

**CONSUMPTION OF UNREGULATED DRINKING WATER  
AND HUMAN HEALTH RISK IN RURAL COMMUNITIES**

A Thesis Submitted to the College of Graduate and Postdoctoral Studies  
In Partial Fulfillment of the Requirements  
For the Degree of Masters of Environment and Sustainability  
In the School of Environment and Sustainability  
University of Saskatchewan  
Saskatoon

By

Lorelei Lynne Ford

## Permission to Use Statement

In presenting this thesis in partial fulfillment of the requirements for a Postgraduate degree from the University of Saskatchewan, I agree that the Libraries of this University may make it freely available for inspection. I further agree that permission for copying of this thesis in any manner, in whole or in part, for scholarly purposes may be granted by the professor who supervised my thesis work or, in their absence, by the Head of the Department or the Dean of the College in which my thesis work was done. It is understood that any copying or publication or use of this thesis or parts thereof for financial gain shall not be allowed without my written permission. It is also understood that due recognition shall be given to me and to the University of Saskatchewan in any scholarly use which may be made of any material in my thesis.

Requests for permission to copy or to make other uses of materials in this thesis in whole or part should be addressed to:

Head of the School of Environment and Sustainability  
University of Saskatchewan  
Room 323 Kirk Hall, 117 Science Place  
Saskatoon, Saskatchewan S7N 5C8 Canada

College of Graduate and Postdoctoral Studies  
University of Saskatchewan  
Room 116 Thorvaldson Building, 110 Science Place  
Saskatoon, Saskatchewan S7N 5C9 Canada

## Abstract

Establishing safe drinking water for rural populations dependent on unregulated water is a global challenge. Despite initiatives to improve access to drinking water, hazards associated with unregulated sources pose a potential risk to human health for rural populations. In the absence of accurate information and monitoring of water quality, consumers form heuristic perceptions of risk associated with their drinking water. Risk perception affects water consumption contributing to uncertainty in risk exposure. Quantifying risk through human health risk assessments (HHRA) has been implemented since the 1940s and advances in risk assessment modeling have created an opportunity to improve HHRA by applying probabilistic Bayesian risk assessment methods. A holistic HHRA integrating risk perception, as it relates to exposure, can quantify uncertainty and provide feedback to improve risk communication and management. The literature lacks a review or summary that characterizes the type and frequency of HHRAs applied to rural populations dependent on unregulated drinking water. The purpose of this thesis is to: (1) summarize studies with HHRA methods applied to unregulated drinking water and rural communities, and describe the characteristics of methods, publications, and current literature gaps; and, (2) characterize and quantify risk perception as it relates to unregulated groundwater wells, and determine the impact of risk perception on human health risk using a holistic HHRA.

A systematic scoping review of peer-reviewed literature (Jan 2000 to May 2014) was used to identify studies with HHRAs applied to unregulated or unspecified drinking water. At least one drinking water source was identified as unregulated (21%) or unspecified (79%) in 100 studies, and 7% identified rural communities dependent on unregulated drinking water. No studies integrated non-traditional factors (e.g. risk perception) into a holistic HHRA. HHRAs applied to rural populations dependent on unregulated water are poorly represented in the literature even though almost half of the global population is rural. The scoping review confirmed a lack of HHRA studies addressing unregulated drinking water risks, and the absence of applied methods that facilitated the quantification and integration of non-traditional factors.

Based on the review findings, a community-based participatory observational case study and holistic HHRA was applied using arsenic concentrations and survey responses from two communities dependent on unregulated groundwater wells. Risk perception and human health risk was determined using probabilistic (Bayesian) risk assessment methods. Community tap water quality exceeded at least one health standard at a rate of 67% and 56%. Households in RM184 or with an annual income > \$50,000 were most likely to have in-house water treatment. Integration of risk perception did not change the overall risk status but lowered the cancer risk for arsenic by 3% for both communities. The probability of exposure to arsenic concentrations over 1:100,000 negligible cancer risk for the two communities was 23% and 22%. This study achieved a holistic Bayesian risk assessment through the integration of risk perception and provided a probability of risk that can be used to inform risk communication and management specific to the participating communities.

## Acknowledgements

I would like to thank Dr. Lalita Bharadwaj and my Advisory Committee for their support and guidance throughout my masters. Dr. Bharadwaj and Dr. Cheryl Waldner have become my most valuable mentors and have allowed me to accomplish my own goals with the support of a knowledgeable research team. I thank Dr. Phillip Bailey, Roy Romanow Provincial Laboratory, for his support of my project and for funding the water quality analysis. I acknowledge the Saskatchewan Health Research Foundation for their financial support dedicated to community-based risk assessment and drinking water, and the support for tuition and books provided by the Water Security Agency. I am very thankful for the data collection and management carried out by Shea Shirley and Karlee McLaughlin, and Dr. Liane McLeod's collaboration, advice, friendship, and support. Thanks to my husband and sons for understanding each time I said 'Mommy has to do her school work'. So much appreciation goes to my parents who have consistently supported me and have helped me form the foundation of my confidence.

I would like to extend my most sincere thanks to the communities that participated in the research. The Lower Qu'Appelle Watershed Stewards, the Rural Municipality of Grayson, and private landowners within the RM were a pleasure to work with. I thank the Chief and Council, the Band, and community members of Beardy's and Okemasis First Nation for their trust, support, and participation. Much appreciation goes to Alfred Gamble who has shown exemplary strength fighting for the well-being of his community. Alfred has inspired me with the empathy and determination required when conducting research with and for communities.

## Table of Contents

Permission to Use Statement.....	i
Abstract .....	ii
Acknowledgements .....	iii
List of Tables.....	vii
List of Figures .....	viii
1 Introduction and Literature Review .....	1
1.1 Introduction .....	1
1.2 Literature Review .....	2
1.2.1 Rural Populations and Unregulated Drinking Water .....	2
1.2.2 Paradigm Shift from Deterministic Risk Assessment .....	3
1.2.3 Probabilistic Bayesian Methods and Holistic Risk Assessment .....	5
1.2.4 Drinking Water Risk Perception .....	7
1.3 Research Opportunity .....	9
1.4 Research Purpose and Objectives.....	9
1.4.1 Characterizing Methods and Approaches of Human Health Risk Assessment Applied to Rural Communities Dependent on Unregulated Drinking Water Sources .....	9
1.4.2 Using Probabilistic Bayesian Human Health Risk Assessment to Quantify and Determine the Impact of Risk Perception on Human Health Risk .....	10
1.5 References .....	12
2 Human Health Risk Assessment Applied to Unregulated Drinking Water Sources: A Scoping Review.....	18
2.1 Abstract.....	19
2.2 Introduction .....	19
2.3 Methods .....	21
2.3.1 Research Question.....	21
2.3.2 Data Sources and Search Strategy.....	21
2.3.3 Citation Management .....	22
2.3.4 Eligibility Criteria .....	22
2.3.5 Title and Abstract Relevance Screening .....	23
2.3.6 Data Characterization .....	23

2.3.7	Data Summary and Synthesis.....	23
2.4	Results .....	24
2.4.1	Search and Selection .....	24
2.4.2	Human Health Risk Assessment Characteristics .....	24
2.4.3	Literature Characteristics .....	29
2.4.4	Literature Gaps .....	33
2.5	Discussion.....	34
2.5.1	Strengths and Limitations.....	39
2.6	Conclusion .....	39
2.7	Recommendations .....	40
2.8	References .....	41
3	Risk Perception and Human Health Risk in Rural Communities Consuming Unregulated Well Water in Saskatchewan, Canada.....	52
3.1	Abstract.....	53
3.2	Introduction .....	54
3.3	Purpose and Objectives .....	55
3.4	Methods .....	55
3.4.1	Research Approach and Community Partnerships.....	55
3.4.2	Study Area and Potential Drinking Water Hazards .....	56
3.4.3	Community Participant Recruitment and Survey Data Collection .....	56
3.4.4	Hazard Data Collection .....	58
3.4.5	Human Health Risk Assessment .....	58
3.4.6	Holistic Bayesian Human Health Risk Assessment.....	60
3.4.7	Data Management and Analysis.....	61
3.5	Results .....	65
3.5.1	Community Profile and Study Survey .....	65
3.5.2	Water Quality and Hazard Identification .....	66
3.5.3	Characterization of Risk Perception and Probability of Using Tap Water for Drinking .....	67
3.5.4	Risk Characterization .....	68
3.6	Discussion.....	71

3.6.1	Holistic Human Health Risk Assessment.....	71
3.6.2	Risk Perception and Exposure.....	72
3.6.3	Unregulated Drinking Water Risks .....	73
3.6.4	Risk Communication and Management .....	74
3.6.5	Limitations and Uncertainty.....	75
3.7	Conclusions .....	76
3.8	References .....	77
4	Conclusion .....	84
4.1	Introduction .....	84
4.2	Future Research .....	86
4.3	Limitations.....	86
4.4	Conclusion.....	87
4.5	References .....	88
5	Supplementary Materials .....	91
5.1	Database Search Terms and Results.....	91
5.2	Full-Text Review Categorization .....	93
5.2.1	References .....	99
5.3	Household Survey.....	100
5.4	Example of Model Code for Bayesian Human Health Risk Assessment.....	101
5.5	Community Tap Water Excursions .....	103

## List of Tables

Table 1.1 Comparison of deterministic and probabilistic human health risk assessment methods.	4
Table 1.2 Advantages and disadvantages of Bayesian risk assessment.	6
Table 2.1 Scoping review inclusion criteria to identify human health risk assessments applied to unregulated or unspecified drinking water.	22
Table 2.2 Human health risk assessment characteristics from scoping review literature ( $n = 100$ ).	25
Table 2.3 Literature characteristics from scoping review ( $n = 100$ ).	30
Table 2.4 Description and references for research, management, and community gaps identified in the scoping review literature ( $n = 67$ ).	33
Table 3.1 Parameters, data sources and descriptions for holistic Bayesian human health risk assessment (HHRA).	63
Table 3.2 Priors on HHRA model parameters.	64
Table 3.3 Characteristics of the community and study population informed by the 2014 <i>Saskatchewan Health Survey</i> , the communities, and the study survey.	66
Table 3.4 Descriptive statistics for community arsenic concentrations ( $\mu\text{g/L}$ ) in household tap water.	67
Table 3.5 The probability of community water use for drinking given household perception it was safe or not safe.	68
Table 3.6 Incremental lifetime cancer risk and probability of risk exceeding Health Canada's negligible cancer risk of 1:100,000 for adults consuming tap water from communities of RM184 and BOFN.	68



## List of Figures

Figure 1.1 Diagram of potential factors affecting risk management of regulated and unregulated drinking water. ....	11
Figure 2.1 PRISMA flowchart of scoping review process. PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analysis).....	24
Figure 2.2 Number of scoping review studies by world region. ....	31
Figure 2.3 Scoping review studies by sector and year. Sectors are not mutually exclusive. ....	32
Figure 3.1 Participating rural communities include Beardy's and Okemasis First Nation and Rural Municipality of Grayson #184 in proximity to the major urban cities of Prince Albert, Saskatoon and Regina in the province of Saskatchewan, Canada. ....	57
Figure 3.2 Directed acyclic graph for the holistic Bayesian human health risk assessment (HHRA) model to determine the incremental lifetime cancer risk (ILCR) from arsenic exposure for adults consuming unregulated well water, according to Health Canada (HC 2010a). ....	62
Figure 3.3 Cumulative probability distribution indicating the probability that the incremental lifetime cancer risk (ILCR) from arsenic was exceeded in drinking water for the Rural Municipality of Grayson #184 (RM184) and Beardy's and Okemasis First Nation (BOFN). ....	70

# 1 Introduction and Literature Review

## 1.1 Introduction

Access to safe drinking water is not a universal human right; however, in 2010, the United Nations and their members adopted *The Human Right to Water and Sanitation - 64/292* resolution based on their concerns ‘that approximately 884 million people lacked access to safe drinking water...’ (UN 2010). The United Nations considers safe drinking water a component of their commitment to the promotion and protection of all human rights, and supports universal access to water through the Millennium Development Goals (MDG). From 2000 to 2015, the United Nations’ Millennium Development Goal was to reduce, by half, the number of the world’s population without ‘sustainable access to safe drinking water (WHO/UNICEF 2015). However, reporting accuracy for this MDG has been challenging with respect to water quality. For example, establishing access does not guarantee the safety of the drinking water sources which may have poor quality due natural contamination or insufficient water management (Shaheed et al. 2014; Wescoat, Headington, and Theobald 2007). Water sources lacking oversight with regard to monitoring and management within a regulatory context are considered unregulated. Given the inability to provide timely and ongoing data on the water quality associated with the water sources identified in the MDG implies that even improved water sources are likely unregulated. In the absence of water regulations or effective management of unregulated water supplies, there exists a knowledge gap where consumption of water is subject to human risk perception (Shaheed et al. 2014; Martz 1983; Maxwell et al. 1998; Hynds, Misstear, and Gill 2013; Charrois 2010). This thesis summarizes the current research on applied human health risk assessment (HHRA) methods associated with unregulated drinking water in rural communities, and characterizes risk perception associated with drinking water to develop a community-based holistic HHRA.

## 1.2 Literature Review

This literature review details research in the areas of: rural populations and unregulated drinking water; the paradigm shift in applied human health risk assessment methods; the benefits of probabilistic Bayesian risk assessment; and, risk perception as it relates to drinking water. These areas of research have coalesced as the methods and approaches of risk assessment have changed over time, thus creating new opportunities to integrate and improve the analysis of risk. The theories and conclusions of researchers in risk assessment, mathematics, statistics, computer science, psychology, sociology, economics, and epidemiology, outside the discipline of human health toxicology, have much to contribute to the evolution of the methods and approaches of HHRA. Broadening the research in this area is necessary to develop a holistic HHRA approach through the integration of non-traditional factors (e.g. risk perception) to improve the determination of risk that better informs risk communication for the management of unregulated drinking water in rural communities globally. To accomplish this, it is imperative to determine the recent trends in HHRA approaches and methods, and explore human perception in an effort to understand how it influences exposure and risk.

### 1.2.1 Rural Populations and Unregulated Drinking Water

In 2015, the World Bank identified 46% (3.38 billion) of the world's population as rural, and determined that 15% of that population lacked adequate access to water (2015). The *25 Year Progress on Sanitation and Drinking Water* report on the Millennium Development Goal 7 reported that 84% of global rural populations had improved drinking water sources; however, 80% of those lacking adequate access to improved water sources were rural residents (WHO/UNICEF 2015). Highly variable by region, rural populations are more likely to be dependent on unimproved surface water sources (WHO/UNICEF 2015). Though rural populations have experienced a 15% increase in access to improved water sources since 1990, these sources are defined as improved only due to their resistance to contamination and not the quality of the water in comparison to the previously accessed source (WHO/UNICEF 2015). For example, a bored well would be considered an improved source when compared to surface water due to the decreased risk of bacterial contamination associated with groundwater wells; however, the well may be naturally contaminated with arsenic. Therefore, improved access through the use of groundwater wells may be an improvement; however, the quality of the water may still pose human health risks for rural populations (Shaheed et al. 2014).

Without sufficient water testing and mitigation of drinking water risks, rural populations are vulnerable to increased health risks associated with drinking water hazards (Shaheed et al. 2014; WHO/UNICEF 2012; WHO 2013). Recognizing the importance of water quality associated with improved drinking water for human health, the WHO/UNICEF *Joint Monitoring Programme (JMP) for Water Supply and Sanitation* initiated a water quality monitoring program in 2010 (WHO/UNICEF 2013). However, there are limitations to the JMP water quality monitoring

program including a lack of epidemiological data to determine the health risks associated with water storage and intermittent end of pipe service, and water quality testing restricted to *E.coli*, arsenic, and fluoride (Shaheed et al. 2014; WHO/UNICEF 2013). In addition, inconsistent or ‘one off’ sampling of individual, private and unregulated water sources limits the temporal interpretation of risk.

For example, Canada is recognized as a developed country with 100% access to water for its citizens (WHO/UNICEF 2015); however, rural populations, including First Nations, are exposed to unregulated water sources that can pose a risk to human health when there is a lack of education, monitoring, and effective treatment of individual and private wells (Charrois 2010; Spence and Walters 2012; Corkal, Schutzman, and Hilliard 2004; Jones et al. 2005). Establishing safe drinking water for rural populations may also be hindered by a lack of resources (e.g. financial) and increased vulnerability (e.g. poverty, illness, minority status, etc.), making it difficult to cope with the responsibility of drinking water management (Nsiah-Kumi 2008; Wescoat, Headington, and Theobald 2007; Zheng and Ayotte 2015). Globally, rural populations face similar challenges when attempting to achieve access to safe drinking water regardless of how they are defined regionally (WHO/UNICEF 2015). Therefore, researching rural communities in specific regions (e.g. Canada) may provide insight and information that is transferrable to rural populations throughout the globe.

### 1.2.2 Paradigm Shift from Deterministic Risk Assessment

In 1999, Roger O. McClellan, a distinguished toxicologist in HHRA, provided a keynote speech which outlined the history and development of human health risk assessment starting with the earliest research on radiation conducted by Cantril and Parker in 1945. His historical summary of HHRA identifies a landmark decade, from 1960 to 1970, in which the US EPA and several national environmental, health, and toxicological institutes were developed. Organizations such as these throughout the world continue to provide the structure for the development and standardization of frameworks and guidelines on human health risk assessment that are applied by the public and academia (e.g. HC 2010; US EPA 2015; WHO/PCS 2001).

In 2001, the World Health Organization (WHO) and International Programme on Chemical Safety (IPCS) advanced the scope of human and environmental risk assessments by providing the *Framework for the Integration of Health and Ecological Risk Assessment*. This framework defined integrated risk assessment as ‘a science-based approach that combines the processes of risk estimation for humans, biota, and natural resources in one assessment’ but did not identify any particular method by which the risk assessment should be carried out. Following the introduction of the IRA framework, publications comparing traditional (deterministic) and IRA methods supported a need to shift towards an integrated approach often with the use of probabilistic methods (Bridges 2003; Sekizawa and Tanabe 2005; Suter II et al. 2005; Ryan

2003). Table 1.1 provides a comparison of the traditional deterministic approach of HHRA to a probabilistic approach as summarized by Richardson (1996) and US EPA (2014).

Table 1.1 Comparison of deterministic and probabilistic human health risk assessment methods.

Deterministic	Probabilistic
<ul style="list-style-type: none"> <li>Provides a single point estimate of individual risk (e.g. 90<sup>th</sup> percentile).</li> <li>Commonly applies average or typical exposure values but has historically used ‘worst-case’ estimates of exposure contributing to over-estimation of risk.</li> <li>Typically applied and supported by regulatory agencies.</li> <li>Cannot integrate non-traditional data and requires quantitative data.</li> <li>Population exposure is only interpreted as above or below a threshold.</li> <li>Manipulation of parameters continues to yield results above or below a threshold.</li> <li>Limited to interpretation of risk relative to the average or worst-case estimates</li> <li>Use of inaccurate point estimate data (e.g. average, maximum) can yield inaccurate results.</li> <li>Provides a cost effective and timely estimation of risk with minimal resources.</li> </ul>	<ul style="list-style-type: none"> <li>Provides an estimate of the potential or probable risk for an individual or community.</li> <li>More likely to provide ‘realistic’ estimates of exposure and risk, and less prone to over-estimation.</li> <li>Typically applied by the private sector; however, regulatory agencies have included it as an advanced or upper tiered method.</li> <li>Allows non-traditional or incomplete data to be characterized and integrated.</li> <li>Provides the proportion of the probability density function exceeding a threshold.</li> <li>Manipulation of parameter inputs and re-running the model can be used to assess options for risk management or prioritize research.</li> <li>Possible to quantify uncertainty and measure model reliability.</li> <li>Use of uncertain and incomplete data or assigning inaccurate probability density functions can yield inaccurate results.</li> <li>May require additional resources and time to develop.</li> </ul>

Support for integrated risk assessment was apparent and the academic research provided feedback on the costs and benefits of its implementation. In his research, Bridges (2003) concluded a paradigm shift from traditional risk assessment was necessary to meet the demand for the quantification of uncertainty and increased transparency. Almost 10 years later researchers were still discussing the implementation of probabilistic methods. For example, Liu et al. (2012) noted that integrated risk assessment can require higher initial resource investment than its traditional counterpart; however, the product may reduce future costs associated with poor decision-making and negative impacts on human health. With the benefits clear, one

wonders if the lack skilled researchers in integrated risk assessment and probabilistic methods remains a limitation on the application as suggested by Bridges (2003).

Despite the benefits of integrated risk assessment acknowledged by a number of groups (Bridges 2003; Ryan 2003; Sekizawa and Tanabe 2005; Suter et al. 2003, 2005; Vermeire et al. 2007), there remains an ongoing need for consistency in the use of terminology and the application of integrated risk assessments (Wilks et al. 2015). The concepts of integrated risk assessment can be applied in the context of community-based risk assessments, taking integrated risk assessment in the direction of tailored assessments specifically responding to the needs of different communities (Wilks et al. 2015). Within the context of rural communities and unregulated drinking water addressed in this thesis, there has been no review and summary of the literature as it relates to community-based approaches and applied integrated risk assessment methods.

### 1.2.3 Probabilistic Bayesian Methods and Holistic Risk Assessment

Though limited, last decade in risk assessment has seen an increase in the application of probabilistic methods (US EPA 2015b), and a desire to integrate data from alternative sources to support holistic risk assessments that include non-traditional factors (e.g. economic, social, and human behaviour variables; Ryan 2003; Wilks et al. 2015). The traditional approach using deterministic methods of HHRA provide ‘...a single point estimate of “individual” risk’, while probabilistic methods can estimate ‘the range of probable risk across a population’ (Richardson 1996). In 1996, Richardson used the Health Canada HHRA framework to compare deterministic and probabilistic methods and concluded that both methods can produce similar results except where the probabilistic methods allow for better characterization of exposure data.

Richardson (1996) also noted that the probability density function produced by probabilistic risk assessment could be used to estimate the proportion of the population exceeding a specified reference dose. This advantage of probabilistic methods provides an indication of how HHRA can be holistic through the inclusion of non-traditional variables that influence risk. For example, research by Doria (2010), Jones et al. (2006), and Spence and Walters (2012) suggests that perception of drinking water influences consumption and exposure which can improve the accuracy of risk when integrated in HHRA. Therefore, the development and standardization of probabilistic methods provides an opportunity to improve the determination of risk, quantify uncertainty, and provide feedback to support risk management.

In 2012, Liu et al. produced evidence that Bayesian belief networks, using conditional probabilities, could better describe mortality and morbidity rates while reducing the over-estimation of risk and additive uncertainty produced by traditional HHRA. Although studies have applied probabilistic Bayesian analysis in risk assessment (Serre et al. 2003; McCann, Marcot, and Ellis 2006; Uusitalo 2007; Sahmel et al. 2010b; Chowdhury, Champagne, and McLellan 2009; Liu et al. 2012; Schmidt et al. 2013), there remains challenges when putting it

into practice. Table 1.2 provides the advantages and disadvantages of probabilistic Bayesian analysis in the context of environmental management or risk assessment as summarized by Liu et al. (2012), McCann et al. (2006), Sahmel et al. (2010), and Uusitalo (2007).

Table 1.2 Advantages and disadvantages of Bayesian risk assessment.

<b>Advantages</b>
<ul style="list-style-type: none"> <li>• Can use historical data, expert judgement or a combination of data sources.</li> <li>• Produces graphical representation which is easy to create, revise and communicate knowledge.</li> <li>• Probability distributions over decision options can enable managers to make reasonable decisions.</li> <li>• Some software has ability to facilitate model construction.</li> <li>• Some software conducts sensitivity analysis.</li> <li>• Backwards inference can be made to determine the most causal conditions for a given outcome.</li> <li>• Characterizes variability inherent in the parameters used for the exposure reconstruction and uncertainty.</li> <li>• The range and likelihood of expected exposure decreases potential for exposure misclassification</li> <li>• Qualitative and quantitative data can be used in exposure reconstruction</li> <li>• Suitable for small and incomplete data sets</li> </ul>
<b>Disadvantages</b>
<ul style="list-style-type: none"> <li>• Model construction (e.g. conditional probability distribution tables) can be challenging to implement.</li> <li>• Temporal dynamics are not well represented.</li> <li>• Precision may be undermined if continuous variables are discretized.</li> <li>• Unable to handle feedback loops within the model.</li> <li>• Availability of relevant epidemiological studies can increase the number of assumptions that need to be made.</li> <li>• Development of appropriate distributions for exposure reconstruction can be a challenge.</li> <li>• Use of probabilistic techniques is common in the environmental and engineering fields but is still rare in specific areas of risk assessment.</li> </ul>

Probabilistic methods, such as Monte Carlo techniques paired with Bayesian methods, are considered to be the next step to improving risk assessment (Sahmel et al. 2010). In their recommended framework for exposure reconstruction in occupational HHRA, Sahmel et al. (2010) provide examples where both techniques are used to address data gaps, and characterize uncertainty and variability. Zargar et al. (2014) highlight the importance of the information located at the tail-end of the probability distributions, and the need to characterize uncertainty in HHRA. This flexibility allows for a data fusion approach which requires a structure, described as ‘architecture’ by Zargar et al. (2014), and goals which suit the unique circumstances of the

HHRA (Esteban et al. 2005). It is important to point out that Zargar et al. (2014) consider integrated HHRA to be the melding or fusion of data for the purpose of more informed risk assessment. For example, they point to the large volume and vast amount of data that require methods of HHRA that can integrate data from multiple sources to improve decision-making. To meet these needs, currently and in the future, data fusion combining multiple data sources can decrease variability and uncertainty in the data (Dasarathy 1997), and provide a mathematical way to simplify data from multiple sources (Zargar et al. 2014). For the purpose of this thesis the integration of data results in a ‘holistic’ HHRA, rather than simply integrated, because it supports the inclusion of a new data type that is not traditionally used in HHRA (i.e. non-traditional factors – risk perception).

Although Bayesian methods are applied more frequently in the field of environmental risk assessment (Liu et al. 2012), there are examples where probabilistic and Bayesian techniques have been applied to HHRA (Serre et al. 2003; Chowdhury, Champagne, and McLellan 2009; Liu et al. 2012; Schmidt et al. 2013; Ramachandran 2001). These methods can allow for the integration of uncertain or qualitative data that supports or compliments quantitative data. The integrated data can then contribute to a confidence interval that assists public health planning and policy (Serre et al. 2003). For this thesis, the use of Monte Carlo and Bayesian techniques in HHRA presents an opportunity to quantify uncertainty and integrate qualitative information (e.g. non-traditional factor - risk perception). In turn these methods will improve the accuracy of HHRA for a holistic view of risk that can be applied to inform risk communication and management for rural communities dependent on unregulated drinking water.

#### 1.2.4 Drinking Water Risk Perception

Risk perception associated with drinking water affects water consumption but does not necessarily correlate with the safety of drinking water for human consumption (Martz 1983; Maxwell et al. 1998; Hynds, Misstear, and Gill 2013; Patrick 2011; Orgill et al. 2013; Chen et al. 2012). If perception of drinking water is inaccurate then the opportunity exists to over-use non-potable water and under-use potable water. Similarly, avoidance of drinking water may encourage consumption of higher risk water sources or sugary beverages (Onufrak et al. 2012; Onufrak et al. 2014; Dupont, Adamowicz, and Krupnick 2010). For example, Bogart et al. (2013) studied the perceptions of youth and their parents as it related to sugar sweetened beverages and tap water and found that 49% of parents and youth had similar perceptions about their tap water; however, 71% of those in agreement on their perceptions thought the water had negative effects on their health. In addition, they noted that tap water was perceived as unsafe concurrently with the high consumption of sugar drinks (Bogart et al. 2013). This study shows how a group of individuals may have similar perceptions of water and how that misinformation can negatively affect their health.



Most studies addressing perception of risk and water quality appear to be focused on tap water and customer satisfaction according to Doria (2010). This finding indicates that perception of risk as it relates to unregulated drinking water sources does not attract the attention of water treatment plants and commercial water providers. Doria (2010) states that direct organoleptic experiences with drinking water sources are the primary driver determining perception of drinking water risk, and that most studies in risk perception focus on hazards that the public know little about. In the Canadian context, Charrois (2010) points out that the majority of people with private drinking water wells are not sufficiently educated on the potential health risks associated with consumption of groundwater; therefore, people dependent on private water wells rely on their perception. Jones et al. (2006) uses Canada as an example of how the lack of regulation, testing, monitoring, and treatment associated with private water management may pose a risk to human health; however, under similar circumstances, drinking water risks can exist for any individual or global community dependent on unregulated water.

In the context of unregulated water and the dependence on perception, the many factors affecting drinking water risk perception clearly illustrates the difficulty, from a risk management perspective, in ensuring people's perception of risk is accurate. Doria (2010) provides a thorough discussion of the factors affecting perception of drinking water. For comparison, Figure 1.1 shows a list of factors that support risk management decisions for regulated and unregulated sources. Often based on defensible science and policy, regulated water sources have a multi-barrier approach to protecting human health. In contrast, unregulated water is highly dependent on less quantitative factors as summarized by Doria (2010). Chowdhury et al. (2009) suggests the uncertainty associated with human behaviour may be reduced if behaviour can be integrated as a variable in HHRA.

Given there can be indirect health effects associated with human behaviour (i.e. consumption) in response to poor drinking water (Wescoat, Headington, and Theobald 2007), it is imperative that risk assessment captures individual's or communities' perception of risk. Furthermore, with an increased pressure for governments to include human perceptions/perspectives in water resource management (Jackson 2006), it would be beneficial to determine the impact those perceptions have on exposure and human health risk. Furthermore, understanding the effect of risk perception on human health risk could lead to improved risk management and communication (Markon and Lemyre 2013). As stated by Serre et al. (2003), "Uncertain knowledge obtained about important exposure parameters could be more valuable than the certain knowledge obtained about less important parameters". In other words, the inclusion of risk perception in HHRA may contribute uncertain knowledge for a more holistic and accurate determination of risk.

Doria (2010) provides two examples by Slovic (2000) and Hagerty (2003) that suggest that the public's perception of water quality may be decreasing over time due to "inter-temporal

pessimism”. If this is an accurate prediction, the perception of risk associated with drinking water quality and its importance in the determination of risk may be increasing. By studying the impact of risk perception on unregulated water consumption and human health risk we can provide insight on the exposure and risk outcomes that are required for risk communication and management to establish safe drinking water for rural populations that distrust their water.

### 1.3 Research Opportunity

This research contributes to a larger Saskatchewan Health Research Foundation (SHRF) grant objective to “Use community-based risk assessment to characterize challenges related to poor drinking water quality in Saskatchewan not included in current surveillance initiatives.” To achieve the SHRF objective, this research conducts an observational case-study with two communities and assesses risk perception as it relates to the use of unregulated water wells in Saskatchewan. Bayesian HHRA methods provide an alternative approach to risk assessment that is not frequently used in Saskatchewan. In the global context, this research can provide valuable knowledge to rural populations dependent on unregulated drinking water by integrating qualitative data into HHRA to improve risk management in the absence of drinking water regulation. On a global scale, the research supports a global goal to increase access to safe drinking water sources for rural communities by showing the importance of integrating non-traditional variables impacting exposure and risk.

### 1.4 Research Purpose and Objectives

The purpose of this thesis is to determine the methods of HHRA currently applied to rural communities that are dependent on unregulated drinking water and, using this information, integrating risk perception of drinking water into a holistic HHRA to support the improvement of risk communication and management. The objectives of the research are to:

- 1) Conduct a scoping review to characterize the methods and approaches of HHRA applied to rural communities dependent on unregulated drinking water sources.
- 2) Quantify risk perception, using probabilistic Bayesian HHRA, to determine its impact on human health risk.

#### 1.4.1 Characterizing Methods and Approaches of Human Health Risk Assessment Applied to Rural Communities Dependent on Unregulated Drinking Water Sources

Chapter 2 provides a summary of the current literature characteristics that summarize the methods and approaches of human health risk assessment applied to rural communities dependent on unregulated drinking water sources. Eligible peer-reviewed literature identified those studies applying a drinking water HHRA to communities dependent on an unregulated or unspecified water source. Characteristics of studies were summarized and primary areas of interest included: the frequency of HHRA applied to rural communities; the application of

deterministic and probabilistic methods; and the integration of non-traditional data into the quantitative risk assessment. The results of this study identified a lack of applied HHRA to rural communities dependent on unregulated water sources, and the limited use of methods that could facilitate the integration of non-traditional data. This scoping review provides a valuable summary of the literature to researchers, regulatory agencies, and organizations that can use the information to inform future HHRA application, approach, and reporting.

#### 1.4.2 Using Probabilistic Bayesian Human Health Risk Assessment to Quantify and Determine the Impact of Risk Perception on Human Health Risk

Supported by the conclusions of the scoping review, Chapter 3 takes a holistic HHRA approach to determine the impact of drinking water risk perception on the lifetime incremental cancer risk due to the presence of arsenic in the drinking water of two rural communities in Saskatchewan, Canada. The need to explore new approaches and tools (i.e. holistic HHRA, and Bayesian risk assessment methods) support the desire to include data of different types and quality in risk assessment. Demonstrated for the first time in the context of unregulated drinking water, this study allows for qualitative risk perception data to be quantified and integrated into a quantitative risk assessment to decrease uncertainty associated with exposure and risk. In addition, the results of the risk assessment provide the communities with a better understanding of the discrepancy between their perception and the safety of their drinking water.

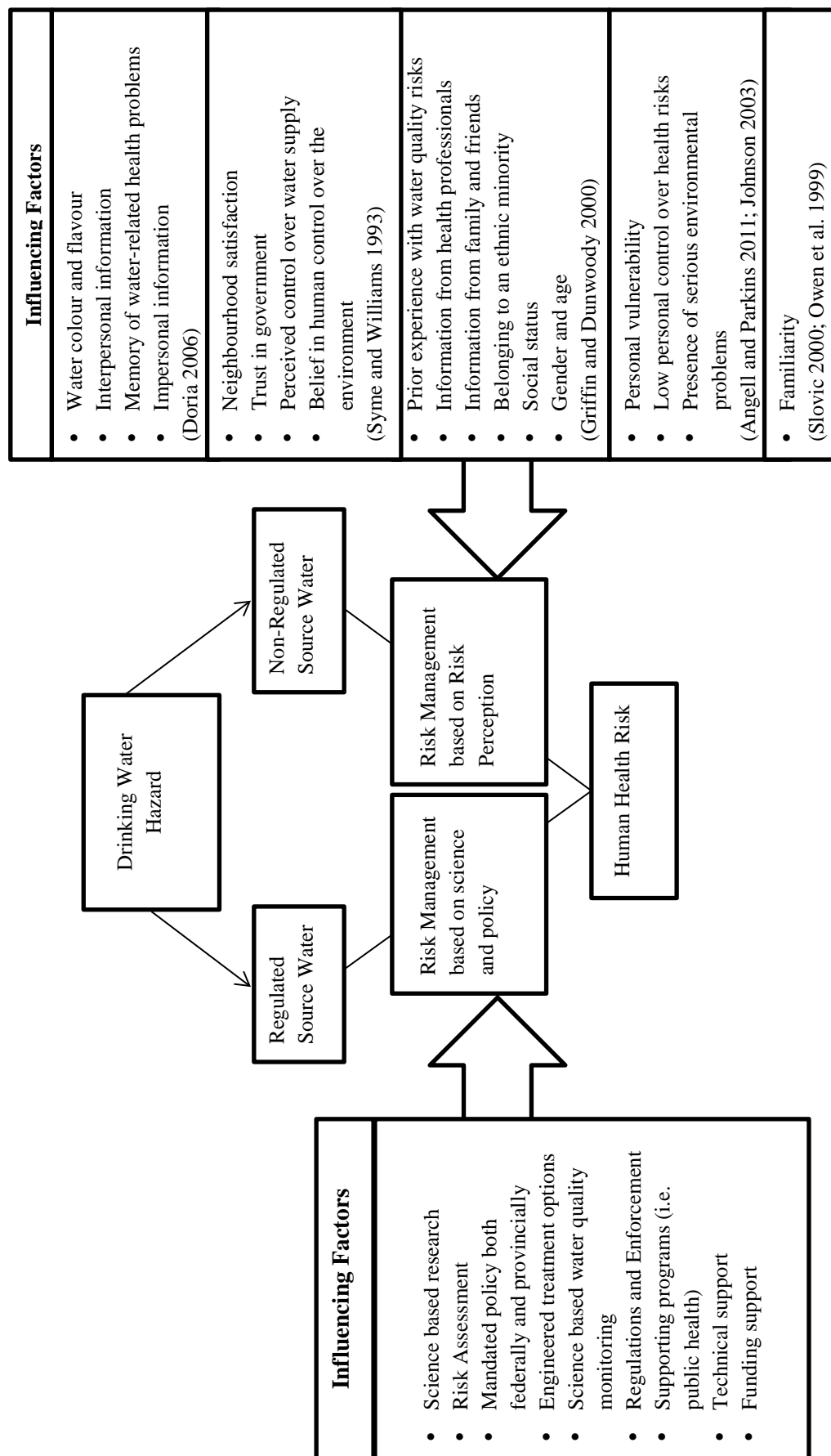


Figure 1.1 Diagram of potential factors affecting risk management of regulated and unregulated drinking water.

## 1.5 References

- Angell, Angela C., and John R. Parkins. 2011. "Resource Development and Aboriginal Culture in the Canadian North." *Polar Record* 47 (1): 67–79. doi:10.1017/S0032247410000124.
- Bogart, Laura M., Burton O. Cowgill, Andrea J. Sharma, Kimberly Uyeda, Laurel A. Sticklor, Katie E. Alijewicz, and Mark A. Schuster. 2013. "Parental and Home Environmental Facilitators of Sugar-Sweetened Beverage Consumption among Overweight and Obese Latino Youth." *Academic Pediatrics* 13 (4). Elsevier Ltd: 348–55. doi:10.1016/j.acap.2013.02.009.
- Bridges, Jim. 2003. "Human Health and Environmental Risk Assessment: The Need for a More Harmonised and Integrated Approach." *Chemosphere* 52 (9): 1347–51. doi:10.1016/S0045-6535(03)00469-7.
- Charrois, J. W. A. 2010. "Private Drinking Water Supplies: Challenges for Public Health." *Canadian Medical Association Journal* 182 (10): 1061–64. doi:10.1503/cmaj.090956.
- Chen, Hanyi, Yaying Zhang, Linlin Ma, Fangmin Liu, Weiwei Zheng, Qinfeng Shen, Hongmei Zhang, et al. 2012. "Change of Water Consumption and Its Potential Influential Factors in Shanghai: A Cross-Sectional Study." *BMC Public Health* 12 (1). BMC Public Health: 450. doi:10.1186/1471-2458-12-450.
- Chowdhury, Shakhawat, Pascale Champagne, and P. James McLellan. 2009. "Uncertainty Characterization Approaches for Risk Assessment of DBPs in Drinking Water: A Review." *Journal of Environmental Management* 90 (5). Elsevier Ltd: 1680–91. doi:10.1016/j.jenvman.2008.12.014.
- Corkal, Darrell, W. C. Schutzman, and Clint R. Hilliard. 2004. "Rural Water Safety From the On-Farm Tap." *Journal of Toxicology and Environmental Health, Part A* 67 (20–22): 1619–42. doi:10.1080/15287390490491918.
- Dasarathy, Belur V. 1997. "Sensor Fusion Potential Exploitation: Innovative Architectures and Illustrative Approaches." *Proceedings of the IEEE* 85 (1): 24–38.
- Doria, Miguel de França. 2006. "Bottled Water versus Tap Water: Understanding Consumers-Preferences." *Journal of Water and Health* 4 (2): 271–76. doi:10.2166/wh.2006.008.
- . 2010. "Factors Influencing Public Perception of Drinking Water Quality." *Water Policy* 12 (1): 1–19. doi:10.2166/wp.2009.051.
- Dupont, Diane, W. L. Adamowicz, and Alan Krupnick. 2010. "Differences in Water Consumption Choices in Canada: The Role of Socio-Demographics, Experiences, and Perceptions of Health Risks." *Journal of Water and Health* 8 (4): 671–86. doi:10.2166/wh.2010.143.
- Esteban, Jaime, Andrew Starr, Robert Willetts, Paul Hannah, and Peter Bryanston-Cross. 2005. "A Review of Data Fusion Models and Architectures: Towards Engineering Guidelines." *Neural Computing and Applications* 14 (4): 273–81. doi:10.1007/s00521-004-0463-7.

- Griffin, Robert J., and Sharon Dunwoody. 2000. "The Relation of Communication to Risk Judgment and Preventive Behavior Related to Lead in Tap Water." *Health Communication* 12 (1): 81–107. doi:10.1207/S15327027HC1201.
- Hagerty, Michael R. 2003. "Was Life Better in the 'Good Old Days'? Intertemporal Judgments of Life Satisfaction." *Journal of Happiness Studies* 4 (2): 115–39.
- Health Canada (HC). 2010. "Guidance on Human Health Preliminary Quantitative Risk Assessment (PQRA), Version 2.0. Part I." Ottawa, Ont.
- Hynds, Paul D., Bruce D. Misstear, and Laurence W. Gill. 2013. "Unregulated Private Wells in the Republic of Ireland: Consumer Awareness, Source Susceptibility and Protective Actions." *Journal of Environmental Management* 127 (September). Elsevier Ltd: 278–88. doi:10.1016/j.jenvman.2013.05.025.
- Jackson, Sue. 2006. "Compartmentalising Culture: The Articulation and Consideration of Indigenous Values in Water Resource Management." *Australian Geographer* 37 (1): 19–31. doi:10.1080/00049180500511947.
- Johnson, Branden B. 2003. "Do Reports on Drinking Water Quality Affect Customers' Concerns? Experiments in Report Content." *Risk Analysis* 23 (5): 985–98. doi:10.1111/1539-6924.00375.
- Jones, Andria Q., Catherine E. Dewey, Kathryn Doré, Shannon E. Majowicz, Scott A. McEwen, David Waltner-Toews, Mathews Eric, Deborah J. Carr, and Spencer J. Henson. 2006. "Public Perceptions of Drinking Water: A Postal Survey of Residents with Private Water Supplies." *BMC Public Health* 6 (January): 94. doi:10.1186/1471-2458-6-94.
- Jones, Andria Q., Catherine E. Dewey, Kathryn Doré, Shannon E. Majowicz, Scott A. McEwen, David Waltner-Toews, Spencer J. Henson, and Eric Mathews. 2005. "Public Perception of Drinking Water from Private Water Supplies: Focus Group Analyses." *BMC Public Health* 5 (January): 129. doi:10.1186/1471-2458-5-129.
- Liu, Kevin Fong-Rey, Che-Fan Lu, Cheng-Wu Chen, and Yung-Shuen Shen. 2012. "Applying Bayesian Belief Networks to Health Risk Assessment." *Stochastic Environmental Research and Risk Assessment* 26 (3): 451–65. doi:10.1007/s00477-011-0470-z.
- Markon, Marie-Pierre L., and L. Lemyre. 2013. "Public Reactions to Risk Messages Communicating Different Sources of Uncertainty: An Experimental Test." *Human and Ecological Risk Assessment: An International Journal* 19 (4): 1102–26. doi:10.1080/10807039.2012.702015.
- Martz, Diane J. F. 1983. "Variations in Resident Appraisals of Groundwater Quality on Saskatchewan Farms." *Thesis*. University of Saskatchewan.
- Maxwell, Reed M., Susan D. Pelmulder, Andrew F. B. Thompson, and William E. Kastenberg. 1998. "On the Development of a New Methodology for Groundwater-Driven Health Risk Assessment." *Water Resources Research* 34 (4): 833–47.
- McCann, Robert K., Bruce G. Marcot, and Rick Ellis. 2006. "Bayesian Belief Networks:

- Applications in Ecology and Natural Resource Management.” *Canadian Journal of Forest Research* 36 (12): 3053–62. doi:10.1139/x06-238.
- McClellan, Roger O. 1999. “KEYNOTE ADDRESS: HUMAN HEALTH RISK ASSESSMENT: A Historical Overview and Alternative Paths Forward.” *Inhalation Toxicology* 11 (6–7): 477–518. doi:10.1080/089583799196880.
- Nsiah-Kumi, Phyllis A. 2008. “Communicating Effectively with Vulnerable Populations during Water Contamination Events.” *Journal of Water and Health* 6 (S1): s63. doi:10.2166/wh.2008.041.
- Onufrak, Stephen J., Sohyun Park, Joseph R. Sharkey, Caitlin Merlo, Wesley R. Dean, and Bettylou Sherry. 2014. “Perceptions of Tap Water and School Water Fountains and Association With Intake of Plain Water and Sugar-Sweetened Beverages.” *Journal of School Health* 84 (3): 195–204. doi:10.1111/josh.12138.
- Onufrak, Stephen, Sohyun Park, Joseph R. Sharkey, and Bettylou Sherry. 2012. “The Relationship of Perceptions of Tap Water Safety with Intake of Sugar-Sweetened Beverages and Plain Water among US Adults.” *Public Health Nutrition* 17 (1): 1–7. doi:10.1017/S1368980012004600.
- Orgill, Jennifer, Ameer Shaheed, Joe Brown, and Marc Jeuland. 2013. “Water Quality Perceptions and Willingness to Pay for Clean Water in Peri-Urban Cambodian Communities.” *Journal of Water and Health* 11 (3): 489–506. doi:10.2166/wh.2013.212.
- Owen, A. J., J. S. Colbourne, C. R. I. Clayton, and C. Fife-Schaw. 1999. “A Mental Model’s Approach to Customer Perception of Drinking-Water Supply and Quality.” *Journal of Chartered Institution of Water and Environmental Management* 13: 241–44.
- Patrick, Robert J. 2011. “Uneven Access to Safe Drinking Water for First Nations in Canada: Connecting Health and Place through Source Water Protection.” *Health & Place* 17 (1). Elsevier: 386–89. doi:10.1016/j.healthplace.2010.10.005.
- Ramachandran, G. 2001. “Retrospective Exposure Assessment Using Bayesian Methods.” *Annals of Occupational Hygiene* 45 (8): 651–67. doi:10.1093/annhyg/45.8.651.
- Richardson, G. Mark. 1996. “Deterministic Versus Probabilistic Risk Assessment: Strengths and Weaknesses in a Regulatory Context.” *Human and Ecological Risk Assessment: An International Journal* 2 (1): 44–54. doi:10.1080/10807039.1996.10387459.
- Ryan, Louise. 2003. “Epidemiologically Based Environmental Risk Assessment.” *Statistical Science* 18 (4): 466–80. doi:10.1214/ss/1081443230.
- Sahmel, Jennifer, Kathryn Devlin, Dennis Paustenbach, Dana Hollins, and Shannon Gaffney. 2010a. “The Role of Exposure Reconstruction in Occupational Human Health Risk Assessment: Current Methods and a Recommended Framework.” *Critical Reviews in Toxicology* 40 (9): 799–843. doi:10.3109/10408444.2010.501052.
- . 2010b. “The Role of Exposure Reconstruction in Occupational Human Health Risk Assessment: Current Methods and a Recommended Framework.” *Critical Reviews in*

- Toxicology* 40 (9): 799–843. doi:10.3109/10408444.2010.501052.
- Schmidt, P. J., K. D. M. Pinter, A. M. Fazil, C. A. Flemming, M. Lanthier, N. Laprade, M. D. Sunohara, et al. 2013. “Using *Campylobacter* Spp. and *Escherichia Coli* Data and Bayesian Microbial Risk Assessment to Examine Public Health Risks in Agricultural Watersheds under Tile Drainage Management.” *Water Research* 47 (10). Elsevier Ltd: 3255–72. doi:10.1016/j.watres.2013.02.002.
- Sekizawa, Jun, and Shinsuke Tanabe. 2005. “A Comparison between Integrated Risk Assessment and Classical Health/environmental Assessment: Emerging Beneficial Properties.” *Toxicology and Applied Pharmacology* 207 (2): 617–22. doi:10.1016/j.taap.2005.01.047.
- Serre, M. L., A. Kolovos, G. Christakos, and K. Modis. 2003. “An Application of the Holistochastic Human Exposure Methodology to Naturally Occurring Arsenic in Bangladesh Drinking Water.” *Risk Analysis* 23 (3): 515–28. doi:10.1111/1539-6924.t01-1-00332.
- Shaheed, Ameer, Jennifer Orgill, Maggie A. Montgomery, Marc A. Jeuland, and Joe Brown. 2014. “Why ‘improved’ Water Sources Are Not Always Safe.” *Bulletin of the World Health Organization* 92 (4): 283–89. doi:10.2471/BLT.13.119594.
- Slovic, Paul. 2000. *The Perception of Risk*. London, UK: Earthscan Publications.
- Spence, Nicholas, and Dan Walters. 2012. “‘Is It Safe?’ Risk Perception and Drinking Water in a Vulnerable Population ‘Is It Safe?’ Risk Perception and Drinking Water in a Vulnerable.” *The International Indigenous Policy Journal* 3 (3): 1–23. <http://ir.lib.uwo.ca/iipj/vol3/iss3/9>.
- Suter, G. W., W. R. Munns, and J. Sekizawa. 2003. “Types of Integration in Risk Assessment and Management, and Why They Are Needed.” *Human and Ecological Risk Assessment* 9 (1): 273–79. doi:10.1080/713609864.
- Suter II, Glenn W., Theo Vermeire, Wayne R. Munns Jr., and Jun Sekizawa. 2005. “An Integrated Framework for Health and Ecological Risk Assessment.” *Toxicology and Applied Pharmacology* 207 (2): 611–16. doi:10.1016/j.taap.2005.01.051.
- Syme, Geoffrey J., and K. D. Williams. 1993. “The Psychology of Drinking Water Quality: An Exploratory Study.” *Water Resources Research* 29 (12): 4003–10. doi:10.1029/93WR01933.
- United Nations (UN). 2010. “General Assembly.” *General Assembly Resolution A/Res/64/292 - The Human Right to Water and Sanitation*. Vol. Sixty-four. <http://www.un.org/es/comun/docs/?symbol=A/RES/64/292&lang=E>.
- United States Environmental Protection Agency (US EPA). 2014. “Risk Assessment Forum White Paper: Probabilistic Risk Assessment Methods and Case Studies.” Washington, DC. [www.epa.gov/raf](http://www.epa.gov/raf).
- . 2015a. “Human Health Risk Assessment.” <https://www.epa.gov/risk/human-health-risk->



assessment.

- . 2015b. “Probabilistic Risk Assessment White Paper and Supporting Documents.” <https://www.epa.gov/osa/probabilistic-risk-assessment-white-paper-and-supporting-documents>.
- Uusitalo, Laura. 2007. “Advantages and Challenges of Bayesian Networks in Environmental Modelling.” *Ecological Modelling* 203 (3–4): 312–18. doi:10.1016/j.ecolmodel.2006.11.033.
- Vermeire, Theo, Wayne R. Munns, Jun Sekizawa, Glenn Suter, and Glen Van der Kraak. 2007. “An Assessment of Integrated Risk Assessment.” *Human and Ecological Risk Assessment: An International Journal* 13 (2): 339–54. doi:10.1080/10807030701226848.
- Wescoat, James L., Lisa Headington, and Rebecca Theobald. 2007. “Water and Poverty in the United States.” *Geoforum* 38 (5): 801–14. doi:10.1016/j.geoforum.2006.08.007.
- WHO. 2013. “Water Quality and Health Strategy 2013-2020.” New York.
- WHO/IPCS. 2001. “Framework for the Integration of Health and Ecological Risk Assessment.” *Integrated Risk Assessment*. Vol. II. Geneva, Switzerland. doi:WHO/IPCS/IRA/01/12.
- WHO/UNICEF. 2012. “Progress on Drinking Water and Sanitation: 2012 Update.” *New York, NY: United Nation’s Children’s Fund and World Health Organisation*. doi:978-924-1503279.
- . 2013. “Meeting Report.” *Report: Second Meeting of the WHO/UNICEF JMP Task Force on Monitoring Drinking-Water Quality*. Geneva, Switzerland. doi:10.1038/sj.embor.7400207.
- . 2015. “25 Years Progress on Sanitation and Drinking Water: 2015 Update and MDG Assessment.” *New York, NY: United Nation’s Children’s Fund and World Health Organisation*. [http://www.who.int/water\\_sanitation\\_health/publications/jmp-2015-update/en/](http://www.who.int/water_sanitation_health/publications/jmp-2015-update/en/).
- Wilks, M. F., N. Roth, L. Aicher, M. Faust, P. Papadaki, A. Marchis, M. Calliera, et al. 2015. “White Paper on the Promotion of an Integrated Risk Assessment Concept in European Regulatory Frameworks for Chemicals.” *Science of The Total Environment* 521–522 (1). Elsevier B.V.: 211–18. doi:10.1016/j.scitotenv.2015.03.065.
- World Bank. 2015. “The World Bank: Data.” *The World Bank: Data*. <http://data.worldbank.org/indicator/SP.RUR.TOTL.ZS>.
- Zargar, Amin, Roberta Dyck, M. Shafiqul Islam, Asish Mohapatra, and Rehan Sadiq. 2014. “Data Fusion Methods for Human Health Risk Assessment: Review and Application.” *Human and Ecological Risk Assessment: An International Journal* 20 (3): 807–38. doi:10.1080/10807039.2012.746145.
- Zheng, Yan, and Joseph D. Ayotte. 2015. “At the Crossroads: Hazard Assessment and Reduction of Health Risks from Arsenic in Private Well Waters of the Northeastern United States and

Atlantic Canada.” *Science of the Total Environment* 505. Elsevier B.V.: 1237–47.  
doi:10.1016/j.scitotenv.2014.10.089.

## 2 Human Health Risk Assessment Applied to Unregulated Drinking Water Sources: A Scoping Review

This manuscript is published in a Human Impact Assessment special issue in the International Journal of Environmental Research and Public Health on July 28<sup>th</sup>, 2017:

<http://www.mdpi.com/1660-4601/14/8/846> doi:[10.3390/ijerph14080846](https://doi.org/10.3390/ijerph14080846)

Ford, Lorelei, Lalita Bharadwaj, Lianne McLeod, and Cheryl Waldner. 2017. “Human Health Risk Assessment Applied to Rural Populations Dependent on Unregulated Drinking Water Sources: A Scoping Review.” *International Journal of Environmental Research and Public Health* 14 (8): 846. doi: 10.3390/ijerph14080846.

This manuscript reproduced under licence with minor modifications for formatting. Author contributions are as follows: Lorelei Ford participated in the entire process of the research including: design, search, screening, full text review, data extraction, data analysis, interpretation of results, writing, and editing the manuscript. Lianne McLeod participated in the design, search and screening and contributed to editing. Cheryl Waldner and Lalita Bharadwaj participated in the design of research, full text review, data extraction, interpretation of results, and writing/editing of the manuscript. Proportion of full text review for Lorelei Ford, Lalita Bharadwaj, and Cheryl Waldner was 44%, 46%, and 10%, respectively.

## 2.1 Abstract

Safe drinking water is a global challenge for rural populations dependent on unregulated water. A scoping review of research on human health risk assessments (HHRA) applied to this vulnerable population may be used to improve assessments applied by government and researchers. This review aims to summarize and describe the characteristics of HHRA methods, publications, and current literature gaps of HHRA studies on rural populations dependent on unregulated or unspecified drinking water. Peer-reviewed literature was systematically searched (January 2000 to May 2014) and identified at least one drinking water source as unregulated (21%) or unspecified (79%) in 100 studies. Only 7% of reviewed studies identified a rural community dependent on unregulated drinking water. Source water and hazards most frequently cited included groundwater (67%) and chemical water hazards (82%). Most HHRAs (86%) applied deterministic methods with 14% reporting probabilistic and stochastic methods. Publications increased over time with 57% set in Asia, and 47% of studies identified at least one literature gap in the areas of research, risk management, and community exposure. HHRAs applied to rural populations dependent on unregulated water are poorly represented in the literature even though almost half of the global population is rural.

## 2.2 Introduction

In 2015, the World Bank identified 46% (3.38 billion) of the world's population as rural, and determined that 15% of that population lacks adequate access to water (World Bank 2015). In 2000, the Millennium Declaration was signed by the United Nations to establish the Millennium Development Goal (MDG) to reduce, by half, the number of the world's population without 'sustainable access to safe drinking water' (WHO/UNICEF 2015). However, increased access to water does not guarantee water sources are safe for consumption, and without sufficient water testing and mitigation of drinking water risks, rural populations are vulnerable to increased health risks associated with drinking water hazards (Shaheed et al. 2014; WHO/UNICEF 2012; WHO 2013). Global rural populations remain an 'at risk' priority due to: exposure to unknown drinking water hazards; a lack of oversight associated with the use of unregulated water sources; a failure to mitigate known drinking water risks (e.g. avoidance or non-regulated treatment); and, their vulnerability and inequality as it relates to education and financial resources to establish safe water in comparison to urban populations (Fawell and Nieuwenhuijsen 2003; Nsiah-Kumi 2008; WHO/UNICEF 2015). To support the management of the risks to rural communities and to further the field of human health risk assessment (HHRA) it is imperative to understand the research undertaken in this area. To this point, there has not been a review or summary of the research literature that provides the type and frequency of applied HHRA methods to determine the drinking water risks to rural communities dependent on unregulated source water.

Human health risk assessment has been used to quantify risk as it relates to human exposure to potential hazards since the late 1940s. With its origins in environmental risk assessment, HHRA

has since evolved independently from the environmental discipline (Suter II et al. 2005). The fields of human health and environmental risk assessment have not paralleled one another in their development of integrated risk assessment despite similarities in the traditional application of methods (Bridges 2003). In 2003, Bridges hypothesized that the departmental separation of human health and environment by governments; the lack of integrated risk assessment training in universities; and, the requirement for communication and collaboration between disciplines are sources of resistance to the integration of human health and environmental risk assessment. Bridges (2003), Munns et al. (2003), Suter et al. (2005), and Vermeire et al. (2007) have acknowledged the need for guidelines and frameworks to facilitate integrated assessment, and there are examples that have been suggested or developed (Briggs 2008; Sexton and Linder 2014; Suter et al. 2005; WHO/IPCS 2001). However, a recent publication by Wilks et al. (2015) suggests the integration of environment and health risk assessments remains a challenge due to lack of agreement between ‘...terminology, models and methodologies across chemical categories and regulatory agencies...’. In addition to concerns regarding the implementation of integrated risk assessment, Wilks et al. (2015) acknowledge that non-traditional factors such as behaviour, socio-economics, perceptions, and values could improve the determination and management of risk through a more holistic approach.

The terms *integrated* and *holistic* are inconsistently defined as noted by Bridges in 2003. Integrated risk assessment, generally, refers to the inclusion of both human health and environmental risk in one assessment (Bridges 2003; Hart and Pollino 2009; Sekizawa and Tanabe 2005; WHO/IPCS) 2001), while the term *holistic* suggests a systems approach where different data types and sources influencing risk can be utilized (e.g. social, economic, perception, etc.; Arquette et al. 2002; Serre et al. 2003). Adopting a holistic approach using probabilistic and stochastic methods can benefit HHRA by allowing for the use of alternative data sources and types (Bridges 2003; Serre et al. 2003; Zargar et al. 2014) which can increase the accuracy through the quantification of uncertainty (Liu et al. 2012). As mentioned previously, Wilks et al. (2015) suggests that a holistic approach would consider economic, social, cultural, and political factors; however, they do not describe the inclusion of these factors as a data source per se. For the purpose of this scoping review, we define integrated risk assessment according to the WHO/IPCS (2001) as ‘...a science-based approach that combines the processes of risk estimation for humans, biota, and natural resources in one assessment.’ Alternatively, we suggest that a holistic risk assessment would be similar to that described by Arquette et al. (2002) and include non-traditional factors, that may be gathered from qualitative data sources or multiple disciplines, in the determination of risk that is specific and relevant to the humans or environment of concern. A holistic human health risk assessment would be inherently integrated; however, an integrated risk assessment is not necessarily holistic.

Deterministic methods of HHRA have been applied to comply with structured national and international guidance documents and frameworks. Despite studies that identify the benefits of integrated risk assessment (Bridges 2003; Ryan 2003; Sekizawa and Tanabe 2005; Suter II et al.

2005; Liu et al. 2012; Briggs 2008), there has not been a systematic review of application frequency of deterministic, integrated or holistic methods. A scoping review of recent HHRA practices may be used to inform and support the adoption and use of holistic frameworks by government and researchers. This could improve methods and quantify uncertainty, which would support effective risk communication and management (Markon and Lemyre 2013). This paper summarizes HHRA methods used to assess human health risks associated with unregulated drinking water and describes the frequency of HHRA applied to rural communities, the characteristics of methods and publications, and current literature gaps.

## 2.3 Methods

This scoping review involved a multi-disciplinary team of four researchers in the fields of water quality, human health, epidemiology, and toxicology. Analysis and writing remained the responsibility of the lead (Lorelei Ford) with all team members participating in the review process, meetings, and editing. A health sciences research librarian was consulted on the selection of databases and search terms to ensure the identification of relevant studies. The framework chosen for the review was that presented by Pham et al. (2014) which is based on the works of Arksey and O'Malley (2005), and Levac et al. (2010). This review utilized the first five steps of the Arksey and O'Malley (2005) framework, including: identification of the research question; identification of relevant studies; study selection; charting the data; and, collating, summarizing and reporting results.

### 2.3.1 Research Question

The research question had two parts and asked, 'What methods of HHRA have been used to determine the health risks associated with consumption of unregulated drinking water, and how often are they applied within the context of rural communities?'

### 2.3.2 Data Sources and Search Strategy

In January 2014, two researchers (Lorelei Ford and Lianne McLeod), with the assistance of a research librarian at the University of Saskatchewan Health Sciences Library, identified the databases, search terms, and limitations that would define the review. Search databases included ProQuest - Public Health (multidisciplinary), EMBASE – Embase + Embase Classic (biomedical, broad), MEDLINE – Ovid (biomedical, specific), Global Health (global), and Scopus (multidisciplinary, broad). These databases provided comprehensive coverage of a wide range of disciplines as they relate to human health risk assessment. Search terms included: 'risk', 'risk assessment\*' or 'analys\*', 'water', 'groundwater', and 'health'. The search terms 'risk assessment', 'water' and 'groundwater' were expanded to ensure inclusion given the diverse range of terminology for HHRA. The concatenated term 'groundwater' was specifically included because search terms for 'ground' and 'water' returned fewer results. Search terms did not include 'drinking water' because studies using the term were included using the search term

‘water’. Searches were restricted to English language publications between January 1<sup>st</sup>, 2000 and May 8<sup>th</sup>, 2014. The Scopus search excluded newspaper articles due to the otherwise high number of non-peer reviewed articles. Detailed search strategies are provided in Supplementary Materials – 5.1 Database Search Terms.

### 2.3.3 Citation Management

Search results were exported to Microsoft Excel and imported to Microsoft Access (Microsoft Corporation, Redmond, WA) for title and abstract relevance screening. Citation fields included: author, reference, journal, title, and abstract. Each database was independently de-duplicated and then combined. Duplicates were identified and eliminated independently by two researchers (Lorelei Ford and Lianne McLeod) and agreement confirmed.

### 2.3.4 Eligibility Criteria

Study selection was a two-step screening process involving a title and abstract screen. In addition to title and abstract screening methods identified by Arksey and O’Malley (2005) and Pham et al. (2014), abstracts were categorized, according to the inclusion criteria in Table 2.1, by two researchers (Lorelei Ford and Lianne McLeod) during screening to enable reliable sorting for full-text review.

Scoping review inclusion criteria to identify human health risk assessments applied to unregulated or unspecified drinking water.

Table 2.1 Scoping review inclusion criteria to identify human health risk assessments applied to unregulated or unspecified drinking water.

<b>Inclusion Criteria</b>
English language
Published between January 1 <sup>st</sup> , 2000 to May 8 <sup>th</sup> , 2014
Peer-reviewed
Identified applied HHRA
Identified water use for human consumption
Identified the water source as unregulated or unspecified <sup>a</sup>

<sup>a</sup> Professional judgement and consensus was used to categorize studies that did not identify the water source as unregulated but provided evidence that the source water was not regulated.

Titles were included if it was clear they *were* or *could* be about risk assessment and drinking water, to minimize the potential for exclusion of relevant articles. For this scoping review, regulated water sources (e.g. municipal treatment, community treatment, or centralized water sources for cities and towns) were excluded to focus the review on unregulated water sources (e.g. private drinking water wells, raw water sources, etc.). Unspecified water sources represent a category of studies that failed to confirm the water source as unregulated and did not describe the

site, hazards tested, or circumstances to suggest water was regulated. Unspecified water sources, likely unregulated, were included in analysis to identify shortfalls in reporting but excluded from descriptive statistics when specifically addressing unregulated water sources.

### 2.3.5 Title and Abstract Relevance Screening

Titles and abstracts were scanned independently by two researchers (Lorelei Ford and Lianne McLeod) to prevent exclusion of valid citations. Disagreements between reviewers during this scan resulted in the article's inclusion for full-text review. A form was created in Microsoft Access to categorize the abstracts to reach consensus on meeting inclusion criteria. The title and abstract scans were completed November 6<sup>th</sup> and November 20<sup>th</sup>, 2014, respectively.

### 2.3.6 Data Characterization

Articles meeting inclusion criteria were eligible for full-text review. Themes and categories were developed and defined based on specific references and terms to ensure characterization of data was consistent. Three broad themes were developed to include HHRA characteristics, literature characteristics; and literature gaps. Categories within the human health risk assessment characteristics theme included the exposure population, exposure pathway, hazard identification, applied methods, framework used, HHRA terminology, factors and uncertainty, and outcomes specific to the application of risk assessment. Literature characteristics related to the world region in which the studies took place, publication dates, and publication sector (or field). Literature gaps, defined as any gap identified in the study by authors, general fit into three categories including gaps in HHRA research, risk management, and community exposure. Except for a few cases, in which researchers contacted authors by email via ResearchGate (ResearchGate GmbH, Berlin, Germany), full-text articles were accessed through the University of Saskatchewan online library. Non-peer reviewed literature was eliminated from the review. If studies did not provide sufficient evidence for exclusion they were retained for analysis and identified as 'unspecified'. Prior to full-text review, all researchers independently reviewed one randomly selected article (i.e. Kavcar et al. 2009) and discussed themes, categories, and definitions as suggested by Levac et al. (2010). Full-text review was conducted by three researchers (Lorelei Ford, Lalita Bharadwaj and Cheryl Waldner) and studies which failed to meet requirements of inclusion criteria for abstract scan and full-text review were removed from further analysis. Individual reviews were summarized and discrepancies or questionable categorizations were re-examined prior to combining results. Final categorization was completed on June 30<sup>th</sup>, 2015. A detailed list of the themes and categories, including examples, and references for the full-text review, are summarized in Supplementary Materials – 5.2 Full-Text Review Categorizations.

### 2.3.7 Data Summary and Synthesis

Screening and full-text review were compiled using Microsoft Excel 2010 (Microsoft Corporation, Redmond, WA). All data entries were reviewed and scanned for manual errors or



incomplete entries prior to analysis. Calculation of descriptive statistics, frequencies, and percentages on nominal data was performed using Microsoft Excel 2010. Charts were designed using Tableau 9.1 (Tableau Software Inc., Seattle, WA).

## 2.4 Results

### 2.4.1 Search and Selection

One-hundred papers met the inclusion criteria for data extraction and scoping review. A total of 7,838 unique articles were found after database results were de-duplicated (Figure 2.1). Further title and abstract screening resulted in the selection of 158 studies for full text review; however, three articles could not be located (i.e. Maqsood 2011; Titilayo et al. 2012; Zhao et al. 2012) and the remaining 55 did not meet inclusion criteria.

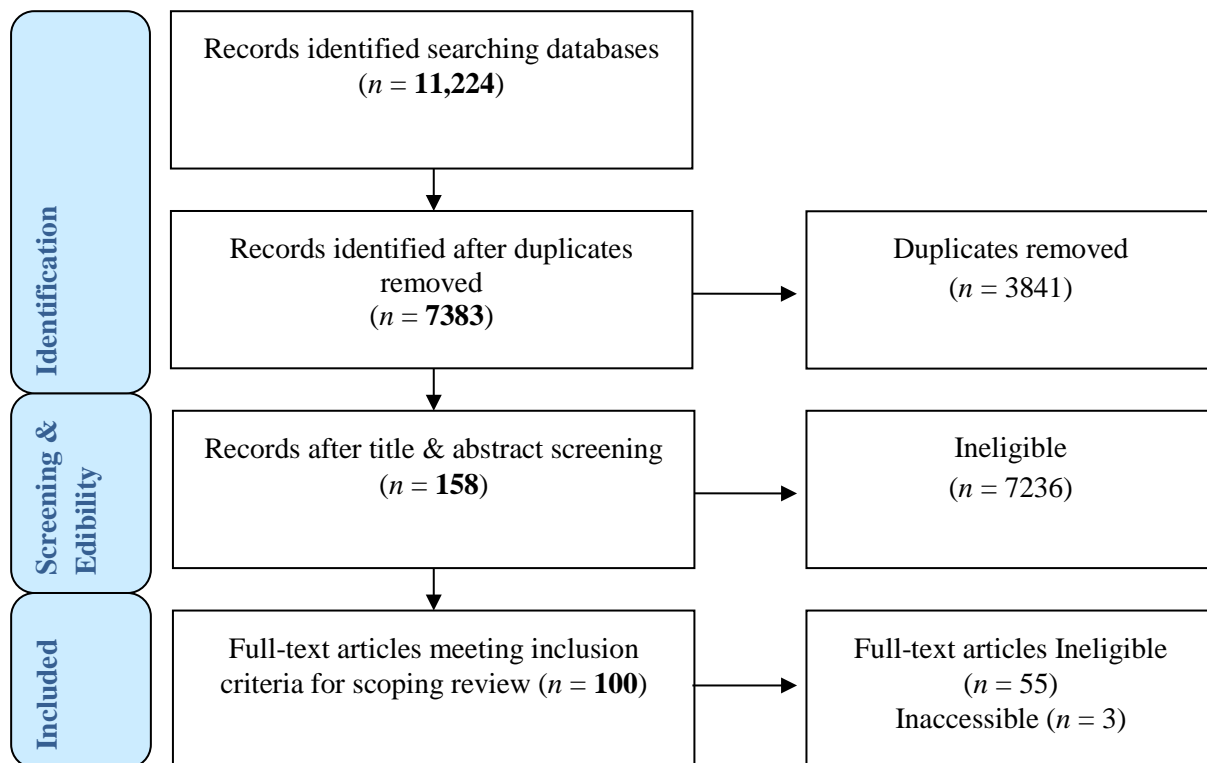


Figure 2.1 PRISMA flowchart of scoping review process. PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analysis).

### 2.4.2 Human Health Risk Assessment Characteristics

Table 2.2 provides a summary of the characteristics of applied HHRAs and categorized into exposure population, exposure pathway, hazard identification (including status of drinking

water), applied method, scope, framework used, HHRA terminology, factors and uncertainty, and outcomes.

Table 2.2 Human health risk assessment characteristics from scoping review literature ( $n = 100$ ).

Characteristic	Number ( $n = 100$ )	Percentage (%)
<b>Exposure Population</b>		
<i>Geographic Area of Population</i>		
Rural (rural and unregulated)	28 (7)	28 (7)
Urban (urban and rural)	16 (4)	16 (4)
Remote (remote and rural)	2 (0)	2 (0)
Unspecified	54	54
<i>Community<sup>a</sup></i>		
Geography	86	86
Topography	27	27
Cultural/Spiritual	2	2
Unspecified	20	20
<i>Receptors<sup>a</sup></i>		
Adults	66	66
Local Residents	41	41
Child	31	31
Toddler	15	15
Teen	15	15
Responsible for source water	13	13
Seniors	11	11
General Public	10	10
Infants	10	10
Local Farmers and Families	5	5
Employees	2	2
First Nation/Indigenous	0	0
Age categories not defined	39	39
Other (e.g., gender, visitors, etc.)	6	6
Unspecified	8	8
<i>Exposure Pathway<sup>a</sup></i>		
Oral	100	100
Dermal	23	23
Inhalation	4	4
<b>Hazard Identification</b>		
<i>Status of drinking water</i>		
Unregulated (unregulated and untreated)	21 (14)	21 (14)
Unspecified (unspecified and untreated)	79 (51)	79 (51)
<i>Source of drinking water<sup>a</sup></i>		
Groundwater (unregulated groundwater)	67 (14)	67 (14)
Surface water (unregulated surface water)	39 (7)	39 (7)
Other (e.g., bottled, rain, cistern, etc.)	21	21
Unspecified	5	5

<i>Type of drinking water</i>		
Untreated	56	56
Untreated and Treated	9	9
Unspecified	35	35
<i>Hazard in drinking water</i>		
Anthropogenic chemical	35	35
Natural chemical	22	22
Anthropogenic and natural chemical	25	25
Microbiological/Pathogen (microbiological/pathogen and chemical)	10 (2)	10 (2)
Radiological (radiological and chemical)	1 (3)	1 (3)
Unspecified	7	7
At least two hazards identified	5	5
<i>Data source <sup>a</sup></i>		
Source water sampled	96	96
Historical data	13	13
Predicted/Extrapolated	11	11
Biomarkers (i.e., hair samples)	3	3
Unspecified	2	2
<i>Applied Method</i>		
Deterministic	86	86
Probabilistic/Stochastic	9	9
Deterministic and Probabilistic/Stochastic	5	5
<i>Scope <sup>a</sup></i>		
Human Health Risk Assessment	100	100
Integrated (human and environmental)	4	4
Holistic (integration of non-traditional data)	0	0
<i>Framework Used <sup>a</sup></i>		
US EPA	75	75
World Health Organization	6	6
Other (i.e., studies, government)	15	15
Unspecified	12	12
<i>HHRA Terminology</i>		
Health (risk) Assessment	47	47
Human Health Risk Assessment	25	25
Risk Assessment	24	24
Other (e.g., cancer risk, risk estimate, etc.)	14	14
<i>Factors and Uncertainty</i>		
<i>Non-Traditional Factors acknowledged <sup>a</sup></i>		
At least one non-traditional factor	90	90
Geography	76	76
Social	23	23
Economic	13	13
Risk Perception	3	3
Cultural/Spiritual	2	2
Other (e.g., behaviours, additional risks, temporal	22	22

effects, etc.)		
<i>Non-Traditional Factors applied<sup>a</sup></i>		
At least one non-traditional factor	69	69
Geography	56	56
Social	4	4
Economic	2	2
Risk Perception	1	1
Cultural/Spiritual	1	1
Other (e.g., behaviours, additional risks, temporal effects, etc.)	16	16
<i>Uncertainty acknowledged<sup>a</sup></i>		
At least one uncertainty acknowledged	83	83
Dedicated section to uncertainty	20	20
Quality Assurance/Quality Control	47	47
Analytical detection limits	38	38
Seasonal/Environment	38	38
Data gaps	30	30
Sufficiency of sampling	28	28
Quality of historical data	10	10
Other (e.g., exposures, toxicological factors, effects of unknown variables, etc.)	18	18
<i>Outcomes</i>		
<i>Result<sup>a</sup></i>		
Exposure Assessment	96	96
Hazard Assessment	95	95
Hazard Quotient/Index	81	81
Epidemiological Assessment	4	4
Other (i.e., quantitative microbial risk assessment and cancer risk)	27	27
<i>Conclusion by Authors</i>		
Quantitative	94	94
Quantitative and Qualitative	4	4
Qualitative	2	2

<sup>a</sup> not mutually exclusive.

Human health risk assessments were applied to rural populations dependent on unregulated source water was found in only 7% (7/100) of the scoped studies (Table 2.2). Overall unregulated water sources were identified in only 21% (21/100) of the studies, while the remaining (79%, 79/100) failed to specify the regulatory status but did not provide enough information to be excluded as regulated. Over half (54%, 54/100) of the geographic areas for the population were insufficiently described and could not be categorized as rural, urban, or remote.

Source water categories including ground and surface water were, not exclusively, identified in 67% (67/100) and 39% (39/100) of the reviewed studies, respectively (Table 2.2). Groundwater was categorized as unregulated in 14% (14/100) of the studies, which doubled the percentage of

surface water sources found to be unregulated (7%, 7/100). Regardless of the source water's regulatory status, groundwater was identified as untreated in 64% (43/67) of the articles (e.g. Çelebi et al. 2014; Phan et al. 2013; Su et al. 2013; Sultana et al. 2014; Wu et al. 2014) versus only 10% (4/39) surface water. Only three studies identified a rural population dependent on unregulated and untreated groundwater (i.e. Jamaludin 2013; Papić et al. 2012; Peplow and Edmonds 2004).

Drinking water hazards were identified as natural or anthropogenic chemicals in 82% (82/100) of articles reviewed (Table 2.2). Risks associated with bacteria, viruses, parasites, and radiological parameters were studied in 11% (11/100) of the HHRAs, exclusive of chemicals; with a small proportion (5%, 5/100) including a chemical hazard in addition to microbes, pathogens, or radiological parameters.

Receptors, defined as the specific group of people exposed to potential risk, were inconsistently described throughout the reviewed literature. Not mutually exclusive, the literature identified adult or local residents as receptors in 66% (66/100) and 41% (41/100) of the studies, respectively (Table 2.2). A specific age category for receptor descriptions was not defined in 39% (39/100) of the studies. Other receptor categories identified (Table 2.2) included: children, toddlers, teens, '(people) responsible for source water', the 'general public', infants, 'local farmers and families', or 'employees'. No studies identified receptors as First Nations, or indigenous communities. When the exposure population was described as a community, the population was delineated by a geographic area (86%, 86/100), topography (27%, 27/100), cultural or spiritual characteristics (2%, 2/100), or were unspecified (20%, 20/100) due to their vague descriptions they were in proximity to sources of pollution, source water, or hydro-geological influences.

Table 2.2 shows that 86% (86/100) of HHRAs applied to unregulated or unspecified drinking water were deterministic with 14% (14/100) utilizing probabilistic and/or stochastic methods in their analysis (i.e. Busset et al. 2010; Deng et al. 2012; Donovan et al. 2008; Hunter et al. 2011; Kavcar et al. 2009; Kumar et al. 2010; Li et al. 2007; Marara et al. 2013a; Mondal et al. 2010; Nzihou et al. 2013; Razzolini et al. 2011; Wang et al. 2009; Williams et al. 2000). Only four studies had an integrated environmental risk in addition to human health (i.e. Buczyńska and Szadkowska-Stańczyk 2005; Genthe et al. 2013; Liu et al. 2011; Yu et al. 2010). The USEPA risk assessment framework was applied in 75% (75/100) of the studies, while 6% (6/100) of the studies utilized the standardized international methods of the World Health Organization. Peer-reviewed, other government or non-government methods of HHRA were applied in 15% (15/100) of the studies while 12% (12/100) had no clear methodological framework.

Use of terminology describing HHRAs was inconsistent within and between studies. The term 'health risk' or 'health risk assessment' was used in 47% (47/100) of the scoped articles. Less frequently the terms 'human risk assessment' or 'human health risk assessment', and 'risk assessment' described the assessment in 25% (25/100) and 24% (24/100), respectively. Other

articles (14%, 14/100) specifically described the assessments as quantitative microbial (or health) risk assessment, cancer risk, risk estimates, and hazard evaluations.

Non-traditional factors were acknowledged or applied qualitatively, by lending to the interpretation of risk, but were not quantified variables within the risk assessment. Non-traditional factors were acknowledged in 90% (90/100) of the studies, however, their qualitative application to the interpretation of risk was only 69% (69/100; Table 2.2). Geographical (76%, 76/100), social (23%, 23/100) and economic (13%, 13/100) factors were acknowledged most frequently. Only 5% (5/100) of studies recognized risk perception, or cultural/spiritual non-traditional factors. The 'other' categories included: health variables (e.g. Giri and Singh 2014; Singh and Ghosh 2012), temporal influences (e.g. Giri and Singh 2014; Jamaludin 2013; Sultana et al. 2014; Yacoub et al. 2013), differences in water sources (e.g. Çelebi et al. 2014; de Jongh et al. 2012; Hynds et al. 2014), effectiveness of risk management (i.e. Machdar et al. 2013), and human behaviors or proximity to human activities (i.e. Ahmed et al. 2010; Armah et al. 2012; Buczyńska and Szadkowska-Stańczyk 2005; Lee et al. 2005, 2006; Ramirez-Andreotta et al. 2013; Santos et al. 2013; Zheng et al. 2013).

Uncertainty was acknowledged at least once in 83% (83/100) of the articles, but only 20% (20/100) provided a section specifically dedicated to the discussion of uncertainty (Table 2.2). Quality assurance and quality control, and analytical detection limits were mentioned in 47% (47/100) and 38% (38/100) of the articles, respectively. Seasonal or environmental influences, such as changes in hazard concentrations over time, were identified in 38% (38/100) of studies. Data gaps (30%, 30/100) and sufficiency of sampling (28%, 28/100) were more frequently mentioned than the quality of historical data for use in the calculation of risk (10%, 10/100). Other sources of uncertainty were disclosed in 18% (18/100) of the articles and included: uncertainty associated with reference to supplementary material or methods (i.e. Kim et al. 2004; Törnqvist et al. 2011); variation in exposure (i.e. Ahmed et al. 2010; Kavcar et al. 2009; Kelepertzis 2014; Lee et al. 2007; Ni et al. 2009; Steyn et al. 2004; Zheng et al. 2013); insufficient toxicological data or guidelines (i.e. Lee et al. 2007; Li et al. 2007; Ramirez-Andreotta et al. 2013; Rapant and Krcmová 2007); error in methods or their application (i.e. Ahmed et al. 2010; Hynds et al. 2014; Kavcar et al. 2009; Kazama et al. 2012; Lee et al. 2010; Li et al. 2007; Steyn et al. 2004); unknown immunity, virulence, reporting and diagnosis (i.e. Hunter et al. 2011); and, failure to consider secondary effects or multiple sources of risk (i.e. Addo et al. 2013; Lee et al. 2010; Papić et al. 2012; Santos et al. 2013).

#### 2.4.3 Literature Characteristics

Table 2.3 provides a summary of the literature characteristics including: the region(s) in which the research was conducted; the number of studies published; and, the sector or discipline the studies were published in.

Most (57.4%, 58/100) of the studies were conducted in Asia and included the countries of China, Pakistan, India, and Bangladesh. All studies reported one region in which the research took place with exception of Hunter et al. (2011) research conducted in France and the United Kingdom; therefore, there were 101 study regions identified in the scoped literature. Figure 2.2 provides a visual summary of the number of studies by world region.

Table 2.3 Literature characteristics from scoping review ( $n = 100$ ).

<b>Characteristic</b>		
<b>World Region</b>	<b>Number (<math>n = 101</math> <sup>a</sup>)</b>	<b>Percentage (%)</b>
Asia	58	57.4
West Africa	9	8.9
Europe	7	6.9
European Union	8	7.9
North America	7	6.9
South America	4	4.0
South Africa	3	3.0
Middle East	2	2.0
Caribbean	1	1.0
East Africa	1	1.0
Oceania	1	1.0
<b>Publication Year</b>	<b>Number (<math>n = 100</math>)</b>	<b>Percentage (%)</b>
January 2010–May 2014	75	75
January 2005–December 2009	20	20
January 2000–December 2004	5	5

<sup>a</sup> not mutually exclusive, one study took place in two regions.

The number of articles published annually increased during the review period from January 2000 to May 2014. Twenty-five percent (25/100) of the articles were published from 2000 to 2009, while the remaining 75% (75/100) were published in less than half that period from January 2010 to May 2014. The highest number of publications per year (19%, 19/100) occurred in 2013. From January 2000 to 2013 is the average publishing rate is 6.6% per year excluding studies from January to May 2014. Figure 2.3 provides the number of publications by sector and year where sectors are not mutually exclusive. Articles were predominately published in journals indicating a focus on human health (94%, 94/100), toxicology (81%, 81/100), and environment/resource management (79%, 79/100).

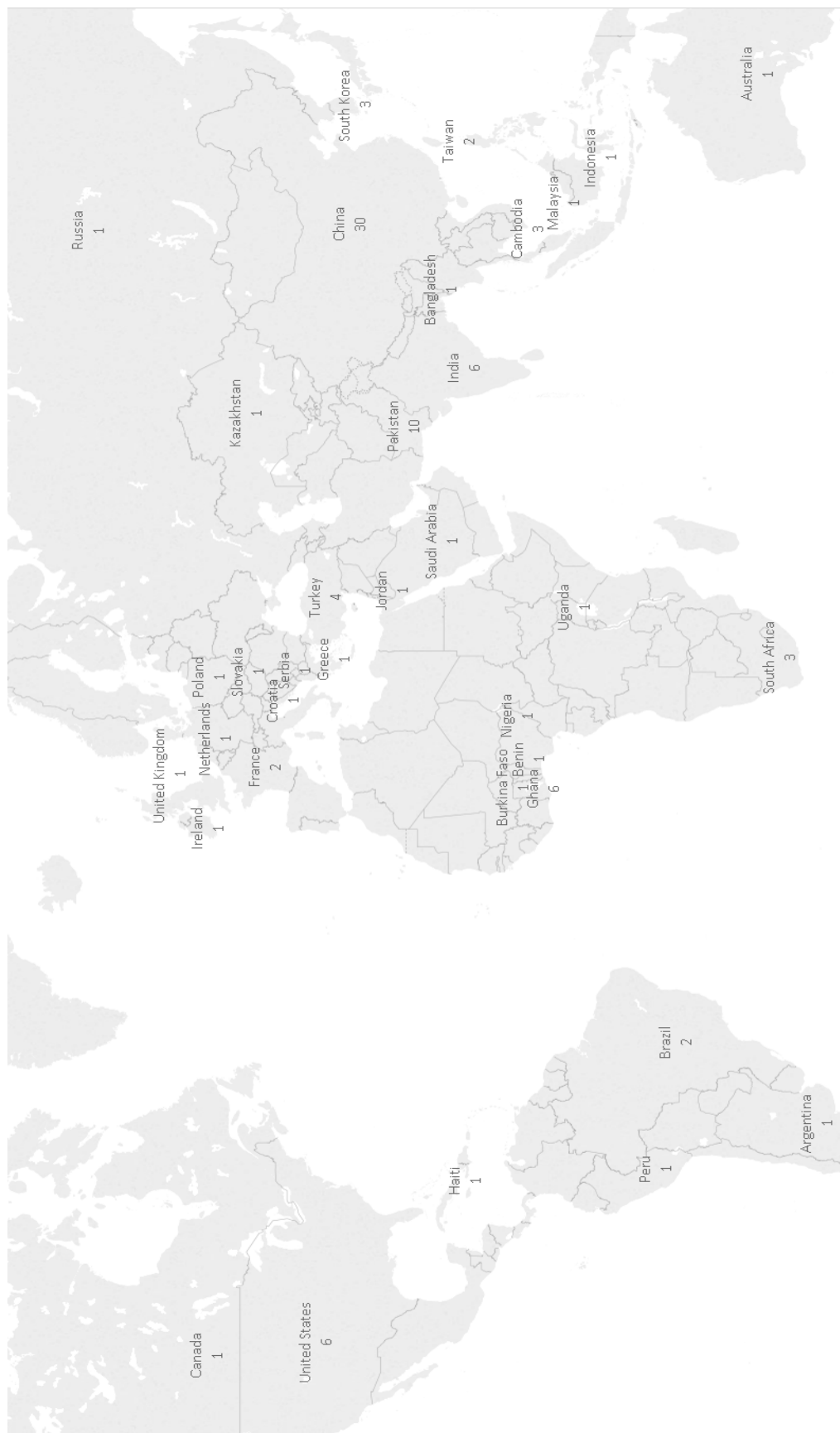


Figure 2.2 Number of scoping review studies by world region.



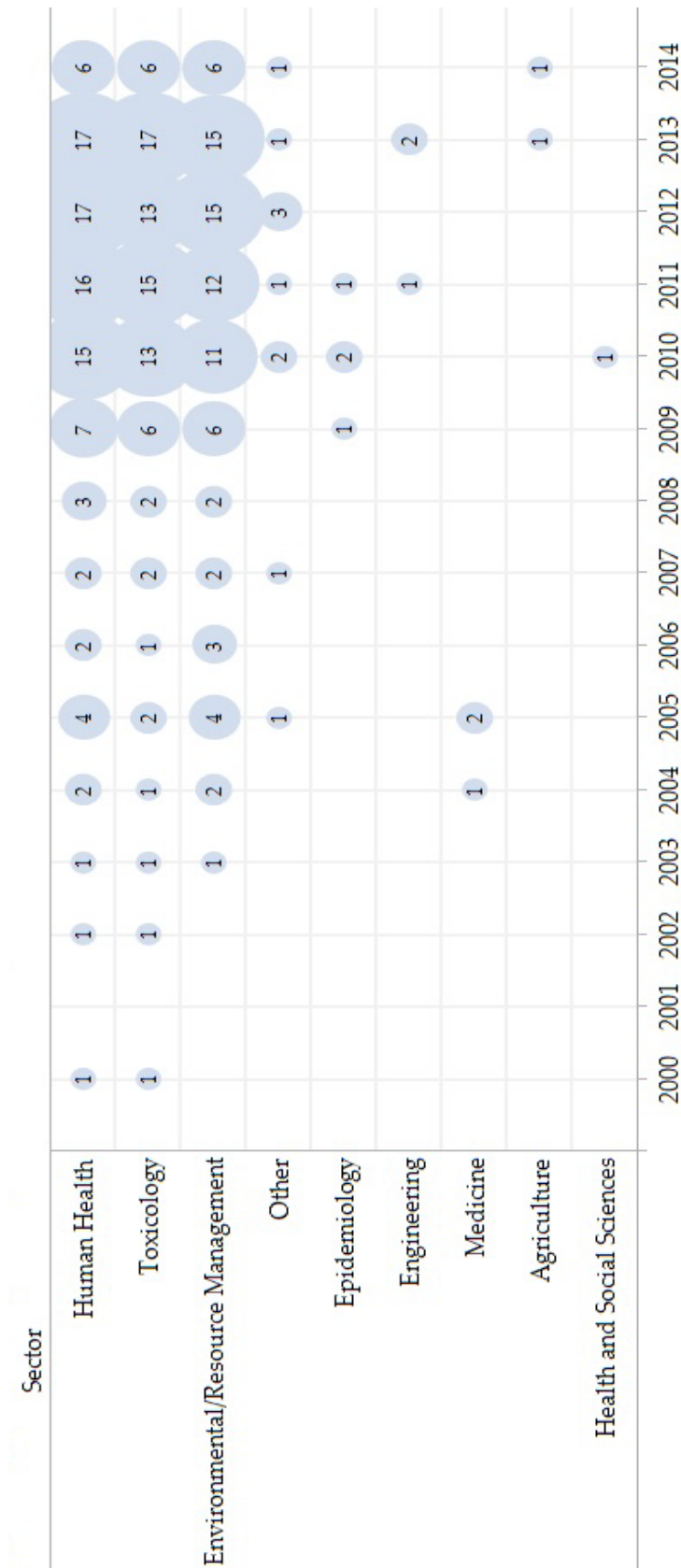


Figure 2.3 Scoping review studies by sector and year. Sectors are not mutually exclusive.

#### 2.4.4 Literature Gaps

At least one gap in the literature was identified in 47% (47/100) of the studies. Literature gaps were not mutually exclusive and were summarized into three main categories including: the research field of HHRA research (35%, 35/100), risk management gaps associated with mitigations to reduce risk (22%, 22/100), and community exposure (10%, 10/100). Table 2.4 provides detailed descriptions of the gaps identified in the literature and relevant studies.

Table 2.4 Description and references for research, management, and community gaps identified in the scoping review literature ( $n = 67$ ).

Gap Description	References
<i>Research in HHRA</i>	
Use of biomonitoring	(Obiri et al. 2010)
Improved methods or application	(B. Wu et al. 2010; Hunter et al. 2011; Ahmed et al. 2010; Steyn, Jagals, and Genthe 2004; Howard, Pedley, and Tibatemwa 2006)
Sources of uncertainty	(Peplow and Edmonds 2004; B. Wang et al. 2009)
Determining temporal exposures	(Momot and Synzynys 2005)
Determining future exposures	(Rapant and Krčmová 2007; P. R. D. Williams et al. 2000)
Considering all pathways of exposure	(Busset et al. 2010; Buchhamer et al. 2012; Qiao et al. 2010; Chai et al. 2010; Ujević Bošnjak et al. 2012; Mondal et al. 2010)
Exposure to additional hazard sources	(Busset et al. 2010; P. Williams et al. 2002; Kelepertzis 2014; Zheng et al. 2013)
Exposure to mixtures	(Phan et al. 2013; Qiao et al. 2010; de Jongh et al. 2012; B. Wang et al. 2009; Ma et al. 2014; Genthe et al. 2013)
Guides to direct researchers	(Caylak 2012a)
Gather more epidemiological evidence and toxicological data	(Marara, Palamuleni, and Ebenso 2013; Emmanuel, Pierre, and Perrodin 2009; Wu et al. 2010; Peplow and Edmonds 2004; Kelepertzis 2014; Ramirez-Andreotta et al. 2013; Lee et al. 2006; Razzolini et al. 2011)
<i>Risk Management</i>	
Collect data to inform management	(Ni et al. 2009; Machdar et al. 2013; Razzolini et al. 2011; Ramirez-Andreotta et al. 2013)
Knowledge of geochemistry and aquifers	(Singh et al. 2014; Emmanuel, Pierre, and Perrodin 2009)
Monitoring	(Wu et al. 2010; de Jongh et al. 2012)
Evaluation of exposures	(Williams et al. 2002; Hynds, Gill, and Misstear 2014; Buczyńska and Szadkowska-Stańczyk 2005)
Establish national/regional HHRAs	(Kumar et al. 2010; Addo et al. 2013; Etchie, Etchie, and Adewuyi 2012; Ahmed et al. 2010)
Standardize methods for mixtures	(Wang et al. 2009)

Standardize regulations	(Yacoub et al. 2013; Leung et al. 2013)
Improved communication, response and determination of risk	(Santos et al. 2013; Genthe et al. 2013; Steyn, Jagals, and Genthe 2004; Ramirez-Andreotta et al. 2013)
<i>Community Exposure</i>	
Inclusion of specific community (i.e., sensitive community members)	(Williams et al. 2000)
Isolate risks specific to communities	(Caylak 2012b; Hunter et al. 2011; Giri and Singh 2014)
Consider quality of life, socioeconomic, and political factors	(Lee et al. 2010; Singh et al. 2014; Z. Wang et al. 2011)
Improve community involvement, engagement, education, and risk management	(Karim 2010; Genthe et al. 2013; Razzolini et al. 2011)

---

## 2.5 Discussion

This paper provides an overview of HHRAAs applied to unregulated drinking water in peer-reviewed literature and describes the frequency of their application to rural communities, the characteristics of their methodology, and gaps identified in the literature. Most of the scoped publications (79%) failed to specify the regulatory status of source water. The inclusion of literature with water sources of unknown regulatory status reveals the need to improve characterization of source water hazards in HHRA. Although 28% of applied HHRAAs were identified as taking place rural communities, only 7% clearly identified both a rural population and unregulated water source. Similarly, in a third of the articles the source water was not specifically described as raw or treated. This lack of transparency in identifying the population of concern has been previously described in a review by Pons et al. (2015) of waterborne disease outbreaks in Canada and the United States, and appears to be an ongoing oversight by authors reporting on risk associated with drinking water. It is essential to describe the population of concern and the regulatory status of source water utilized for drinking purposes to effectively assess the potential drinking water risks to global rural communities; to support development of appropriate risk management options; and, to further research in the discipline of human health risk assessment.

The water source (i.e. ground, surface, and other) was highly reported in the studies which suggest that groundwater (67%) was the most frequent source of drinking water; however, only 14 of the studies identified the groundwater source as unregulated. Although only 21 studies identified an unregulated drinking water status, it is possible that 51 of the studies that did not specify the regulatory status but identified untreated water could be identified as unregulated. A high proportion of unregulated groundwater use would be expected given the global effort to meet the needs of increasing populations and improve accessibility of drinking water in rural and

remote locations (Kundzewicz and Döll 2009; WHO/UNICEF 2015; Pons et al. 2015; Famiglietti 2014). Information on source, treatment and regulatory status of drinking water is essential for effective use of reported data. The potential for risk is very different between treated and untreated sources. For example, treated water may pose risks associated with disinfection by-products while raw groundwater sources may focus on naturally occurring heavy metals. The very nature of unregulated source water implies a lack of management options such as regular maintenance and monitoring. Without clear identification of drinking water supplies, and reliable information, data, and reporting, it is difficult to gauge risk and provide risk management options to rural communities.

The application of HHRA methods was largely deterministic with approximately 1 in 7 reporting the use of probabilistic or stochastic methods. Though these methods are not being utilized to *integrate* non-traditional factors into a *holistic* HHRA, more than half of the papers mentioned or qualitatively applied non-traditional factors to the interpretation of risk. For example, the most frequently acknowledged non-traditional factor was geography which was often used to define the area associated with the hazard or to compare risk between specific areas. A shift from deterministic to probabilistic methods (which can utilize stochastic distributions) has benefits including: the quantification of uncertainty (Bridges 2003; Burns et al. 2014); less dependence on animal based studies (Bridges 2003); increased transparency in the process of risk assessment (Burns et al. 2014); the potential inclusion of qualitative information (Serre et al. 2003); and, the use of vast and multiple data types (Zargar et al. 2014). In the context of this review only 4 studies carried out what the WHO/IPCS (2001) defined as an integrated risk assessment; however, only (Genthe et al. 2013) further met the scoping review criteria addressing rural population consuming unregulated source water.

Holistic approaches using probabilistic risk assessment methods and decision-type networks (e.g. Bayesian Risk Assessment) that can utilize qualitative and quantitative data were not applied in the literature despite frequent acknowledgement and use of non-traditional factors to interpret risk (e.g. comparison of risk between geographical areas). The integration of qualitative data, such as behavior, can improve risk management due to its influence on water use and exposure for rural communities (Chowdhury, Champagne, and McLellan 2009; Hertwich, McKone, and Pease 1999). Researchers could explore the benefits of probabilistic and stochastic methods in holistic HHRA to integrate non-traditional factors potentially influencing risk and to better characterize uncertainty (Serre et al. 2003). For example, effective education or government programming to alter human behaviour can be used decrease exposure to hazards, rather than treating illness outcomes. Therefore, by determining how the behaviour changes the overall risk, the strategy for risk communication and management can be tailored to the receptors.

Researchers continue to rely on traditional methods of HHRA despite the advances in software and data processing capability; the ongoing need to improve the use of data and accuracy of risk assessment; and, encouragement to use probabilistic methods by governments (i.e. US EPA 2015). Probabilistic methods in HHRA can enable more holistic risk assessments (e.g. Zargar et

al. 2014), similar to the environmental field (e.g. Hooten and Hobbs 2015), to assess not only multiple hazards but to include non-traditional factors that may influence risk.

The potential influence of non-traditional factors is related to uncertainty if they have an influence on the overall measure of risk (Slovic 1999; Boholm 2010; Renn 1998). Uncertainty is an important part of any risk assessment because it provides the caveats that may affect the interpretation of the risk measure. Fewer than half of the papers reported quality assurance and control within their studies. Declaration of uncertainty is fundamental to risk assessment (Burns et al. 2014) and well-established frameworks provide checklists to ensure users disclose uncertainty (IPCS 2014; US EPA 1989). Twenty percent (20%) of the reviewed research papers addressed uncertainty and limitations under a specific sub-heading in the article. Without full disclosure of uncertainty, it is difficult compare or assesses risk evaluations.

A significant short-coming identified in the literature was a lack of defined at risk exposure populations. This can be improved when thorough descriptions of receptors are provided (Kavlock et al. 1996). For example, age groupings for receptors and terms such as ‘rural’ and ‘urban’ should be defined with geographic area for better characterization of risk. Adult receptors were frequently chosen to represent communities while sub-groups or sensitive populations were less frequently identified. The scoped studies had limited demographic representation of the receptors and considered only a single route of exposure. Despite the perceived need for inclusion of all exposure pathways as identified in the literature (Buchhamer et al. 2012; Chai et al. 2010; Qiao et al. 2010), the oral pathway of exposure was most frequently assessed.

Communities were often defined by a geographic or topographic area, implying a natural link between groundwater hazards and the physical environment, notably the geology or land-use. However, a geological approach including the interpretation of hydro-geology could be more relevant when associations between geology and hazards are required (e.g. Rajagopal and Talcott 1983). Typically, receptors in studies were vaguely identified as ‘local residents’, ‘general public’, ‘local farmers and their families’, and, ‘individuals responsible for their source water’. Related to the need to better describe the receptors, researchers identified gaps in community exposure including the necessity to address additional receptor groups or communities to improve aspects of risk assessment or management options. Clearly defining the receptors and communities in the human health risk assessment further improves the research, allowing future research to build on the knowledge associated with the characteristics of similar receptors.

Studies most frequently identified natural and anthropogenic chemicals as potential hazards. The focus of the studies on chemicals, versus bacteriological water quality parameters, suggests that the unspecified and untreated (51%) water sources in the studies may largely be unregulated groundwater; however, we are unable to confirm this. Interestingly, bacteria, pathogens, and radiological parameters were infrequently included in studies despite their presence in surface and groundwater (Ritter et al. 2002 and Villanueva et al. 2014). Thus, future research considering

risk associated with chemical, radiological, and microbiological parameters may provide a more comprehensive measure of risk for communities dependent on unregulated source water. Source water and specific hazards were generally well defined in the scoped studies; however, risk management or mitigation would benefit from comprehensive characterization of hazards and receptors including: mixed chemical or hazard exposures; geographical/geological influences; social/societal factors; and limitations and uncertainties associated with all aspects of HHRA.

The relative frequency of HHRA research on unregulated or unspecified drinking water is variable globally and is primarily focused in Asia. Since 2000, the majority of HHRA studies have taken place in countries with large numbers of rural residents without improved drinking water sources and high exposures to natural and anthropogenic water quality hazards (e.g. China and India; UNICEF/WHO 2012; Zhang et al. 2010). Conversely, there is an absence of HHRA studies conducted in more developed regions with known drinking water hazards (e.g. North America). Villanueva et al. (2014) suggest that assessing drinking water exposure is a challenge due to insufficient information on hazards and exposure. Therefore, global rural populations reliant on unregulated drinking water, regardless of regional socio-economic status, may be at increased health risk due to a mistaken perception that hazards are low. Alternatively, underutilization of safe unregulated drinking water is a missed opportunity to provide sustainable water to rural populations. Considering the development status of countries, a developed region (e.g. North America) would have the resources required to drastically reduce the risks to their population reliant on unregulated drinking water, and the research and risk management strategies carried out may provide insight into the larger global challenge of improving access to safe drinking water.

Research publications focusing on unregulated or unspecified drinking water increased from 2000 to 2014. For the scope of this review, publications prior to 2000 were not included because the literature is dated and data analysis methods have since advanced with mainstream use of computers for analysis (USEPA 2001; Hooten & Hobbs 2015). Regardless, this review determined an approximate 7% annual publication rate which is similar to the global exponentially increasing annual publication rate of approximately 8% from 1980 to 2012 (Bornmann and Mutz 2015). In addition to increased publishing, the Millennium Declaration was established in 2000 by the Member States of the United Nations leading to the Millennium Development Goals and United Nations initiatives which have focused on improving access to safe drinking water and sanitation internationally (WHO/UNICEF 2015). These programs ‘gained momentum in the 2000s’, (Bartram et al. 2014) which may have created increased funding opportunities for drinking water research in countries with large rural populations lacking access to safe and sustainable drinking water. If these global initiatives are influencing publications, drinking water research, particularly in undeveloped countries and vulnerable communities, should continue to increase in the wake of international initiatives such as the World Health Organization’s *Water Quality and Health Strategy 2013-2020* (WHO 2013).

Publications on drinking water quality and human health best fit into journals addressing the interrelationship between disciplines focused on human health, risk assessment, and the environment. Researchers conducting *risk assessments* on *human health* should use the full description *human health risk assessment* instead of variations that introduce ambiguity. Human health risk assessments are defined by the USEPA (2015a) as ‘...the process to estimate the nature and probability of adverse health effects in humans who may be exposed to chemicals in contaminated environmental media, now or in the future’. Use of standardized terminology in title and abstract would ensure risk assessments with human subjects are easily identified during literature searches. Increased consistency in use of terminology, in addition to improvements already discussed (i.e. need to better characterize the hazards and receptors), would improve the clarity and transparency of applied HHRAs.

Fewer than 50% of studies identified gaps in the literature. Risk assessment short-comings were identified more frequently than gaps in risk management or community exposure. Risk assessment gaps often included the need for increased epidemiological and toxicological data, in an effort to understand the toxicological effects when exposed to chemical mixtures through multiple pathways. Risk management gaps, identified by researchers, expressed a similar need for increased data and monitoring, and improved evaluations of exposure. In addition, risk management gaps highlighted the desire by researchers to have specific national or regional HHRAs. We can summarize the gaps identified by researchers in the field of HHRA to say that overall they require: increased data collection and monitoring as well as strong integration with research fields that support HHRA (e.g. toxicology and epidemiology); the determination of risk by way of standardized methods and guides that improve accuracy and account for uncertainty; community-based research approaches that consider how the data and results can be used to support ongoing drinking water management; and, improved communication and involvement with communities to ensure the outcome of HHRA studies are specific and relevant as it relates to the receptors and their exposure.

In addition to risk assessment and management gaps identified in the scoped studies, the need for risk characterization specific to communities has been recognized by researchers. Consideration of non-traditional factors (e.g. quality of life, socioeconomic, and political) have been suggested and supports the need to determine how these factors may influence risk. The importance of HHRA to protect community health requires transparency and diligent data collection, analysis, and reporting. This could be achieved through equal partnerships with communities and would be beneficial, ethical and in practice with a community based participatory research approach where both the researchers and community would benefit (O’Toole et al. 2003; Slovic 1999). In the context of HHRA it is ideal to meet the goals of research and management for applied research that benefits both academia and communities.

### 2.5.1 Strengths and Limitations

This scoping review was carried out with a systematic approach. Inclusion of five databases, each varied in breadth and depth, ensured the necessary coverage required for this review. The multidisciplinary team and frequent communication provided a balanced process and facilitated consensus through screening and full-text review, thus, eliminating the need for reliability statistics. A professional librarian guided initial database searches decreasing the likelihood of bias or error associated with attaining citations relevant for review. Abstract categorization assisted in development of inclusion or exclusion criteria for full-text review while allowing the team to become familiar with the literature as recommended by Daudt, van Mossel, and Scott (2013). Consistent with Pham et al. (2014), full-text reviews did not include qualitative or quantitative assessment of research quality. Research team meetings at each step through the scoping process were necessary to integrate advice from the team, and maintain effective communication (Daudt, van Mossel, and Scott 2013).

The possibility exists that relevant articles were excluded. Ending the search in May 2014 limited interpretations of publication trends up to publication. Despite mutually established and well-defined definitions for charting, the full-text review between researchers is subject to interpretation error. Exclusion of regulated water sources limited our ability to compare the characteristics of the scoped studies to regulated sources; however, the focus of this scoping review was to determine the characteristics associated with HHRA studies focused on unregulated source water.

## 2.6 Conclusion

A summary of the HHRA literature and methods applied to populations dependent on unregulated or unspecified drinking water sources is provided. This review reveals a lack of HHRA research dedicated to rural populations dependent on unregulated source waters in spite of the global concern regarding access to safe drinking water. The majority of the scoped HHRA were applied in countries of proportionally high rural populations globally, of which a large proportion of water is unregulated and untreated. Insufficiently defined and poorly disclosed risk assessments decrease the usefulness of the research when attempting to gather vital information on exposure populations, water sources, and hazards to further this area of study or manage risk. The field of HHRA may be delayed in the adoption of methods that allow for the inclusion of various data types and the quantification of uncertainty for a holistic approach. It is essential that literature gaps identified by researchers and summarized herein, are used to inform the future direction of research and management on unregulated drinking water for the world's rural populations. Furthermore, the adoption of community-based participatory approaches, where possible, will provide the information necessary to support risk management decision-making and improve the health of communities.



## 2.7 Recommendations

Global rural populations face potential health risks related to water quality hazards associated with unregulated source water. Evolution and improvement in the approach and application of HHRA methods are necessary for a better understanding of the human health risks, and improved risk communication and management in rural populations. Recommendations for researchers, based on a summary of studies in the field of HHRA on unregulated and unspecified source waters, are as follows:

- Components of the HHRA (e.g. exposure population, source water, hazards, etc.) should be adequately described to improve the detection of potential relevant literature upon title and abstract searches, and the quality of research reporting. Consistent use of terminology and reporting associated with standardized HHRA frameworks is essential. Uncertainty and limitations should be clearly presented to allow for appropriate interpretation of the research.
- A holistic approach to HHRA should be considered when non-traditional factors are suspected of influencing the human health risk. This can be accomplished with alternative methods of risk assessment (e.g. Bayesian risk assessment) to characterize non-traditional factors and their influence on the human health risks. Gaps in the literature also identify the need to consider the effects and uncertainty non-traditional factors have with respect to multiple hazards, exposures and pathways.
- Identification of gaps in research, management, community, and risk assessments is a necessary component of HHRA. Recognition of gaps in these areas drives research forward, paving the way for new research to better inform future approaches, frameworks, and decision-making.

## 2.8 References

- Addo, M. A., E. O. Darko, C. Gordon, and B. J. B. Nyarko. 2013. "Water Quality Analysis and Human Health Risk Assessment of Groundwater from Open-Wells in the Vicinity of a Cement Factory at Akporkloe, Southeastern Ghana." *E-Journal of Science and Technology* 4 (8): 16–30. doi:10.18780/e-jst.v8i4.838.
- Ahmed, W., A. Vieritz, A. Goonetilleke, and T. Gardner. 2010. "Health Risk from the Use of Roof-Harvested Rainwater in Southeast Queensland, Australia, as Potable or Nonpotable Water, Determined Using Quantitative Microbial Risk Assessment." *Applied and Environmental Microbiology* 76 (22): 7382–91. doi:10.1128/AEM.00944-10.
- Arksey, Hilary, and Lisa O'Malley. 2005. "Scoping Studies: Towards a Methodological Framework." *International Journal of Social Research Methodology* 8 (1): 19–32. doi:10.1080/1364557032000119616.
- Armah, Frederick A., Markku Kuitunen, Isaac Luginaah, and Paul Mkandawire. 2012. "Non Occupational Health Risk Assessment from Exposure to Chemical Contaminants in the Gold Mining Environment of Tarkwa, Ghana." *Trends in Applied Sciences Research* 7 (3): 181–95. doi:10.3923/tasr.2012.181.195.
- Arquette, Mary, Maxine Cole, Katsi Cook, Brenda LaFrance, Margaret Peters, James Ransom, Elvera Sargent, Vivian Smoke, and Arlene Stairs. 2002. "Holistic Risk-Based Environmental Decision Making: A Native Perspective." *Environmental Health Perspectives* 110 (s2): 259–64. doi:10.1289/ehp.02110s2259.
- Bartram, Jamie, Clarissa Brocklehurst, Michael Fisher, Rolf Luyendijk, Rifat Hossain, Tessa Wardlaw, and Bruce Gordon. 2014. "Global Monitoring of Water Supply and Sanitation: History, Methods and Future Challenges." *International Journal of Environmental Research and Public Health* 11 (8): 8137–65. doi:10.3390/ijerph110808137.
- Boholm, Åsa. 2010. "On the Organizational Practice of Expert-Based Risk Management: A Case of Railway Planning." *Risk Management* 12 (4): 235–55. doi:10.1057/rm.2010.4.
- Bornmann, L., and R. Mutz. 2015. "Growth Rates of Modern Science: A Bibliometric Analysis Based on the Number of Publications and Cited References." *Journal of the Association for Information Science and Technology* 66 (11): 2215–22. doi:10.1002/asi.
- Bridges, Jim. 2003. "Human Health and Environmental Risk Assessment: The Need for a More Harmonised and Integrated Approach." *Chemosphere* 52 (9): 1347–51. doi:10.1016/S0045-6535(03)00469-7.
- Briggs, David J. 2008. "A Framework for Integrated Environmental Health Impact Assessment of Systemic Risks." *Environmental Health* 7: 61–77. doi:10.1186/1476-069X-7-61.
- Buchhamer, Edgar E., Patricia S. Blanes, Rosa M. Osicka, and M. Cecilia Giménez. 2012. "Environmental Risk Assessment of Arsenic and Fluoride in the Chaco Province, Argentina: Research Advances." *Journal of Toxicology and Environmental Health, Part A* 75 (22–23): 1437–50. doi:10.1080/15287394.2012.721178.

- Buczyńska, Alina, and Irena Szadkowska-Stańczyk. 2005. "Identification of Health Hazards to Rural Population Living near Pesticide Dump Sites in Poland." *International Journal of Occupational Medicine and Environmental Health* 18 (4): 331–39. <http://www.ncbi.nlm.nih.gov/pubmed/16617849>.
- Burns, Carol J., J. Michael Wright, Jennifer B. Pierson, Thomas F. Bateson, Igor Burstyn, Daniel A. Goldstein, James E. Klaunig, et al. 2014. "Evaluating Uncertainty to Strengthen Epidemiologic Data for Use in Human Health Risk Assessments." *Environmental Health Perspectives* 122 (11): 1160–65. doi:10.1289/ehp.1308062.
- Busset, G., C. Vialle, C. Sablayrolles, M. C. Huau, S. Jacob, and M. Montrejaud-Vignoles. 2010. "Health Risk Assessment Case Study for Trace Metals in Rainwater for Domestic Uses." *Fresenius Environmental Bulletin* 20 (9): 2277–83.
- Caylak, Emrah. 2012a. "Health Risk Assessment for Trace Metals, Polycyclic Aromatic Hydrocarbons and Trihalomethanes in Drinking Water of Cankiri, Turkey." *E-Journal of Chemistry* 9 (4): 1976–91. doi:10.1155/2012/172135.
- . 2012b. "Health Risk Assessment for Arsenic in Water Sources of Cankiri Province of Turkey." *Clean - Soil, Air, Water - Soil, Air, Water* 40 (7): 728–34. doi:10.1002/clen.201100439.
- Çelebi, Ahmet, Bülent Şengörür, and Bjørn Kløve. 2014. "Human Health Risk Assessment of Dissolved Metals in Groundwater and Surface Waters in the Melen Watershed, Turkey." *Journal of Environmental Science and Health, Part A* 49 (2): 153–61. doi:10.1080/10934529.2013.838842.
- Chai, Liyuan, Zhenxing Wang, Yunyan Wang, Zhihui Yang, Haiying Wang, and Xie Wu. 2010. "Ingestion Risks of Metals in Groundwater Based on TIN Model and Dose-Response Assessment — A Case Study in the Xiangjiang Watershed, Central-South China." *Science of The Total Environment* 408 (16). Elsevier B.V.: 3118–24. doi:10.1016/j.scitotenv.2010.04.030.
- Chowdhury, Shakhawat, Pascale Champagne, and P. James McLellan. 2009. "Uncertainty Characterization Approaches for Risk Assessment of DBPs in Drinking Water: A Review." *Journal of Environmental Management* 90 (5). Elsevier Ltd: 1680–91. doi:10.1016/j.jenvman.2008.12.014.
- Daudt, Helena M. L., Catherine van Mossel, and Samantha J. Scott. 2013. "Enhancing the Scoping Study Methodology: A Large, Inter-Professional Team's Experience with Arksey and O'Malley's Framework." *BMC Medical Research Methodology* 13 (1). BMC Medical Research Methodology: 48. doi:10.1186/1471-2288-13-48.
- de Jongh, Cindy M., Pascal J. F. Kooij, Pim de Voogt, and Thomas L. ter Laak. 2012. "Screening and Human Health Risk Assessment of Pharmaceuticals and Their Transformation Products in Dutch Surface Waters and Drinking Water." *Science of The Total Environment* 427–428 (June). Elsevier B.V.: 70–77. doi:10.1016/j.scitotenv.2012.04.010.
- Deng, Yu, Fuquan Ni, and Zhenquang Yao. 2012. "The Monte Carlo-Based Uncertainty Health

- Risk Assessment Associated with Rural Drinking Water Quality.” *Journal of Water Resource and Protection* 4: 772–78. doi:10.4236/jwarp.2012.49088.
- Donovan, E., K. Unice, J. D. Roberts, M. Harris, and B. Finley. 2008. “Risk of Gastrointestinal Disease Associated with Exposure to Pathogens in the Water of the Lower Passaic River.” *Applied and Environmental Microbiology* 74 (4): 994–1003. doi:10.1128/AEM.00601-07.
- Emmanuel, Evens, Marie Gisèle Pierre, and Yves Perrodin. 2009. “Groundwater Contamination by Microbiological and Chemical Substances Released from Hospital Wastewater: Health Risk Assessment for Drinking Water Consumers.” *Environment International* 35 (4). Elsevier Ltd: 718–26. doi:10.1016/j.envint.2009.01.011.
- Etchie, Ayotunde Titilayo, Tunde Ogbemi Etchie, and Gregory Olufemi Adewuyi. 2012. “Systemic Chronic Health Risk Assessment of Residential Exposure to Cd 2+ and Cr 6+ in Groundwater.” *Toxicological & Environmental Chemistry* 94 (1): 181–94. doi:10.1080/02772248.2011.633913.
- Famiglietti, J. S. 2014. “The Global Groundwater Crisis.” *Nature Climate Change* 4 (11). Nature Publishing Group: 945–48. doi:10.1038/nclimate2425.
- Fawell, John, and Mark J. Nieuwenhuijsen. 2003. “Contaminants in Drinking Water.” *British Medical Bulletin* 68 (1): 199–208. doi:10.1093/bmb/ldg027.
- Genthe, B., W. J. Le Roux, K. Schachtschneider, P. J. Oberholster, N. H. Aneck-Hahn, and J. Chamier. 2013. “Health Risk Implications from Simultaneous Exposure to Multiple Environmental Contaminants.” *Ecotoxicology and Environmental Safety* 93 (July). Elsevier: 171–79. doi:10.1016/j.ecoenv.2013.03.032.
- Giri, Soma, and Abhay Kumar Singh. 2014. “Risk Assessment, Statistical Source Identification and Seasonal Fluctuation of Dissolved Metals in the Subarnarekha River, India.” *Journal of Hazardous Materials* 265. Elsevier B.V.: 305–14. doi:10.1016/j.jhazmat.2013.09.067.
- Hart, Barry T., and Carmel A. Pollino. 2009. “Bayesian Modelling for Risk-Based Environmental Water Allocation.” *Waterlines Report Series No. 14*. Canberra.
- Hertwich, Edgar G., Thomas E. McKone, and William S. Pease. 1999. “Parameter Uncertainty and Variability In Evaluative Fate and Exposure Models.” *Risk Analysis* 19 (6): 1193–1204. doi:10.1111/j.1539-6924.1999.tb01138.x.
- Hooten, M. B., and N. T. Hobbs. 2015. “A Guide to Bayesian Model Selection for Ecologists.” *Ecological Monographs* 85 (1): 3–28. doi:10.1890/14-0661.1.
- Howard, Guy, Steve Pedley, and Sarah Tibatemwa. 2006. “Quantitative Microbial Risk Assessment to Estimate Health Risks Attributable to Water Supply: Can the Technique Be Applied in Developing Countries with Limited Data?” *Journal of Water and Health* 4 (1): 49–65. doi:10.2166/wh.2005.058.
- Hunter, Paul R., Marianna Anderle de Saylor, Helen L. Risebro, Gordon L. Nichols, David Kay, and Philippe Hartemann. 2011. “Quantitative Microbial Risk Assessment of Cryptosporidiosis and Giardiasis from Very Small Private Water Supplies.” *Risk Analysis* 31

(2): 228–36. doi:10.1111/j.1539-6924.2010.01499.x.

- Hynds, Paul D., Laurence W. Gill, and Bruce D. Misstear. 2014. “A Quantitative Risk Assessment of Verotoxigenic *E. Coli* (VTEC) in Private Groundwater Sources in the Republic of Ireland.” *Human and Ecological Risk Assessment: An International Journal* 20 (6): 1446–68. doi:10.1080/10807039.2013.862065.
- International Programme on Chemical Safety (IPCS). 2014. “Guidance Document on Evaluating and Expressing Uncertainty in Hazard Characterization.” Geneva, Switzerland. doi:ISBN 978 92 4 150761 5.
- Jamaludin, Noraziah, Shaharuddin Mohd Sham, and Sharifah Norkhadijah Syed Ismail. 2013. “Health Risk Assessment of Nitrate Exposure in Well Water of Residents in Intensive Agriculture Area” *American Journal of Applied Sciences* 10 (5): 442–48. doi:10.3844/ajassp.2013.442.448.
- Karim, M. R. 2010. “Microbial Contamination and Associated Health Burden of Rainwater Harvesting in Bangladesh.” *Water Science & Technology* 61 (8): 2129. doi:10.2166/wst.2010.031.
- Kavcar, Pınar, Aysun Sofuoglu, and Sait C. Sofuoglu. 2009. “A Health Risk Assessment for Exposure to Trace Metals via Drinking Water Ingestion Pathway.” *International Journal of Hygiene and Environmental Health* 212 (2): 216–27. doi:10.1016/j.ijheh.2008.05.002.
- Kavlock, Robert J., George P. Daston, Chris DeRosa, Penny Fenner-Crisp, L. Earl Gray, Steve Kaattari, George Lucier, et al. 1996. “Research Needs for the Risk Assessment of Health and Environmental Effects of Endocrine Disruptors: A Report of the U.S. EPA-Sponsored Workshop.” *Environmental Health Perspectives* 104 (August): 715. doi:10.2307/3432708.
- Kazama, So, Toshiki Aizawa, Toru Watanabe, Priyantha Ranjan, Luminda Gunawardhana, and Ayako Amano. 2012. “A Quantitative Risk Assessment of Waterborne Infectious Disease in the Inundation Area of a Tropical Monsoon Region.” *Sustainability Science* 7 (1): 45–54. doi:10.1007/s11625-011-0141-5.
- Kelepertzis, Efstratios. 2014. “Investigating the Sources and Potential Health Risks of Environmental Contaminants in the Soils and Drinking Waters from the Rural Clusters in Thiva Area (Greece).” *Ecotoxicology and Environmental Safety* 100. Elsevier: 258–65. doi:10.1016/j.ecoenv.2013.09.030.
- Kim, Ye-shin, Hoa-sung Park, Jin-yong Kim, Sun-ku Park, Byong-wook Cho, Ig-hwan Sung, and Dong-chun Shin. 2004. “Health Risk Assessment for Uranium in Korean Groundwater.” *Journal of Environmental Radioactivity* 77 (1): 77–85. doi:10.1016/j.jenvrad.2004.03.001.
- Kumar, Arun, Probas Adak, Patrick L. Gurian, and John R. Lockwood. 2010. “Arsenic Exposure in US Public and Domestic Drinking Water Supplies: A Comparative Risk Assessment.” *Journal of Exposure Science and Environmental Epidemiology* 20 (3). Nature Publishing Group: 245–54. doi:10.1038/jes.2009.24.
- Kundzewicz, Zbigniew W., and Petra Döll. 2009. “Will Groundwater Ease Freshwater Stress under Climate Change?” *Hydrological Sciences Journal* 54 (4): 665–75.

doi:10.1623/hysj.54.4.665.

- Lee, Jin-Jing, Cheng-Shin Jang, Sheng-Wei Wang, and Chen-Wuing Liu. 2007. "Evaluation of Potential Health Risk of Arsenic-Affected Groundwater Using Indicator Kriging and Dose Response Model." *Science of The Total Environment* 384 (1–3): 151–62. doi:10.1016/j.scitotenv.2007.06.021.
- Lee, Jin Soo, Hyo Taek Chon, and Kyoung Woong Kim. 2005. "Human Risk Assessment of As, Cd, Cu and Zn in the Abandoned Metal Mine Site." *Environmental Geochemistry and Health* 27 (2): 185–91. doi:10.1007/s10653-005-0131-6.
- Lee, Lukas Jyuhn-Hsiarn, Chien-Hung Chen, Yu-Yin Chang, Saou-Hsing Liou, and Jung-Der Wang. 2010. "An Estimation of the Health Impact of Groundwater Pollution Caused by Dumping of Chlorinated Solvents." *Science of The Total Environment* 408 (6). Elsevier B.V.: 1271–75. doi:10.1016/j.scitotenv.2009.12.036.
- Lee, Sang-woo, Byung-tae Lee, Ju-yong Kim, Kyoung-woong Kim, and Jin-soo Lee. 2006. "Human Risk Assessment for Heavy Metals and as Contamination in the Abandoned Metal Mine Areas, Korea." *Environmental Monitoring and Assessment* 119 (1–3): 233–44. doi:10.1007/s10661-005-9024-5.
- Leung, Ho Wing, Ling Jin, Si Wei, Mirabelle Mei Po Tsui, Bingsheng Zhou, Liping Jiao, Pak Chuen Cheung, Yiu Kan Chun, Margaret Burkhardt Murphy, and Paul Kwan Sing Lam. 2013. "Pharmaceuticals in Tap Water: Human Health Risk Assessment and Proposed Monitoring Framework in China." *Environmental Health Perspectives* 121 (7): 839–46. doi:10.1289/ehp.1206244.
- Levac, Danielle, Heather Colquhoun, and Kelly K. O'Brien. 2010. "Scoping Studies: Advancing the Methodology." *Implementation Science* 5 (1): 69. doi:10.1186/1748-5908-5-69.
- Li, Jianbing, Gordon H. Huang, Guangming Zeng, Imran Maqsood, and Yuefei Huang. 2007. "An Integrated Fuzzy-Stochastic Modeling Approach for Risk Assessment of Groundwater Contamination." *Journal of Environmental Management* 82 (2): 173–88. doi:10.1016/j.jenvman.2005.12.018.
- Liu, Chen-Wuing, Chun-Nan Lin, Cheng-Shin Jang, Min-Pei Ling, and Jeng-Wei Tsai. 2011. "Assessing Nitrate Contamination and Its Potential Health Risk to Kinmen Residents." *Environmental Geochemistry and Health* 33 (5): 503–14. doi:10.1007/s10653-010-9367-x.
- Liu, Kevin Fong-Rey, Che-Fan Lu, Cheng-Wu Chen, and Yung-Shuen Shen. 2012. "Applying Bayesian Belief Networks to Health Risk Assessment." *Stochastic Environmental Research and Risk Assessment* 26 (3): 451–65. doi:10.1007/s00477-011-0470-z.
- Ma, Lixin, Xuebo Qin, Nan Sun, and Guoting Yang. 2014. "Human Health Risk of Metals in Drinking-Water Source Areas from a Forest Zone after Long-Term Excessive Deforestation." *Human and Ecological Risk Assessment: An International Journal* 20 (5): 1200–1212. doi:10.1080/10807039.2013.854134.
- Machdar, E., N. P. van der Steen, L. Raschid-Sally, and P. N. L. Lens. 2013. "Application of Quantitative Microbial Risk Assessment to Analyze the Public Health Risk from Poor

- Drinking Water Quality in a Low Income Area in Accra, Ghana.” *Science of The Total Environment* 449 (April). Elsevier B.V.: 134–42. doi:10.1016/j.scitotenv.2013.01.048.
- Maqsood, I. 2011. “Integrated Fuzzy-Stochastic Risk Assessment for Contaminated Groundwater Systems.” *International Journal of Risk Assessment and Management* 15 (1): 43–65.
- Marara, Tafadzwa, L. G. Palamuleni, and Eno E. Ebenso. 2013. “Chemical and Radiological Risks of Drinking Water from Communities in Wonderfonteinspruit Catchment, South Africa.” *Asian Journal of Chemistry* 25 (16): 9302–8. doi:10.14233/ajchem.2013.15502.
- Markon, Marie-Pierre L., and L. Lemyre. 2013. “Public Reactions to Risk Messages Communicating Different Sources of Uncertainty: An Experimental Test.” *Human and Ecological Risk Assessment: An International Journal* 19 (4): 1102–26. doi:10.1080/10807039.2012.702015.
- Momot, Olga, and Boris Synzynys. 2005. “Toxic Aluminium and Heavy Metals in Groundwater of Middle Russia: Health Risk Assessment.” *International Journal of Environmental Research and Public Health* 2 (2): 214–18. doi:10.3390/ijerph2005020003.
- Mondal, Debapriya, Mayukh Banerjee, Manjari Kundu, Nilanjana Banerjee, Udayan Bhattacharya, Ashok K. Giri, Bhaswati Ganguli, Sugata Sen Roy, and David A. Polya. 2010. “Comparison of Drinking Water, Raw Rice and Cooking of Rice as Arsenic Exposure Routes in Three Contrasting Areas of West Bengal, India.” *Environmental Geochemistry and Health* 32 (6): 463–77. doi:10.1007/s10653-010-9319-5.
- Munns, Wayne R., Robert Kroes, Gilman Veith, Glenn W. Suter II, Terri Damstra, and Michael D. Waters. 2003. “Approaches for Integrated Risk Assessment.” Edited by Intergovernmental Panel on Climate Change. *Human and Ecological Risk Assessment: An International Journal* 9 (1). Cambridge: Cambridge University Press: 267–72. doi:10.1080/713609863.
- Ni, Fuquan, Guodong Liu, Jian Ye, Huazhun Ren, and Shangchun Yang. 2009. “ArcGIS-Based Rural Drinking Water Quality Health Risk Assessment.” *Journal of Water Resource and Protection* 1 (5): 351–61. doi:10.4236/jwarp.2009.15042.
- Nsiah-Kumi, Phyllis A. 2008. “Communicating Effectively with Vulnerable Populations during Water Contamination Events.” *Journal of Water and Health* 6 (S1): s63. doi:10.2166/wh.2008.041.
- Nzihou, Jean Fidèle, Médard Bouda, Salou Hamidou, and Jean Diarra. 2013. “Arsenic in Drinking Water Toxicological Risk Assessment in the North Region of Burkina Faso.” *Journal of Water Resource and Protection* 5 (8): 46–52. doi:10.4236/jwarp.2013.58A007.
- O’Toole, Thomas P., Kaytura Felix Aaron, Marshall H. Chin, Carol Horowitz, and Frederick Tyson. 2003. “Community-Based Participatory Research.” *Journal of General Internal Medicine* 18 (7): 592–94. doi:10.1046/j.1525-1497.2003.30416.x.
- Obiri, S., D. K. Dodoo, F. A. Armah, D. K. Essumang, and S. J. Cobbina. 2010. “Evaluation of Lead and Mercury Neurotoxic Health Risk by Resident Children in the Obuasi Municipality, Ghana.” *Environmental Toxicology and Pharmacology* 29 (3). Elsevier B.V.: 209–12.

doi:10.1016/j.etap.2010.01.001.

- Papić, Petar, Marina Ćuk, Maja Todorović, Jana Stojković, Bojan Hajdin, Nebojša Atanacković, and Dušan Polomčić. 2012. "Arsenic in Tap Water of Serbia's South Pannonian Basin and Arsenic Risk Assessment." *Pol. J. Environ. Stud.* 21 (6): 1783–90.
- Peplow, Daniel, and Robert Edmonds. 2004. "Health Risks Associated with Contamination of Groundwater by Abandoned Mines Near Twisp in Okanogan County, Washington, USA." *Environmental Geochemistry and Health* 26 (1): 69–79.  
doi:10.1023/B:EGAH.0000020974.52087.cb.
- Pham, Mai T., Andrijana Rajić, Judy D. Greig, Jan M. Sargeant, Andrew Papadopoulos, and Scott A. McEwen. 2014. "A Scoping Review of Scoping Reviews: Advancing the Approach and Enhancing the Consistency." *Research Synthesis Methods* 5 (4): 371–85.  
doi:10.1002/jrsm.1123.
- Phan, Kongkea, Samrach Phan, Laingshun Huoy, Bunseang Suy, Ming Hung Wong, Jamal Hisham Hashim, Mohamed Salleh Mohamed Yasin, Syed Mohamed Aljunid, Suthipong Sthiannopkao, and Kyoung-Woong Kim. 2013. "Assessing Mixed Trace Elements in Groundwater and Their Health Risk of Residents Living in the Mekong River Basin of Cambodia." *Environmental Pollution* 182 (November). Elsevier Ltd: 111–19.  
doi:10.1016/j.envpol.2013.07.002.
- Pons, Wendy, Ian Young, Jenifer Truong, Andria Jones-Bitton, Scott McEwen, Katarina Pintar, and Andrew Papadopoulos. 2015. "A Systematic Review of Waterborne Disease Outbreaks Associated with Small Non-Community Drinking Water Systems in Canada and the United States." Edited by Paula V Morais. *PLOS ONE* 10 (10): e0141646.  
doi:10.1371/journal.pone.0141646.
- Qiao, Meng, Taicheng An, Xiangying Zeng, Delin Zhang, Guiying Li, Guoying Sheng, Jiamo Fu, Guoxia Zhang, and Jun Guo. 2010. "Safety Assessment of the Source Water within the Pearl River Delta on the Aspect of Organochlorine Pesticides Contamination." *Journal of Environmental Monitoring* 12 (9): 1666. doi:10.1039/c002019b.
- Rajagopal, R., and Richard L. Talcott. 1983. "Patterns in Groundwater Quality: Selected Observations in Iowa." *Environmental Management* 7 (5): 465–73.  
doi:10.1007/BF01867126.
- Ramirez-Andreotta, Monica D., Mark L. Brusseau, Paloma Beamer, and Raina M. Maier. 2013. "Home Gardening near a Mining Site in an Arsenic-Endemic Region of Arizona: Assessing Arsenic Exposure Dose and Risk via Ingestion of Home Garden Vegetables, Soils, and Water." *Science of The Total Environment* 454–455 (June). Elsevier B.V.: 373–82.  
doi:10.1016/j.scitotenv.2013.02.063.
- Rapant, Stanislav, and Katarína Krčmová. 2007. "Health Risk Assessment Maps for Arsenic Groundwater Content: Application of National Geochemical Databases." *Environmental Geochemistry and Health* 29 (2): 131–41. doi:10.1007/s10653-006-9072-y.
- Razzolini, Maria Tereza Pepe, Mark H. Weir, Maria Helena Matte, Glavur Rogerio Matte, Licia



- Natal Fernandes, and Joan B. Rose. 2011. "Risk of Giardia Infection for Drinking Water and Bathing in a Peri-Urban Area in São Paulo, Brazil." *International Journal of Environmental Health Research* 21 (3): 222–34. doi:10.1080/09603123.2010.533367.
- Renn, Ortwin. 1998. "Three Decades of Risk Research: Accomplishments and New Challenges." *Journal of Risk Research* 1 (1): 49–71. doi:10.1080/136698798377321.
- Ritter, Len, Keith Solomon, Paul Sibley, Ken Hall, Patricia Keen, Gevan Mattu, and Beth Linton. 2002. "Sources, Pathways, and Relative Risks of Contaminants in Surface Water and Groundwater: A Perspective Prepared for the Walkerton Inquiry." *Journal of Toxicology and Environmental Health, Part A* 65 (1): 1–142. doi:10.1080/152873902753338572.
- Ryan, Louise. 2003. "Epidemiologically Based Environmental Risk Assessment." *Statistical Science* 18 (4): 466–80. doi:10.1214/ss/1081443230.
- Santos, Miriam dos Anjos, Bruno Esteves Tavora, Sergio Koide, and Eloisa Dutra Caldas. 2013. "Human Risk Assessment of Benzene after a Gasoline Station Fuel Leak." *Revista de Saúde Pública* 47 (2): 335–44. doi:10.1590/S0034-8910.2013047004381.
- Sekizawa, Jun, and Shinsuke Tanabe. 2005. "A Comparison between Integrated Risk Assessment and Classical Health/environmental Assessment: Emerging Beneficial Properties." *Toxicology and Applied Pharmacology* 207 (2): 617–22. doi:10.1016/j.taap.2005.01.047.
- Serre, M. L., A. Kolovos, G. Christakos, and K. Modis. 2003. "An Application of the Holistochastic Human Exposure Methodology to Naturally Occurring Arsenic in Bangladesh Drinking Water." *Risk Analysis* 23 (3): 515–28. doi:10.1111/1539-6924.t01-1-00332.
- Sexton, Ken, and Stephen H. Linder. 2014. "Integrated Assessment of Risk and Sustainability in the Context of Regulatory Decision Making." *Environmental Science & Technology* 48 (3): 1409–18. doi:10.1021/es4043066.
- Shaheed, Ameer, Jennifer Orgill, Maggie A. Montgomery, Marc A. Jeuland, and Joe Brown. 2014. "Why 'improved' Water Sources Are Not Always Safe." *Bulletin of the World Health Organization* 92 (4): 283–89. doi:10.2471/BLT.13.119594.
- Singh, S. K., A. K. Ghosh, A. Kumar, K. Kislay, R. R. Tiwari, R. Parwez, N. Kumar, and M. D. Imam. 2014. "Groundwater Arsenic Contamination and Associated Health Risks in Bihar, India." *International Journal of Environmental Research* 8 (1): 49–60. www.SID.ir.
- Singh, Sushant Kumar, and Ashok Kumar Ghosh. 2012. "Health Risk Assessment Due to Groundwater Arsenic Contamination: Children Are at High Risk." *Human and Ecological Risk Assessment: An International Journal* 18 (4): 751–66. doi:10.1080/10807039.2012.688700.
- Slovic, Paul. 1999. "Trust, Emotion, Sex, Politics, and Science: Surveying the Risk-Assessment Battlefield." *Risk Analysis* 19 (4): 689–701. doi:10.1111/j.1539-6924.1999.tb00439.x.
- Steyn, M., P. Jagals, and B. Genthe. 2004. "Assessment of Microbial Infection Risks Posed by Ingestion of Water during Domestic Water Use and Full-Contact Recreation in a Mid

- Southern African Region.” *Water Science & Technology* 50 (1): 301–8.
- Su, Xiaosi, Huang Wang, and Yuling Zhang. 2013. “Health Risk Assessment of Nitrate Contamination in Groundwater: A Case Study of an Agricultural Area in Northeast China.” *Water Resources Management* 27 (8): 3025–34. doi:10.1007/s11269-013-0330-3.
- Sultana, Jawairia, Abida Farooqi, and Usman Ali. 2014. “Arsenic Concentration Variability, Health Risk Assessment, and Source Identification Using Multivariate Analysis in Selected Villages of Public Water System, Lahore, Pakistan.” *Environmental Monitoring and Assessment* 186 (2): 1241–51. doi:10.1007/s10661-013-3453-3.
- Suter II, Glenn W., Theo Vermeire, Wayne R. Munns Jr., and Jun Sekizawa. 2005. “An Integrated Framework for Health and Ecological Risk Assessment.” *Toxicology and Applied Pharmacology* 207 (2): 611–16. doi:10.1016/j.taap.2005.01.051.
- Titilayo, E. A., A. G. Olufemi, and E. T. Ogbemi. 2012. “Chronic Exposure to Heavy Metals in Public Water Supply and Human Health Risk Assessment.” *Terrestrial and Aquatic Environmental Toxicology* 6 (2): 106–11.
- Törnqvist, Rebecka, Jerker Jarsjö, and Bakhtiyor Karimov. 2011. “Health Risks from Large-Scale Water Pollution: Trends in Central Asia.” *Environment International* 37 (2). Elsevier Ltd: 435–42. doi:10.1016/j.envint.2010.11.006.
- Ujević Bošnjak, M., K. Capak, A. Jazbec, C. Casiot, L. Sipos, V. Poljak, and Ž. Dadić. 2012. “Hydrochemical Characterization of Arsenic Contaminated Alluvial Aquifers in Eastern Croatia Using Multivariate Statistical Techniques and Arsenic Risk Assessment.” *Science of The Total Environment* 420 (March): 100–110. doi:10.1016/j.scitotenv.2012.01.021.
- United States Environmental Protection Agency (US EPA). 1989. “Risk Assessment Guidance for Superfund. Volume I Human Health Evaluation Manual (Part A)” I (December): 289. doi:EPA/540/1-89/002.
- . 2001. “Process for Conducting Probabilistic Risk Assessment.” In *Risk Assessment Guidance for Superfund: Volume III - Part A*, 3rd ed., 3:1–34. Washington, DC: Office of Emergency and Remedial Response, U.S. Environmental Protection Agency. [www.epa.gov/superfund/RAGS3A/index.htm](http://www.epa.gov/superfund/RAGS3A/index.htm).
- . 2015a. “Human Health Risk Assessment.” <https://www.epa.gov/risk/human-health-risk-assessment>.
- . 2015b. “Probabilistic Risk Assessment White Paper and Supporting Documents.” <https://www.epa.gov/osa/probabilistic-risk-assessment-white-paper-and-supporting-documents>.
- Vermeire, Theo, Wayne R. Munns, Jun Sekizawa, Glenn Suter, and Glen Van der Kraak. 2007. “An Assessment of Integrated Risk Assessment.” *Human and Ecological Risk Assessment: An International Journal* 13 (2): 339–54. doi:10.1080/10807030701226848.
- Villanueva, Cristina M., Manolis Kogevas, Sylvaine Cordier, Michael R. Templeton, Roel Vermeulen, John R. Nuckols, Mark J. Nieuwenhuijsen, and Patrick Levallois. 2013.

- “Assessing Exposure and Health Consequences of Chemicals in Drinking Water: Current State of Knowledge and Research Needs.” *Environmental Health Perspectives* 122 (3): 213–21. doi:10.1289/ehp.1206229.
- Wang, B., G. Yu, Y. J. Yu, J. Huang, H. Y. Hu, and L. S. Wang. 2009. “Health Risk Assessment of Organic Pollutants in Jiangsu Reach of the Huaihe River, China.” *Water Science & Technology* 59 (5): 907. doi:10.2166/wst.2009.038.
- Wang, Zhenxing, Liyuan Chai, Yunyan Wang, Zhihui Yang, Haiying Wang, and Xie Wu. 2011. “Potential Health Risk of Arsenic and Cadmium in Groundwater near Xiangjiang River, China: A Case Study for Risk Assessment and Management of Toxic Substances.” *Environmental Monitoring and Assessment* 175 (1–4): 167–73. doi:10.1007/s10661-010-1503-7.
- WHO. 2013. “Water Quality and Health Strategy 2013–2020.” New York.
- WHO/IPCS. 2001. “Framework for the Integration of Health and Ecological Risk Assessment.” *Integrated Risk Assessment*. Vol. II. Geneva, Switzerland. doi:WHO/IPCS/IRA/01/12.
- WHO/UNICEF. 2012. “Progress on Drinking Water and Sanitation: 2012 Update.” *New York, NY: United Nation’s Children’s Fund and World Health Organisation*. doi:978-924-1503279.
- . 2015. “25 Years Progress on Sanitation and Drinking Water: 2015 Update and MDG Assessment.” *New York, NY: United Nation’s Children’s Fund and World Health Organisation*. [http://www.who.int/water\\_sanitation\\_health/publications/jmp-2015-update/en/](http://www.who.int/water_sanitation_health/publications/jmp-2015-update/en/).
- Wilks, M. F., N. Roth, L. Aicher, M. Faust, P. Papadaki, A. Marchis, M. Calliera, et al. 2015. “White Paper on the Promotion of an Integrated Risk Assessment Concept in European Regulatory Frameworks for Chemicals.” *Science of The Total Environment* 521–522 (1). Elsevier B.V.: 211–18. doi:10.1016/j.scitotenv.2015.03.065.
- Williams, Pamela, Laurie Benton, John Warmerdam, and Patrick Sheehan. 2002. “Comparative Risk Analysis of Six Volatile Organic Compounds in California Drinking Water.” *Environmental Science & Technology* 36 (22): 4721–28. doi:10.1021/es020725y.
- Williams, Pamela R. D., Paul K. Scott, Patrick J. Sheehan, and Dennis J. Paustenbach. 2000. “A Probabilistic Assessment of Household Exposures to MTBE from Drinking Water.” *Human and Ecological Risk Assessment: An International Journal* 6 (5): 827–49. doi:10.1080/10807030091124220.
- World Bank. 2015. “The World Bank: Data.” *The World Bank: Data*. <http://data.worldbank.org/indicator/SP.RUR.TOTL.ZS>.
- Wu, Bing, Yan Zhang, Xuxiang Zhang, and Shupeì Cheng. 2010. “Health Risk from Exposure of Organic Pollutants Through Drinking Water Consumption in Nanjing, China.” *Bulletin of Environmental Contamination and Toxicology* 84 (1): 46–50. doi:10.1007/s00128-009-9900-8.

- Wu, Chunfa, Yongming Luo, Tong Gui, and Yujuan Huang. 2014. "Concentrations and Potential Health Hazards of Organochlorine Pesticides in Shallow Groundwater of Taihu Lake Region, China." *Science of The Total Environment* 470–471 (February). Elsevier B.V.: 1047–55. doi:10.1016/j.scitotenv.2013.10.056.
- Yacoub, C., N. Blazquez, A. Pérez-Foguet, and N. Miralles. 2013. "Spatial and Temporal Trace Metal Distribution of a Peruvian Basin: Recognizing Trace Metal Sources and Assessing the Potential Risk." *Environmental Monitoring and Assessment* 185 (10): 7961–78. doi:10.1007/s10661-013-3147-x.
- Yu, Feng Cun, Guo Hua Fang, and Xiang Wen Ru. 2010. "Eutrophication, Health Risk Assessment and Spatial Analysis of Water Quality in Gucheng Lake, China." *Environmental Earth Sciences* 59 (8): 1741–48. doi:10.1007/s12665-009-0156-8.
- Zargar, Amin, Roberta Dyck, M. Shafiqul Islam, Asish Mohapatra, and Rehan Sadiq. 2014. "Data Fusion Methods for Human Health Risk Assessment: Review and Application." *Human and Ecological Risk Assessment: An International Journal* 20 (3): 807–38. doi:10.1080/10807039.2012.746145.
- Zhang, Junfeng, Denise L. Mauzerall, Tong Zhu, Song Liang, Majid Ezzati, and Justin V. Remais. 2010. "Environmental Health in China: Progress towards Clean Air and Safe Water." *The Lancet* 375 (9720). Elsevier Ltd: 1110–19. doi:10.1016/S0140-6736(10)60062-1.
- Zhao, Juan, Rui Wang, Jian Min Bian, and Yang Zhao. 2012. "Potential Health Risk of Arsenic in Groundwater near Tongyu County, Western of Jilin Province: A Case Study for Health Risk Assessment Based on Triangular Fuzzy Number." *Advanced Materials Research* 518–523 (May): 982–86. doi:10.4028/www.scientific.net/AMR.518-523.982.
- Zheng, Jing, Ke-Hui Chen, Xiao Yan, She-Jun Chen, Guo-Cheng Hu, Xiao-Wu Peng, Jian-Gang Yuan, Bi-Xian Mai, and Zhong-Yi Yang. 2013. "Heavy Metals in Food, House Dust, and Water from an E-Waste Recycling Area in South China and the Potential Risk to Human Health." *Ecotoxicology and Environmental Safety* 96 (October). Elsevier: 205–12. doi:10.1016/j.ecoenv.2013.06.017.

### 3 Risk Perception and Human Health Risk in Rural Communities Consuming Unregulated Well Water in Saskatchewan, Canada

This manuscript is published in *Risk Analysis an International Journal* on June 3<sup>rd</sup>, 2019:  
<https://onlinelibrary.wiley.com/doi/10.1111/risa.13335> doi: <https://doi.org/10.1111/risa.13335>

Ford, Lorelei, Cheryl Waldner, Javier Sanchez and Lalita Bharadwaj. 2019. "Risk Perception and Human Health Risk in Rural Communities Consuming Unregulated Well Water in Saskatchewan, Canada." *Risk Analysis an International Journal* doi: 10.1111/risa.13335

This manuscript reproduced under licence with minor modifications for formatting. Author contributions are as follows: Lorelei Ford participated in the entire process of the research including: design, data collection, data analysis, interpretation of results, writing, and editing the manuscript. Cheryl Waldner and Lalita Bharadwaj participated in the design, data analysis, interpretation of results, and editing. Javier Sanchez provided training and consultation on Bayesian risk assessment and editing.

### 3.1 Abstract

Rural communities dependent on unregulated drinking water are potentially at increased health risk from exposure to contaminants. Perception of drinking water safety influences water consumption, exposure, and health risk. A community-based participatory approach and probabilistic Bayesian methods were applied to integrate risk perception in a holistic human health risk assessment. Tap water arsenic concentrations and risk perception data were collected from two Saskatchewan communities. Drinking water health standards were exceeded in 67% (51/76) of households in the Rural Municipality #184 (RM184) and 56% (25/45) in Beardy's and Okemasis First Nation (BOFN). There was no association between the presence of a health exceedance and risk perception. Households in RM184 or with an annual income > \$50,000 were most likely to have in-house water treatment. The probability of consuming tap water perceived as safe (92%) or not safe (0%) suggested households in RM184 were unlikely to drink water perceived as not safe. The probability of drinking tap water perceived as safe (77%) or as not safe (11%) suggested households in BOFN contradicted their perception and consumed water perceived as unsafe. Integration of risk perception lowered the adult Incremental Lifetime Cancer Risk by 3% to  $1.3 \times 10^{-5}$  (95% CI  $8.4 \times 10^{-8}$  to  $9.0 \times 10^{-5}$ ) for RM184 and  $8.9 \times 10^{-6}$  (95% CI  $2.2 \times 10^{-7}$  to  $5.9 \times 10^{-5}$ ) for BOFN. Probability of exposure to arsenic concentrations > 1:100,000 negligible cancer risk was 23% for RM184 and 22% for BOFN.

### 3.2 Introduction

Rural populations represent 46% of the global population and depend on unregulated sources of drinking water (World Health Organization [WHO] & United Nations International Children's Emergency Fund [UNICEF] 2015). Unregulated water sources lack regular monitoring to ensure drinking water quality meet standards for human consumption (Shaheed et al. 2015; WHO and UNICEF 2015). In the absence of effective water regulation, water users bear the responsibility of protecting water sources and ensuring water is safe to drink; however, they often lack the knowledge and resources required to effectively manage drinking water risks (Charrois 2010; Hynds et al. 2013; Maxwell et al. 1998; Shaheed et al. 2014). Concerns about the presence of hazards in unregulated drinking water sources and the potential for associated health risks in vulnerable rural populations have been identified (Charrois 2010; Fox et al. 2016; Hynds et al. 2013; Pons et al. 2015; Simpson 2010; Villanueva et al. 2013).

Approximately 43 million Americans and 6 million Canadians access unregulated groundwater (Fox et al. 2016; Statistics Canada 1996). In Canada, rural and First Nation communities frequently rely on and fail to manage potential health risks associated with unregulated drinking water due to the lack of education, monitoring, and effective treatment of household or private wells (Charrois 2010; Corkal et al. 2004; Jones et al. 2005; Spence and Walters 2012). Interestingly, unregulated drinking water sources pose similar barriers globally (Schwarzenbach et al. 2010; WHO and UNICEF 2015). A lack of financial resources and increased vulnerability due to poverty, illness, and minority status, make it more difficult for households or communities to cope with drinking water management (Nsiah-Kumi 2008; Wescoat et al. 2007; Zheng and Ayotte 2015).

In the absence of effective drinking water monitoring and management, consumers' perception of drinking water safety or risk influence consumption (Charrois 2010; Hynds et al. 2013; Maxwell et al. 1998; Shaheed et al. 2014). Consumers develop perceptions of risk through intuition and judgement formed by their experiences (Slovic 1987), and a lack of correlation between perception of drinking water risk and actual health risk can potentially lead to overuse of unsafe water sources (Chen et al. 2012; Hynds et al. 2013; Maxwell et al. 1998; Orgill et al. 2013; Patrick 2011; Turgeon et al. 2004). Conversely, avoidance of safe drinking water, perceived as unsafe, may increase consumption of alternative unsafe water sources or sugary beverages (Dupont et al. 2010; S. J. Onufrak et al. 2014; S. Onufrak et al. 2012). Factors contributing to drinking water risk perception are complex and can vary considerably by community (Doria 2010; Dupont et al. 2014; Spence and Walters 2012). Characterization of drinking water risk perception could provide a better understanding of uncertainty associated with the amount of water consumed and risk of exposure (Chowdhury et al. 2009; Doria 2010; Wright et al. 2018). For example, Wright et al. (2018) have explored perceptions associated with tap water consumption in Canadian Inuit communities.

Perception of risk associated with drinking water is an example of a ‘non-traditional’ variable for the assessment of exposure and human health risk. Risk perception, economic, social, and human behaviour are other non-traditional variables influencing human health risk assessment (HHRA) according to Ryan (2003) and Wilks et al. (2015). Bayesian risk models can enhance traditional quantitative HHRA by including community knowledge and providing a measure of uncertainty around the estimates of risk that are important for public health policy and planning (Ritter et al. 2002; Serre et al. 2003). A recent scoping review found HHRA publications from 2000 to early 2014 did not include non-traditional variables or quantitatively integrate them in risk assessments of unregulated drinking water (Ford et al. 2017).

Pressure on governments to include human perceptions in sustainable water resource and drinking water management (Jackson 2006; Jackson et al. 2012; Ochoo et al. 2017; Wright et al. 2018) make the integration of perception in risk assessment necessary. Suter II et al. (2005) predicted that effective HHRA would require an integrative approach defined by collaborative, place-based risk assessments that integrate additional data to inform direct and indirect risks. Better informed risk assessment supports risk management by providing scalable assessment relevant to communities (Suter II et al. 2005).

The inclusion of risk perception in HHRA, as it relates to unregulated drinking water, may contribute uncertain knowledge for a more holistic and reliable determination of risk. Understanding the effect of risk perception on human health risk could improve drinking water risk management and communication (Markon and Lemyre 2013). Serre et al. (2003) wrote: “Uncertain knowledge obtained about important exposure parameters could be more valuable than the certain knowledge obtained about less important parameters.”

### 3.3 Purpose and Objectives

The primary objective of this case study was to examine the impact of risk perception on human health risk associated with exposure to arsenic in unregulated well water using Bayesian methods. Secondary objectives were to determine factors influencing the presence of household water treatment and if current tap water safety can be predicted by risk perception and historical knowledge of water quality.

### 3.4 Methods

#### 3.4.1 Research Approach and Community Partnerships

A community-based participatory (CBP) study, utilizing probabilistic Bayesian risk assessment methods, (Serre et al. 2003; Zargar et al. 2014) was applied to integrate drinking water risk perception in a quantitative holistic HHRA. Probabilistic Bayesian methods were used to describe variability of model parameters and quantify perception of drinking water safety. A CBP approach was chosen to inform and support safe drinking water management. Previous



engagement, research and work experience with community residents led to established protocols of engagement and formalization of community research partnerships.

Partnerships with the Rural Municipality #184 (RM184) and Beardy's and Okemasis First Nation (BOFN) provided opportunity for households to receive free water testing, educational materials, and a well assessment. Community members provided guidance and feedback on the purpose and goals of the project, survey development and actively participated in data collection. The study was approved by the University of Saskatchewan Behavioral Research Ethics Board (#Beh14-108, April 14th, 2014).

### 3.4.2 Study Area and Potential Drinking Water Hazards

Private well owners in the RM184 (50°38'05"N 102°37'01"W) and residents with access to a well in BOFN (52°49'56"N 106°17'43"W), Saskatchewan, Canada (Fig. 3.1) participated. Southern Saskatchewan is located within the Western Canada Sedimentary Basin comprised of Precambrian rocks layered under 'flat-lying sedimentary strata' as described by (Maathuis 2008a). Domestic groundwater wells are located within the upper glacial sediments and water quality is aesthetically poor due to high sulphate, sodium, chloride, hardness, iron, manganese, and total dissolved solids (Maathuis 2000, 2008a).

Saskatchewan is known to have natural arsenic concentrations that exceed drinking water guidelines (Maathuis 2008b). Arsenic poses a drinking water hazard and is categorized as a Group 1 carcinogen associated with lung, bladder and liver cancer (Health Canada [HC] 2006). Arsenic species in groundwater are likely to be As (III) speciation due to the pH and reduced environment within aquifers (Wang and Mulligan 2006). Furthermore, Wang et al. (2006) suggest assessment of arsenic exposure in rural populations for increased public awareness and health problem avoidance.

### 3.4.3 Community Participant Recruitment and Survey Data Collection

A census sampling method included all eligible participants recruited through partnerships with municipal government and Chief and Council. A minimum participant number representing the adult population for each community was determined. Inclusion criteria were: consenting adults aged 20 to 59 from households with a well plumbed to the house, a working well pump, the ability to bypass treatment and water storage, and primary residence in the community. Total population, number of adults aged 20 to 59, the number of households, and an estimation of households with access to wells was gathered from eHealth Saskatchewan, and the Rural Municipality and Band Offices. Eligible households were those that met inclusion criteria and non-respondents for which status of eligibility could not be determined.

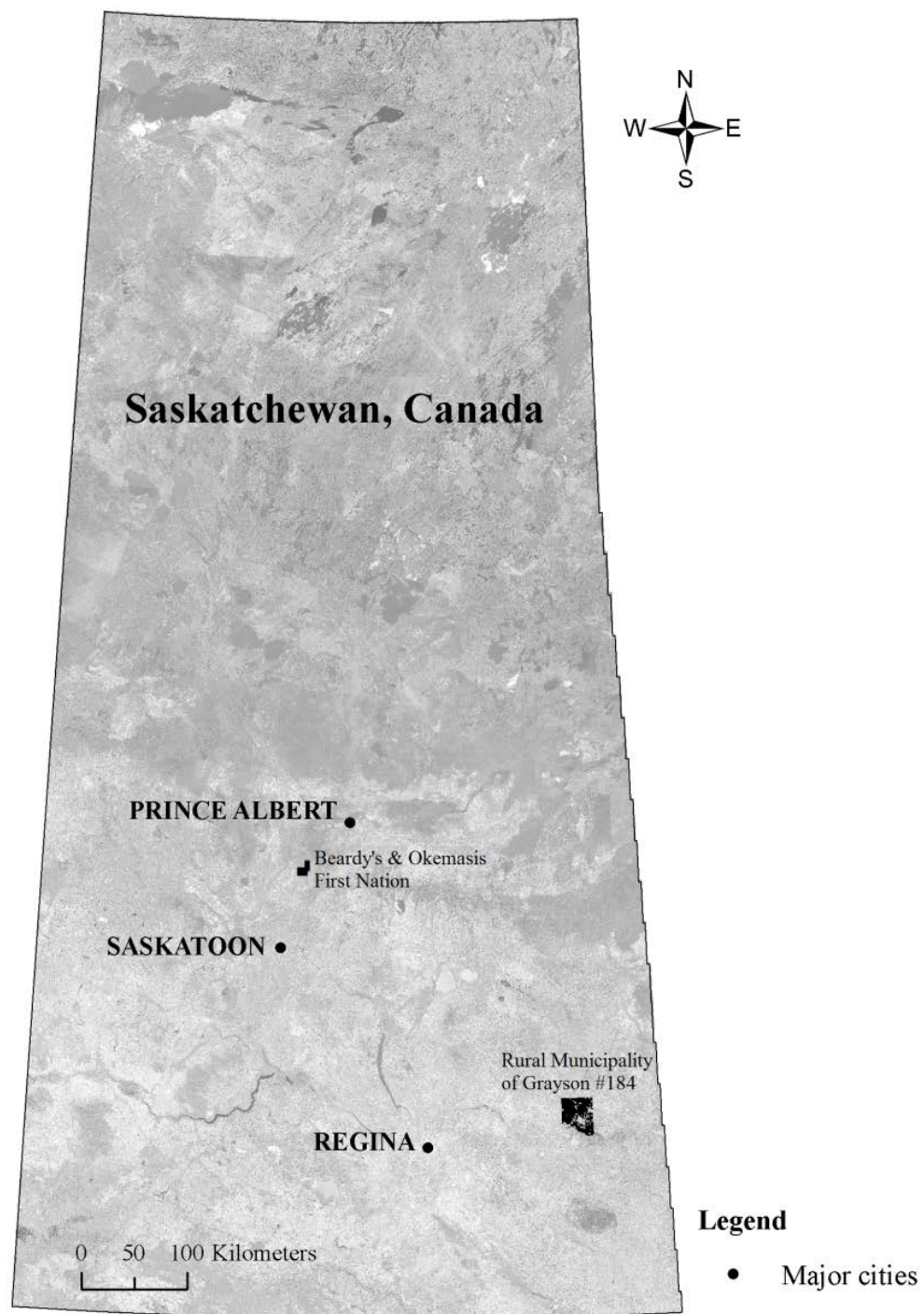


Figure 3.1 Participating rural communities include Beardy's and Okemasis First Nation and Rural Municipality of Grayson #184 in proximity to the major urban cities of Prince Albert, Saskatoon and Regina in the province of Saskatchewan, Canada.

The survey tool was piloted with 45 households outside of RM184 and BOFN. A Cohen's kappa ( $\kappa$ ) was used to determine intra-rater agreement between two telephone interviews, 30 days apart, for each participant (SPSS version 24, IBM Corp., Armonk, NY). Community household survey data was then collected by telephone or in-person in RM184, and in-person with a community member for BOFN from July to October 2014. Close-ended survey questions determined eligibility, household use and drinking water risk perception, well-user characteristics, previous testing, and water treatment for each household. Water treatment included equipment and technology that conditioned or treated drinking water. Survey questions are detailed in Supplementary Materials – 5.3 Household Survey.

#### 3.4.4 Hazard Data Collection

In-house tap water was collected to establish arsenic concentrations (hazard variable) for each community and HHRA application. Drinking water parameters analyzed and reported to each household included major ions, bacteria, and metals. Water quality testing took place from July to October 2014.

Water samples were collected in sealed plastic 250 mL sterile bottles containing sodium thiosulfate for bacteriological analysis, and plastic 500 mL bottles, tripled-rinsed with tap water, for chemical analysis. Water samples were collected as first-flush samples without prior tap disinfection.

Samples were transported on ice, within 48 hours, to the Roy Romanow Provincial Laboratory. Approximately 10% of samples collected were duplicated and sent to Saskatchewan Research Council to determine inter-laboratory variability for arsenic. Lin's Concordance Correlation Coefficient was used to measure lab agreement (Lin 2000; SPSS version 24, IBM Corp., Armonk, NY). Both laboratories are accredited by the Canadian Association for Laboratory Accreditation Inc. and conform to ISO/IEC 17025:2005 international standards.

Laboratory analytical methods were conducted according to the *Standard Methods for the Examination of Water and Wastewater* (Rice et al. 2012) with exception of the provincial lab's use of a modified ICP-MS method for metals (United States Environmental Protection Agency [US EPA] 1994). Water quality results were compared to *Saskatchewan's Drinking Water Quality Standards and Objectives* (Water Security Agency [WSA] 2017) to determine excursions.

#### 3.4.5 Human Health Risk Assessment

Probabilistic risk assessment using Bayesian inference with Gibbs sampling was performed based on Health Canada guidance methods (2010a, 2010b, 2010c).

#### 3.4.5.1 Problem formulation

Arsenic was identified as the chemical of concern. The assumption of potential risk was made prior to testing and was based on Maathuis (2008a) who identified 14.9% of Saskatchewan wells exceeded maximum acceptable arsenic concentrations of 10 µg/L. Oral exposure to arsenic through ingestion of drinking water was considered the primary exposure route (WHO 2011). Health Canada (2006) summarized the potential for increased estimated lifetime cancer risks associated with oral exposure to arsenic concentrations greater than 0.3 µg/L.

#### 3.4.5.2 Characterization of Risk Perception and Probability of Consuming Tap Water and Exposure Assessment

Risk perception was quantified by determining probability of well use given the water was perceived as safe ( $prob.s = P(\text{use}|\text{safe})$ ) or not safe ( $prob.ns = P(\text{use}|\text{not safe})$ ) and multiplying each by the corresponding proportion of the population, accounting for the entire population of well users. Together these probabilities sum to 1.0 and provide the probability tap water was used for drinking in the community ( $t.risk$ ), see Equation 3.1:

$$t.risk = (prop.s * prob.s) + (prop.ns * prob.ns) \quad (3.1)$$

Where:

$t.risk$  is the overall probability the water was used.

$prop.s$  is the proportion of households that perceived the well water was safe.

$prob.s$  is the probability that a household will drink well water if it was perceived as safe.

$prop.ns$  is the proportion of households that perceived the well water as not safe ( $1-prop.s$ ).

Data from questionnaires were summarized as beta distributions before inclusion in exposure assessment. Parameters for beta distribution for household results were as follows:  $\alpha = x + 1$  and  $\beta = n - x + 1$ , where there were  $x$  positive responses out of ' $n$ ' households.

Drinking water intake rate for adults in the Canadian population according to Health Canada is 1.5 L/day. Richardson's (1997) probability density function for Canadian daily tap water consumption rate (both sexes) was applied. The relative absorption factor ( $RAForal$ ) for arsenic in drinking water is 1, or 100% absorption (HC 2010c). Lifetime average daily dose ( $LADD$ ) was calculated according to Health Canada's Guidance on Human Health Preliminary quantitative Risk Assessment (HC 2010b) for ingestion of contaminated drinking water (Equation 3.2).

$$LADD (\mu g/kg bw/day) = \left( \frac{C_W \times IR_W \times RAForal \times D_2 \times D_3 \times D_4}{BW \times LE} \right) \quad (3.2)$$

Where:

$C_W$  = concentration of arsenic in drinking water ( $\mu\text{g/L}$ )

$IR_W$  = adult water intake rate (L/day)

$RAF_{oral}$  = relative absorption factor

$D_2$  = days per week exposed/7days

$D_3$  = weeks per year exposed/52 weeks

$D_4$  = total years exposed to water source

$BW$  = body weight (kg)

$LE$  = life expectancy (years)

#### 3.4.5.3 Toxicity Assessment and Risk Characterization

Incremental Lifetime Cancer Risk ( $ILCR$ ) was calculated using the LADD and arsenic Cancer Slope Factor ( $CSF$ ) of  $1800 \mu\text{g/kg/day}$  (HC 2010a; Equation 3.3). An essentially negligible  $ILCR$  of  $\leq 1$  in  $100,000$  ( $1 \times 10^{-5}$ ) was used as the threshold of acceptability for  $ILCR$  (HC 2010b), representing additional cancer cases per  $100,000$  people over their lifetime.

$$ILCR = LADD (\mu\text{g/kg bw/day}) \times CSF (\mu\text{g/kg/day})^{-1} \quad (3.3)$$

#### 3.4.6 Holistic Bayesian Human Health Risk Assessment

Probabilistic holistic Bayesian HHRA was used to integrate risk perception as a variable and to quantify uncertainty when determining the probability of  $ILCR$ . Bayesian inference was selected for modeling the HHRA due to its flexibility to integrate multiple data types, update prior understanding with future data, and quantify uncertainty through the use of Markov chain Monte Carlo sample simulations (Zargar et al. 2014; Tighe et al. 2013). The model was developed using OpenBUGS version 3.2.3 rev 1012 (Lunn et al. 2009) and the framework provided by Ames et al. (2005) was applied. The model facilitated the integration of perception by multiplying the intake rate ( $IR_W$ ) by the probability of use given well water perceived as safe or not safe to better reflect exposure. The full model in OpenBUGS can be referenced in Supplementary Materials – 5.4 Example of Model Code for Bayesian Human Health Risk Assessment.

Model convergence was qualitatively determined by visually monitoring the Brook-Gelman-Rubin diagram, history, quantiles, and trace plot convergence for three chains according to Spiegelhalter et al. (2012). Burn-in period was determined to be 30,000 followed by 90,000 iterations per chain for a total of 270,000 iterations. Sample iterations were confirmed sufficient with an MC error less than 5% of the standard deviation as stated in Spiegelhalter et al. (2012). Quantitative model convergence was confirmed with R-CODA package to confirm upper 97.5% of the scale reduction factor was less than 1.05 (Plummer et al. 2016; RStudio Team 2016).

Model parameters were, when appropriate, characterized as stochastic nodes represented by probability density distributions. Model parameters and their sources are listed in Table 3.1. Probabilistic parameters, and semi and non-informative priors used for mean and precision ( $\tau$ )

of model parameters are listed in Table 3.2. Precision was defined by  $1/\sigma^2$  where  $\sigma$  was the standard deviation. The semi-informative prior on the likelihood life expectancy (lif.exp) with a standard deviation of 10 and a normal distribution around the Canadian life expectancy of 80 years was developed using Parameter Solver, version 3.0 (MD Anderson Cancer Centre, Houston, TX; Table 3.2). Non-informative Jeffery's priors were used on the likelihood of chemical concentration (chem.c), water intake (h2o.in), and receptor weight (rep.t; Lunn et al. 2013). The human health risk assessment model was informed by conceptual diagram integrating risk perception, water intake and exposure (Fig. 3.2).

A step-function was used to determine the probability the ILCR exceeding 1 in 100,000 for households within community (Spiegelhalter et al. 2012). To assess difference in the model outputs, with and without perception, overlap of the probability density functions was determined for each community using R package 'overlapping' (Pastore 2017).

#### 3.4.7 Data Management and Analysis

Water quality and survey data were entered into Microsoft Access (Microsoft Access, Microsoft Corporation, Redmond, WA) and verified. Distributions for arsenic were determined using R package 'fitdistrplus' (Delignette-Muller et al. 2016). One-half the detection limit was applied to left censored data for deterministic statistics.

Within each community, Chi-squared tests were used to determine any associations between:

1. risk perception or awareness of a health exceedance, and the presence of at least one exceedance of a health guideline for drinking water safety; and,
2. risk perception and the presence of at least one exceedance of a drinking water aesthetic objective.

Generalized estimating equations with a logit link function and binomial distribution were used to evaluate potential risk factors for presence of in-house treatment. An exchangeable correlation coefficient accounted for clustering of responses within a community. Potential risk factors examined at 5% level of significance included associations between income, education, and the presence of a health or aesthetic-related drinking water quality exceedance. Community was considered as a fixed effect in the models to account for potential for unmeasured confounders. Interactions were not examined due to convergence issues associated with the sample being limited to two communities. All analyses were carried out in SPSS (SPSS version 24, IBM Corp., Armonk, NY).

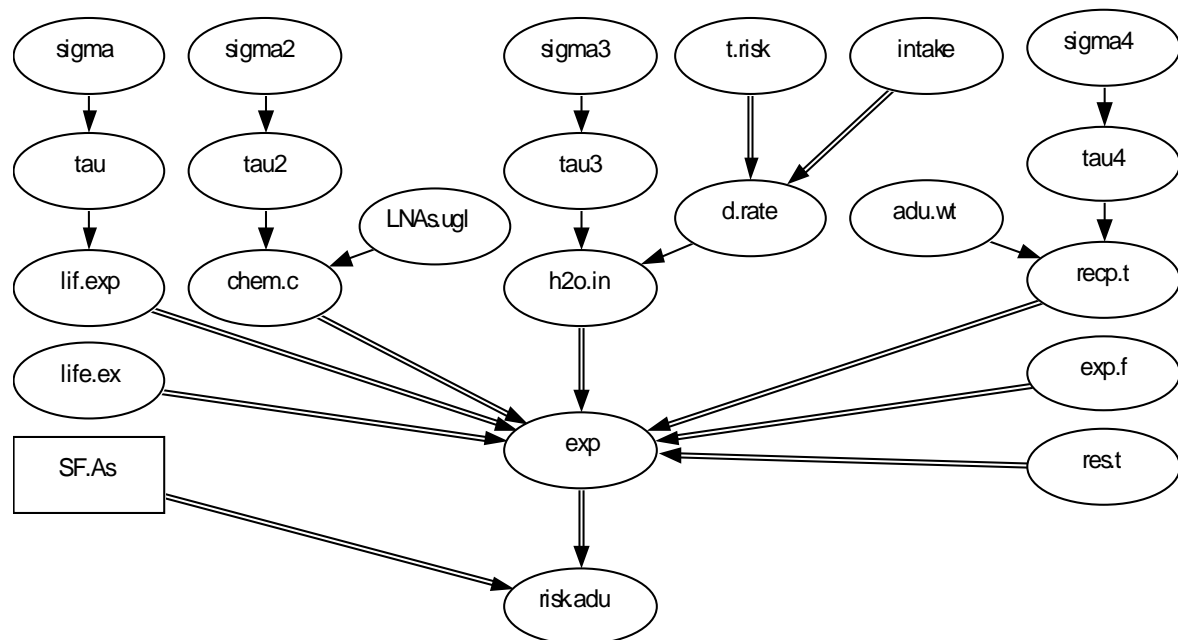


Figure 3.2 Directed acyclic graph for the holistic Bayesian human health risk assessment (HHRA) model to determine the incremental lifetime cancer risk (ILCR) from arsenic exposure for adults consuming unregulated well water, according to Health Canada (HC 2010a). Stochastic nodes, represented by a distribution, are oval shaped while the rectangle shape identifies a constant slope factor or cancer slope factor for arsenic (SF.As). Arrows with double lines indicate a logical relationship between nodes (i.e. multiplication), while single lines represent a stochastic relationship defined by a distribution. Life expectancy (lif.exp) in years has a semi-informative prior (i.e. tau and sigma). The concentration of arsenic (chem.c) is informed by the arsenic concentrations (LNAs.ugl) in µg/L from study households. An uninformative prior is placed on chem.c as to not influence the data (i.e. tau2 and sigma2). Drinking water intake (h2o.in) in litres per day is the drinking water rate (d.rate) and an uninformative prior (i.e. tau3 and sigma3). The overall probability the water was consumed (t.risk) was multiplied by the Canadian national drinking water rate (intake) to determine the influence of risk perception on d.rate. Receptor body weight (recp.t) in kilograms is informed by the national adult body weight (adu.wt) and an uninformative prior (i.e. tau4 and sigma4). All semi and uninformative priors allow for new data into the model or the model to be easily applied to another community. The exposure (exp) is then multiplied by the SF.As to determine the probability density function representing the ILCR for the community. Table 3.1 and 3.2 provide additional detail on the parameters of the model.

Table 3.1 Parameters, data sources and descriptions for holistic Bayesian human health risk assessment (HHRA).

Deterministic Parameters	Probabilistic Parameters	Unit	Data Type	Likelihood Distribution <sup>a</sup>	Reference
Chemical Concentration ( $C_W$ )	chem.c	µg/L	Continuous	LN(-0.207, 0.65) BOFN LN(-0.436, 0.36) RM184	Study
Water Intake Rate ( $IR_W$ )	h2o.in	L/day	Continuous	LN(0.28, 4.0)	Richardson 1997
Risk Perception (t.risk)	prob.s	No units	Binomial	$\beta(21, 7)$ BOFN $\beta(47, 5)$ RM184	Study
	prob.ns	No units	Binomial	$\beta(3, 17)$ BOFN $\beta(1, 27)$ RM184	Study
	prop.s	No units	Binomial	$\beta(27, 19)$ BOFN $\beta(51, 27)$ RM184	Study
	prop.ns	No units	Binomial	$\beta(19, 27)$ BOFN $\beta(27, 51)$ RM 184	Study
Relative Absorption Rate ( $RAF_{Oral}$ )	Not included	No units	Discrete	Assumed a constant of 1	HC 2010a; Schuhmacherwolz-Wolz et al. 2009
Exposure in Days per Week x Weeks per Year ( $D_2 \times D_3$ )	exp.f	days/year	Discrete	U(0.93, 0.96) between 340 to 350 days per year	Deng et al. 2012; Kentel and Aral 2004
Exposure in Years ( $D_4$ )	res.t	years	Discrete	U(20, 59)	Health Canada adult age category
Adult Body Weight ( $BW$ )	recp.t	kg	Continuous	LN(4.24, 25.00)	Richardson 1997
Life Expectancy ( $LE$ )	lif.exp	years	Constant	N(80, 10)	HC 2010b



Cancer Slope Factor (CSF)	SF.As	µg/kg/day	Constant	1800	HC 2010a
Lifetime Average Daily Dose (LADD)	exp	µg/kg bw/day	Logical Calculation	Determined by model	HC 2010b
Incremental Lifetime Cancer Risk (ILCR)	risk.adu	Chance of cancer risk	Logical Calculation	Determined by model	HC 2010b

<sup>a</sup> N,  $\beta$ , LN, and U means normal, beta, lognormal and uniform distribution. N and LN distributions defined by the mean and precision (tau) as required in OpenBUGS (Spiegelhalter et al. 2012).

Table 3.2 Priors on HHRA model parameters.

Probabilistic Parameters	Prior Type <sup>a</sup>	Prior Description <sup>b</sup>	Prior Distribution <sup>c</sup>
lif.exp	Semi-informative	SD = 10	$\sigma = \gamma(125, 12.5)$ , $\tau = 1/\sigma^2$
chem.c	Uninformative (Jeffery's)	$\mu = \text{LNAs.ugl}$ SD = $\sigma^2$	$\text{LNAs.ugl} = N(0, 0.00001)$ $\sigma^2 = \gamma(0.0001, 0.0001)$ , $\tau^2 = 1/\sigma^2$
h2o.in	Uninformative (Jeffery's)	$\mu = \text{intake}$ SD = $\sigma^3$	$\text{intake} = N(0, 0.00001)$ $\sigma^3 = \gamma(0.0001, 0.0001)$ , $\tau^3 = 1/\sigma^3$
recp.t	Uninformative (Jeffery's)	$\mu = \text{adu.wt}$ SD = $\sigma^4$	$\text{adu.wt} = N(0, 0.00001)$ $\sigma^4 = \gamma(0.0001, 0.0001)$ , $\tau^4 = 1/\sigma^4$

<sup>a</sup> Lunn et al. 2013 used to select priors; <sup>b</sup>  $\mu$ =mean, SD=standard deviation; <sup>c</sup>  $\gamma$  (gamma) and N (normal) distribution

### 3.5 Results

#### 3.5.1 Community Profile and Study Survey

A total of 121 households participated in the study from RM184 ( $n = 76$ ) and BOFN ( $n = 44$ ); Table 3.3). Participation rate relative to the number of eligible households with wells was 62% (76/123) for RM184 and 94% (44/47) for BOFN. One household from BOFN did not complete a survey, resulting in 44 survey participants and 45 collected water samples. A minimum sample size of adults required to represent the household for perception of risk with a confidence interval of 95% and proportion of 0.5 was  $n = 81$  with 5% margin of error for RM184, and  $n = 77$  with 2% margin of error for BOFN. The study identified a total of  $n = 80$  adults in RM184 and  $n = 79$  adults in BOFN residing surveyed households. Non-participating respondents from RM184 were not eligible primarily due to a lack of plumbing to the well or interest in the study. Table 3.3 provides community and household characteristics registered by *eHealth Saskatchewan* (2014), the participating communities, and the study.

The estimated average water consumption rate of 1.0 L/day in RM184 and 1.5 L/day in BOFN were within the Canadian average of  $1.5 \pm \text{SD } 0.8$  L/day for adults aged 20 to 59 years of age (Richardson 1997). Household well use and drinking water risk perception were similar within each community and approximately half of the residents were dependent on unregulated groundwater (Table 3.3). A large percentage (73%) of households in BOFN had used bottled water as an alternative water source. Only 47% of households in RM184 used alternative water sources including purchased bottled water, and water hauled from a municipal treatment facility or neighbour's well.

Most households in RM184 had water treatment (88%, 67/76) with a high percentage having previously tested their water source (84%, 64/76). Almost all households (93%, 41/44) on BOFN were aware of previous water testing; however, in-house water treatment was only present in 18% (8/44) of households. The presence of water treatment in households varied ( $p < 0.001$ ) based on income and also between communities. Households were more likely to use treatment if they were in RM184 as compared to BOFN (OR = 18.7; 95%CI 16.2 to 21.5). Households from both communities were more likely to use water treatment if the household annual income was greater than \$50,000 (OR = 4.6; 95%CI 2.9 to 7.1). Level of education was not a significant predictor of household water treatment ( $p = 0.12$ ) when controlling for income.

A smaller percentage of households in RM184 (24%, 18/76) and BOFN (7%, 3/44) were aware of previous water quality testing for metals (Table 3.3). Almost half (47%) of BOFN households surveyed recalled a notice to not consume their well water, while fewer than 10% of RM184 households could recall water quality results that identified the water unsafe to consume. There was no significant relationship between household awareness of a previous water quality health

exceedance or ‘do not consume’ notification and the presence of a drinking water quality health exceedance found in the study for RM184 ( $p = 0.27$ ) and BOFN ( $p = 0.82$ ).

Table 3.3 Characteristics of the community and study population informed by the 2014 Saskatchewan Health Survey, the communities, and the study survey. Where relevant, the total number of household survey answers is followed by percentage or mean and standard deviation (SD) are provided.

Characteristic	RM184	BOFN
<i>Saskatchewan Health (2014)</i>		
Total population	197	1821
Total adult population (incl. those not accessing wells)	101	79
<i>Communities</i>		
Households with access to wells	147	50
<i>Study</i>		
Eligible households with wells (incl. non-responses)	123	47
Total households participating	76	44 <sup>a</sup>
Total study population	190	185
Total study adult population	80	79
Households with income $\geq$ \$50,000	38 (50%)	4 (9%)
Households with highest education $\geq$ high school	58 (76%)	24 (55%)
Mean $\pm$ SD of residence time (years)	28 $\pm$ 17 ( $n = 73$ )	19 $\pm$ 13 ( $n = 43$ )
Mean $\pm$ SD of water consumption rate (L/day)	1.0 $\pm$ 1.2 ( $n = 54$ )	1.5 $\pm$ 1.5 ( $n = 44$ )
Number of people using water for drinking	46 (61%)	22 (50%)
Number of people that perceived the water as safe for drinking	50 (66%)	26 (59%)
Number of people using alternate water for drinking	36 (47%)	32 (73%)
Water Treatment	67 (88%)	8 (18%)
Previous water testing	64 (84%)	41 (93%)
Previous water testing for metals	18 (24%)	3 (7%)
Recalled results or recieved do not consume notice <sup>b</sup>	7 (9%; $n = 76$ )	20 (47%; $n = 43$ )

<sup>a</sup> arsenic concentration was determined for 45 households; however, one participant did not complete the survey; <sup>b</sup> Recall of previous water quality results was not a formal question but was noted as awareness similar to receiving a do not consume notice.

### 3.5.2 Water Quality and Hazard Identification

Tap water samples were collected from RM184 ( $n = 76$ ) and BOFN ( $n = 45$ ). Fourteen percent sample duplicates had almost perfect concordance for arsenic (0.98; 95%CI 0.97 to 0.99). Descriptive statistics for arsenic concentration at the tap for RM184 and BOFN are provided in Table 3.4. Arsenic distributions were right-skewed and fit a lognormal distribution.

Table 3.4 Descriptive statistics for community arsenic concentrations ( $\mu\text{g/L}$ ) in household tap water.

Community	Mean	Standard Deviation	Median	Min	Max
RM184 ( $n = 76$ )	2.58	5.38	0.65	0.12 <sup>a</sup>	36.3
BOFN ( $n = 45$ )	1.57	1.79	0.70	0.12 <sup>a</sup>	8.30

<sup>a</sup> one-half of detection limit

In RM184, at least one health related objective was exceeded in 67% (51/76) of the households' tap water including arsenic, boron, total coliform bacteria, *E.coli*, nitrate, lead, selenium, and uranium. Tap water in BOFN households exceeded at least one health related objective for total coliform bacteria, *E.coli*, and nitrate in 56% (25/45) households. The percentage of households with tap water exceeding at least one aesthetic objective was 91% (69/76) in RM184 and 58% (26/45) in BOFN. The aesthetic water quality excursions were frequent in RM184 and included chloride, iron, hardness, magnesium, manganese, sodium, pH, sulphate, total alkalinity, and total dissolved solids. Aesthetic objectives exceeded on BOFN included copper, iron, manganese, sodium, total alkalinity, and total dissolved solids. The number and percent of tap water quality excursion(s) from the *Saskatchewan's Drinking Water Quality Standards and Objectives* (2017) are provided in Supplementary Materials – 5.6 Community Tap Water Excursions.

### 3.5.3 Characterization of Risk Perception and Probability of Using Tap Water for Drinking

The probability of drinking tap water in RM184 when water was perceived as safe was 0.92 and not safe was 0.00 (Table 3.5). Tap water used for drinking when perceived unsafe in RM184 was verbally expressed as infrequent and circumstantial (e.g. drinking from the hydrant while working in the yard). The probability of drinking tap water in BOFN when water was perceived as safe was 0.77, and 0.11 when perceived as unsafe (Table 3.5). There was no significant relationship between drinking water risk perception and at least one health exceedance (water safety) in RM184 ( $p = 0.95$ ) or BOFN ( $p = 0.89$ ). A significant relationship between drinking water risk perception and exceedance of an aesthetic objective for BOFN ( $p = 0.04$ ) but not RM184 ( $p = 0.22$ ) existed. Households from BOFN were 1.7 times more likely (95%CI 1.0 to 2.7) to have water exceeding an aesthetic objective if the water was reported as unsafe than if reported as safe.

Table 3.5 The probability of community water use for drinking given household perception it was safe or not safe. The probability the water is not used for drinking given it was perceived as safe or not safe is included for completeness but was not used in the characterization of risk perception.

Community	Conditions	Total Number	Prevalence Estimate	Exact-Binomial (Clopper-Pearson) 95% Confidence Interval
RM184	P (not safe used)	0	0.00	0.00, 0.13
Drinking water (n = 76)	P (not safe not used)	26	1.00	0.87, 1.00
	P (safe used)	46	0.92	0.81, 0.98
	P (safe not used)	4	0.08	0.02, 0.19
BOFN	P (not safe used)	2	0.11	0.01, 0.35
Drinking Water (n = 44) <sup>a</sup>	P (not safe not used)	16	0.89	0.65, 0.99
	P (safe used)	20	0.77	0.56, 0.91
	P (safe not used)	6	0.23	0.09, 0.44

<sup>a</sup>one survey was not completed

### 3.5.4 Risk Characterization

The integration of risk perception modified exposure to drinking water risk due to arsenic for both communities when compared to the Canadian national average drinking water rate (Richardson 1997). The integration of risk perception reduced the mean consumption rate by 0.2 L/day for RM184 (1.3 L/day, 95% credible interval 0.44 to 3.16) and BOFN (1.3 L/day, 95% credible interval 0.43 to 3.06). The adjusted drinking water rates decrease the overall ILCR and probability of community risk for RM184 and BOFN; however, the 95% credible intervals overlap and the differences were not significant (Table 3.6).

Table 3.6 Incremental lifetime cancer risk and probability of risk exceeding Health Canada's negligible cancer risk of 1:100,000 for adults consuming tap water from communities of RM184 and BOFN.

Community	Inclusion of Drinking Water Risk Perception	Incremental Lifetime Cancer Risk (Mean) <sup>a</sup>	95% Credible Interval <sup>†</sup>	Probability Community Risk Exceeds Negligible
RM184	No	<b>1.5 x 10<sup>-5</sup></b>	9.5 x 10 <sup>-8</sup> , <b>1.0 x 10<sup>-4</sup></b>	0.26
	Yes	<b>1.3 x 10<sup>-5</sup></b>	8.4 x 10 <sup>-8</sup> , <b>9.0 x 10<sup>-5</sup></b>	0.23
BOFN	No	1.0 x 10 <sup>-5</sup>	2.5 x 10 <sup>-7</sup> , <b>5.1 x 10<sup>-5</sup></b>	0.25
	Yes	8.9 x 10 <sup>-6</sup>	2.2 x 10 <sup>-7</sup> , <b>5.9 x 10<sup>-5</sup></b>	0.22

<sup>a</sup>ILCR over Health Canada's negligible cancer risk in bold

RM184 had a mean ILCR exceeding Health Canada's negligible risk with or without risk perception. RM184 had a 23% chance of being over the acceptable cancer risk if risk perception was included in the model and 26% chance if perception was not included. Although BOFN did not have a mean ILCR greater than the negligible risk, the uncertainty, represented by the upper credible interval, indicated the mean ILCR for the community could be greater than 1:100,000. BOFN had a 22% chance of being over the acceptable cancer risk if risk perception was included in the model and 25% chance if perception was not included. Fig. 3.3 visually indicates the model's sensitivity to the inclusion of risk perception displayed using a cumulative distribution of the ILCR for both communities. The inclusion of risk perception decreased the percent area of the ILCR probability density function by 9% for BOFN and 19% for RM184.

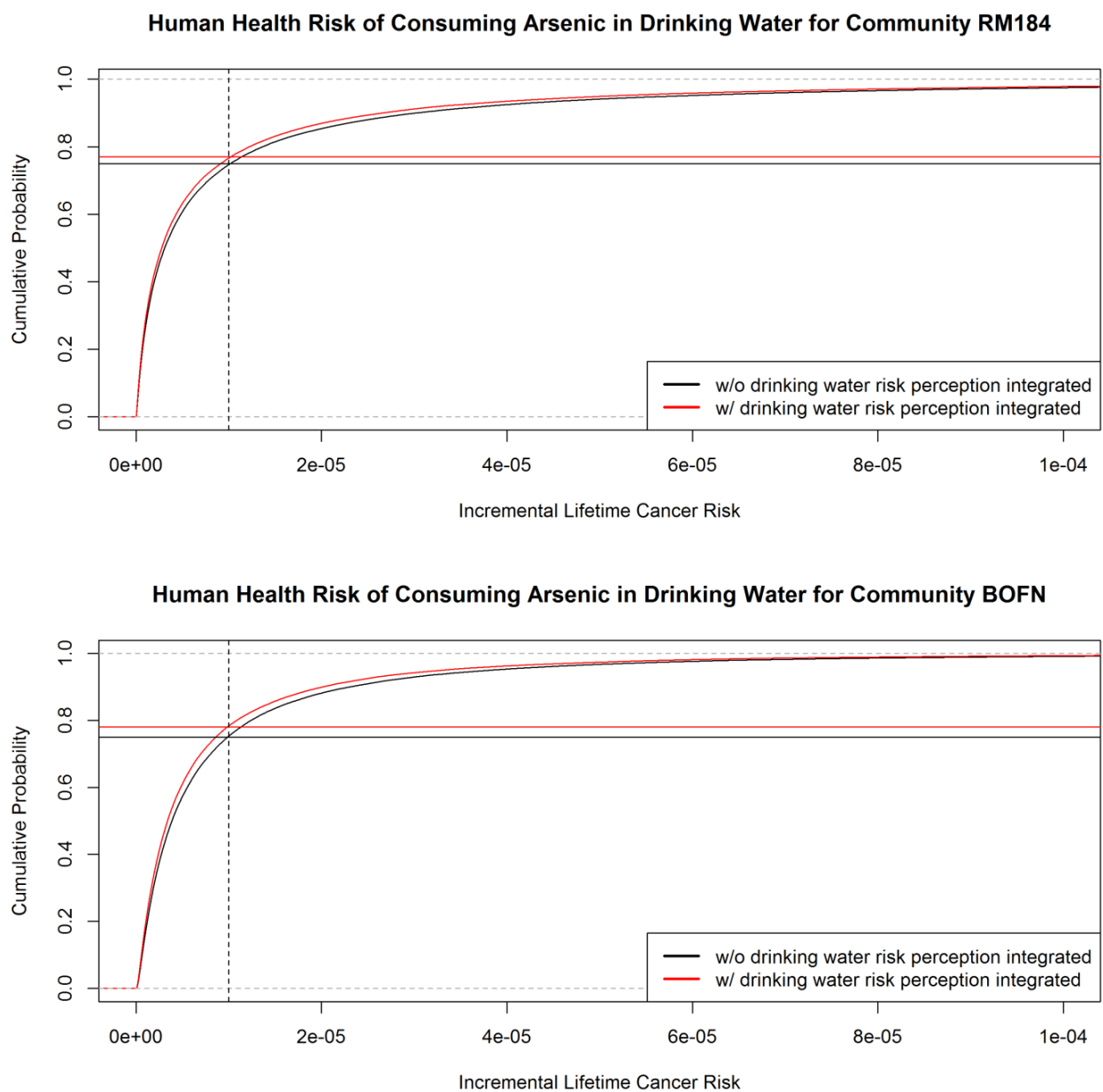


Figure 3.3 Cumulative probability distribution indicating the probability that the incremental lifetime cancer risk (ILCR) from arsenic was exceeded in drinking water for the Rural Municipality of Grayson #184 (RM184) and Beardy's and Okemasis First Nation (BOFN). Vertical dashed line denotes Health Canada's negligible cancer risk of 1:100,000. Horizontal lines represent the difference in the cumulative probability without perception (w/o drinking water risk perception) and with perception (w/ drinking water risk perception) integrated.

### 3.6 Discussion

#### 3.6.1 Holistic Human Health Risk Assessment

This study provides an example of how the integration of a non-traditional factor, drinking water risk perception, can improve the interpretation of risk by quantifying the discrepancy and uncertainty associated with perception and use. Dichotomous data on whether or not a household perceived their water was safe and whether or not they chose to drink it were summarized by community as beta probability density functions. These methods can facilitate holistic human health risk assessment through inclusion of non-traditional factors to better inform risk, communication and management. The probabilistic Bayesian model developed for this study can provide valuable priors for future holistic HHRAEs especially as it relates to communities with limited historical data.

Wilks et al. (2015) acknowledged the need to include behaviours, socio-economics, perceptions, and values to improve risk characterization and management through a holistic approach to benefit communities and improve drinking water management. Bridges (2003), Liu et al. (2012), Serre et al. (2003), and Zargar et al. (2014) explored the use of probabilistic methods to facilitate the integration of different data types and sources to improve risk assessment and quantify uncertainty. This study integrates risk perception but does not consider the other benefits associated with valuing perceptions in calculations of risk, and future decision-making. Future research is required to assess how the integration of perception may influence:

- 1) community ownership and participation in the research,
- 2) trust in collaborators (e.g. institutions or researchers), and
- 3) the approach and effectiveness of risk communication and management.

Integration of perception may provide meaningful context to empirical results, bridge and build trusting partnerships.

The integration of community risk perception did not significantly decrease the estimated mean exposure and overall risk to communities but it provides insight into how perception and consumption are related. Though drinking water rates would not be expected to vary significantly from the Canadian average, these results suggest that assuming Health Canada's deterministic drinking water rate can over-estimate community exposure and human health risk (Liu et al. 2012). Similarly, the mean ILCR decreased but did not change overall average risk in either community. Regardless, the probabilistic Bayesian risk assessment method provides a posterior probability distribution and credible interval that characterizes the potential range of the mean ILCR, better informing risk interpretation in comparison to deterministic methods. Quantification



of uncertainty, represented by the credible interval around the mean, is the non-additive accumulation of the uncertainty associated with the variables in the model (Liu et al. 2012). The probability that the mean ILCR exceeds a negligible risk requires only a simple step-function on the posterior probability distribution. The visual representation of the exposure and risk is easy to understand and provides context to the measure of risk to facilitate risk communication.

### 3.6.2 Risk Perception and Exposure

Similar to findings by Doria (2010) and Owen et al. (1999), there was no association between the presence of a drinking water health hazard and perceived risk in either BOFN or RM184. The lack of association between knowledge of previous ‘do not consume’ notices and current safety of their drinking water suggests households develop their perception of risk in the absence of quantitative information. Doria (2010), Orgill et al. (2013), Syme and Williams (1993), and Turgeon et al. (2004) report similar findings. This supports the need for community-based approaches to risk assessment as perception of drinking water risk is based on unique experiences with water sources (Hynds et al. 2013; Maxwell et al. 1998).

The probability of drinking water use as it relates to risk perception can summarize the effect of multiple factors influencing perception on exposure and risk. Discrepancies between the perception and the actual hazards can be used to improve risk communication and management specific to the community. Doria (2010) identifies multiple factors that influence perception of drinking water such as organoleptic properties, trust, control, experience, and vulnerability which manifest differently among individuals and communities. Perception of drinking water may also be influenced by non-traditional factors such as culture (Doria 2010; Spence and Walters 2012). Determining how each factor influences risk perception in each community may not be realistic, in practice, for communities and governments, however, the effect of risk perception on the probability of drinking water use can identify the need for additional research or enquiry. For example, the majority of households in the RM184 consumed their water when they felt it was safe but also had in-house water treatment which can decrease aesthetic water problems and provide households with a sense of control, assurance, and confidence that the water is safe to drink. Though the majority of households on BOFN also drank their water if it was perceived as safe, they had an 11% probability of consuming water perceived as not safe. Many households on BOFN reported use of bottled water, so the continued use of tap water suggests there are other barriers, in addition to perception, preventing access to safe drinking water.

First Nations are not identical and the barriers associated with colonization continue to impact their governance, economy, society, and perceptions which can manifest in a lack of trust in institutions, a lack of control over their land and resources, and dissatisfaction with the quality and equality associated with their water supplies (Dupont et al. 2014; Spence and Walters 2012). As a result, future research is required to determine the specific reasons unsafe water is consumed on BOFN. Research results were shared and discussed with the health director of BOFN to

attempt to correct risk perceptions in the context of the specific cultural and spiritual needs of their community.

### 3.6.3 Unregulated Drinking Water Risks

Over half of study households were exposed to drinking water hazards and potential health risks associated with unregulated drinking water sources similar to global rural populations (Chakraborti et al. 2015; Fox et al. 2016; Gleick 2002; Schwarzenbach et al. 2010; Villanueva et al. 2013). Unexpectedly, only 7% of households on BOFN were aware of previous metal testing despite years of routine testing conducted by the Band and Health Canada (2013). Although well water quality could have changed since households last tested, the history of water quality testing and communication of results in both communities was insufficient to properly inform residents of potential drinking water associated health risks. These findings are similar to Ochoo et al. (2017) who found a discrepancy between the perception of water quality and the actual water quality of public systems despite government website information improvement efforts.

RM184 households treated their private well water despite not having an accurate perception of their water quality. Though guideline exceedances in tap water did not inform household decisions to install water treatment, exceedances in the aesthetic water quality of the raw water may have been a motivating factor (Doria 2010). The lack of association between perception and the aesthetic exceedances of the tap water in RM184 may be a result of improved perception of the frequency of post treatment tap water relative to raw well water. Fewer BOFN households had water treatment and exceedances in aesthetic objectives were significantly associated with their perception of drinking water safety. Neither tap water exceedances in health or aesthetic guidelines nor education level were significant determinants of the presence of water treatment in either of the communities; however, a household annual income > \$50,000 was significantly associated with water treatment use in both. Studies conflict on the existence of a relationship between income or education level and water treatment (Doria 2010; Jones et al. 2007), indicating that these associations may be community or household specific.

The lack of association between risk perception and the presence of a drinking water health exceedance, and education level and water treatment; and, the influence of community and income on the use of water treatment indicates a need for: 1) monetary incentives or supports for household monitoring or treatment, and 2) accurate drinking water quality and water treatment information specific to households and communities. For example, despite the lack of knowledge around drinking water quality, households in RM184 with higher income were able to afford water treatment equipment. Lower income households were less able to afford water treatment equipment, regardless of their perception, which may be a confounding issue on BOFN where households also lack well ownership and control. At minimum, awareness of drinking water quality is the first step to ensuring appropriateness of water treatment and potential application of

assistance programs to provide guidance to low income households or communities accessing high-risk drinking water sources.

### 3.6.4 Risk Communication and Management

Quantification and awareness of uncertainty, or the potential range of risk, is beneficial information to consumers which can assist them in making safe, sustainable drinking water choices. Markon and Lemyre (2013) studied public reactions to the communication of risk and uncertainty and suggest that authorities be cautious and precise in their approach to communicating uncertainty. Providing complete information on risk and uncertainty to the public will decrease the chance that information is misinterpreted, especially when a behavioural change is required by the public to decrease risk (Markon and Lemyre 2013). Many consumers are capable of making decisions, maintaining trust in authorities, and understanding the meaning of the message when uncertainty is disclosed (Markon and Lemyre 2013). When the public feel they are not well informed, or there is disagreement in messages from authorities, they are less likely to follow risk management instruction and will trust their own experiences. Thus, in the absence of drinking water regulations, which provide the framework to monitor, interpret and mitigate drinking water risks, knowledge of consumer risk perception provides insight on drinking water use (exposure). Quantitative water quality data and a summary of community drinking water risk perception can be used to correct misconceptions to better support household and community drinking water risk management. This information can reduce the use of unsafe drinking water sources and encourage the use of safe drinking water sources for more sustainable water resource management and protection of human health.

A risk communication and management strategy specific to each community in the study can be developed by local health authorities, government, communities, or households based on risk perception and exposure, hazards present, and overall health risk. For example, community members and leaders on BOFN may be aware of the barriers causing households on BOFN to consume tap water despite perceiving it as unsafe to drink. Understanding the community's risk perception is an important subjective component that compliments the objective measure of risk required to create an effective risk communication strategy for First Nations communities (Spence and Walters 2012). Spence and Walters (2012) suggest a community-based partnership of researchers and policy personal work together to develop risk communication and management. This may be an important consideration for future researchers when working with First Nation communities. Risk communication and management in RM184 would differ given the high level of ownership and responsibility households have over their water source. Furthermore, they have a high percentage of household water treatment use and a low probability of consuming water they believe is unsafe. With a greater ability to invest in water treatment equipment, households in RM184 may only require support in the identification of drinking water hazards and guidance on how to establish effective treatment. Incentive programs for testing or

information on treatment equipment may further motivate households in RM184 to decrease exposure to drinking water hazards.

Application of similar strategies for risk communication and management will not work for both communities. Provision of only water quality results to BOFN households offers awareness of risk but does not allow action because households lack control over the water source which may result in a lack of trust and acceptability of the water (Syme and Williams 1993). Solutions would have to be community-based given the unique community culture and governance on BOFN which requires careful consideration to ensure management and communication is relevant and effective (Doria 2010). Not limited to First Nations that are culturally unique, Huerta and Macario (1999) conducted a case study in which they effectively communicated cancer risk to Hispanic-Americans in the United States by tailoring their communication within the community's cultural context. They concluded it is essential to understand the community culture and sub-culture in order to effectively communicate risk, and change community perceptions and behaviours.

This approach to holistic human health risk assessment provides a deeper understanding of behaviour associated with community perception of drinking water risk, the probability of consumption, and potential for human health risk associated with hazards. Morris, Wilson, and Kelly (2016) provide a model to conduct effective outreach to private well owners through the identification of the audience (e.g. community), identification of barriers, motivator selection, partnership building, and trust. Future research can explore the benefits of extending partnerships between researchers and community decision-makers or policy makers as suggested by Spence and Walters (2012), and the application of the outreach model developed by Morris et al. (2016) to a broader audience including First Nations.

### 3.6.5 Limitations and Uncertainty

This research is subject to selection bias given the census approach to data collection; however, communities were small and participation rates were high. In future a randomized sampling approach to characterize perception could be used, especially where drinking water hazards are known. Information bias associated with misclassification of risk perception may be present despite the use of a probability density function to account for uncertainty associated with household responses to the categorical questions on drinking water safety. Community-specific risk perception data may not transfer to another community; however, this model offers informed priors that can support weaker datasets from subsequent communities or data updates within the study communities as suggested by Hooten and Hobbs (2015). Uncertainty was thoroughly assessed by applying the US EPA guidance on the process for conducting probabilistic risk assessment (2001). Parameter uncertainty includes the potential for missing data associated with the arsenic concentrations and survey data due to non-respondent households meeting study inclusion criteria. Systematic error in data collection was minimized through use of standard

methods, a piloted survey, duplication of samples, and accredited laboratories. Model uncertainty was minimized by using Health Canada's standard equations for HHRA (HC 2010b) and values. Non-informative priors and posterior distributions were individually checked to ensure posterior distributions for local parameters represented the data. Life expectancy values may be biased due to assignment of a normally distributed probability density function when it is known to be a non-symmetric distribution with highly variable shapes dependent on population (Román et al. 2007); however, Canadian specific data could not be obtained from Health or Statistics Canada. A conservative exposure frequency between 340 to 350 days per year was selected to reduce bias of over-estimation of risk if households consumed alternative water. Model validation ensured average model outputs matched deterministic values for each parameter (not provided). Sensitivity was determined by calculating the percent overlap between probability density functions with and without perception for each community. This study was limited to oral exposure in adults aged 20 to 59 and excluded additional drinking water hazards representing scenario uncertainty associated with exposure to mixtures of multiple carcinogen and non-carcinogens. Uncertainty associated with changes in arsenic concentrations, risk perception, and water consumption over time or seasonally was not accounted for given the study design (i.e. one survey/water test per household).

### 3.7 Conclusions

Probabilistic Bayesian methods can be used to integrate non-traditional data influencing risk and improved interpretation of risk for community specific risk communication and management. This model allows for the probability that health risk would be over a threshold, the negligible cancer risk, provides informative priors, and could be adapted to include multiple hazards, additional age groups, or used to direct support and funding by prioritizing communities at risk. Community risk perceptions are unique and influence probability of drinking water use. This research agrees with previous studies that found participants were unable to accurately perceive the safety of their drinking water. The influence of community and income on use of water treatment indicates need for:

- 1) monetary incentives or supports for household monitoring or treatment, and
- 2) accurate and community specific drinking water quality and water treatment information.

A discrepancy between risk perception and drinking water use may identify a need for additional research or investigation into barriers to safe drinking water. Maintaining partnerships with community leaders and policy makers to develop risk communication and management strategies could be beneficial. Future research should assess how inclusion of risk perception aids development of effective risk communication and management. A culturally appropriate framework enabling First Nations to conduct effective risk outreach, similar to that for private well owners (Morris et al. 2016) should be developed.

### 3.8 References

- Ames, Daniel P., Bethany T. Neilson, David K. Stevens, and Upmanu Lall. 2005. "Using Bayesian Networks to Model Watershed Management Decisions: An East Canyon Creek Case Study." *Journal of Hydroinformatics* 7 (4): 267–82.  
<https://doi.org/10.2166/hydro.2005.0023>.
- Bridges, Jim. 2003. "Human Health and Environmental Risk Assessment: The Need for a More Harmonised and Integrated Approach." *Chemosphere* 52 (9): 1347–51.  
[https://doi.org/10.1016/S0045-6535\(03\)00469-7](https://doi.org/10.1016/S0045-6535(03)00469-7).
- Chakraborti, Dipankar, Mohammad Mahmudur Rahman, Amitava Mukherjee, Mohammad Alauddin, Manzurul Hassan, Rathindra Nath Dutta, Shymapada Pati, et al. 2015. "Groundwater Arsenic Contamination in Bangladesh-21 Years of Research." *Journal of Trace Elements in Medicine and Biology* 31: 237–48.  
<https://doi.org/10.1016/j.jtemb.2015.01.003>.
- Charrois, J. W. A. 2010. "Private Drinking Water Supplies: Challenges for Public Health." *Canadian Medical Association Journal* 182 (10): 1061–64.  
<https://doi.org/10.1503/cmaj.090956>.
- Chen, Hanyi, Yaying Zhang, Linlin Ma, Fangmin Liu, Weiwei Zheng, Qinfeng Shen, Hongmei Zhang, et al. 2012. "Change of Water Consumption and Its Potential Influential Factors in Shanghai: A Cross-Sectional Study." *BMC Public Health* 12 (1): 450.  
<https://doi.org/10.1186/1471-2458-12-450>.
- Chowdhury, Shakhawat, Pascale Champagne, and P. James McLellan. 2009. "Uncertainty Characterization Approaches for Risk Assessment of DBPs in Drinking Water: A Review." *Journal of Environmental Management* 90 (5): 1680–91.  
<https://doi.org/10.1016/j.jenvman.2008.12.014>.
- Corkal, Darrell, W. C. Schutzman, and Clint R. Hilliard. 2004. "Rural Water Safety From the On-Farm Tap." *Journal of Toxicology and Environmental Health, Part A* 67 (20–22): 1619–42. <https://doi.org/10.1080/15287390490491918>.
- Delignette-Muller, Marie Laure, Christophe Dutang, Regis Pouillot, and Denis Jean-Baptiste. 2016. "Package 'Fitdistrplus', Version 1.0-7."
- Doria, Miguel de França. 2010. "Factors Influencing Public Perception of Drinking Water Quality." *Water Policy* 12 (1): 1–19. <https://doi.org/10.2166/wp.2009.051>.
- Dupont, Diane, W. L. (Vic) Adamowicz, and Alan Krupnick. 2010. "Differences in Water Consumption Choices in Canada: The Role of Socio-Demographics, Experiences, and Perceptions of Health Risks." *Journal of Water and Health* 8 (4): 671–86.  
<https://doi.org/10.2166/wh.2010.143>.
- Dupont, Diane, Cheryl Waldner, Lalita Bharadwaj, Ryan Plummer, Blair Carter, Kate Cave, and Rebecca Zagozewski. 2014. "Drinking Water Management: Health Risk Perceptions and Choices in First Nations and Non-First Nations Communities in Canada." *International*

*Journal of Environmental Research and Public Health* 11 (6): 5889–5903.  
<https://doi.org/10.3390/ijerph110605889>.

- Ford, Lorelei, Lalita Bharadwaj, Lianne McLeod, and Cheryl Waldner. 2017. “Human Health Risk Assessment Applied to Rural Populations Dependent on Unregulated Drinking Water Sources: A Scoping Review.” *International Journal of Environmental Research and Public Health* 14 (8): 846. <https://doi.org/10.3390/ijerph14080846>.
- Fox, Mary A., Kieve E. Nachman, Breeana Anderson, Juleen Lam, and Beth Resnick. 2016. “Meeting the Public Health Challenge of Protecting Private Wells: Proceedings and Recommendations from an Expert Panel Workshop.” *Science of the Total Environment* 554–555: 113–18. <https://doi.org/10.1016/j.scitotenv.2016.02.128>.
- Gleick, Peter H. 2002. “Dirty Water : Estimated Deaths from Water-Related Diseases 2000–2020.” *Pacific Institute for Studies in Development, Environment, and Security*.
- Health Canada (HC). 2006. “Guidelines for Canadian Drinking Water Quality: Guideline Technical Document - Arsenic.” *Water Quality and Health Bureau, Healthy Environments and Consumer Safety Branch*. Ottawa, Ont. <https://doi.org/10.1016/j.scitotenv.2009.04.006>.
- . 2010a. “Federal Contaminated Site Risk Assessment in Canada, Part 11: Health Canada Toxicological Reference Values (TRVs) and Chemical-Specific Factors, Version 2.0.” Ottawa, Ont.
- . 2010b. “Guidance on Human Health Preliminary Quantitative Risk Assessment (PQRA), Version 2.0. Part I.” Ottawa, Ont.
- . 2010c. “Guidance on Peer Review of Human Health Risk Assessments for Federal Contaminates Sites in Canada, Version 2.0.” *Health Canada*. Ottawa, Ont.
- . 2013. “Ensuring Safe Drinking Water in First Nations Communities in Canada.” First Nations & Inuit Health. 2013. <http://www.hc-sc.gc.ca/fniah-spnia/promotion/public-publique/sfw-sep-eng.php>.
- Hooten, M. B., and N. T. Hobbs. 2015. “A Guide to Bayesian Model Selection for Ecologists.” *Ecological Monographs* 85 (1): 3–28. <https://doi.org/10.1890/14-0661.1>.
- Huerta, Elmer. E, and Everly Macario. 1999. “Communicating Health Risk to Ethnic Groups: Reaching Hispanics as a Case Study.” *JNCI Monographs* 1999 (25): 23–26. <https://doi.org/10.1093/oxfordjournals.jncimonographs.a024202>.
- Hynds, Paul D., Bruce D. Misstear, and Laurence W. Gill. 2013. “Unregulated Private Wells in the Republic of Ireland: Consumer Awareness, Source Susceptibility and Protective Actions.” *Journal of Environmental Management* 127: 278–88. <https://doi.org/10.1016/j.jenvman.2013.05.025>.
- Jackson, Sue. 2006. “Compartmentalising Culture: The Articulation and Consideration of Indigenous Values in Water Resource Management.” *Australian Geographer* 37 (1): 19–31. <https://doi.org/10.1080/00049180500511947>.

- Jackson, Sue, Poh Ling Tan, Carla Mooney, Suzanne Hoverman, and Ian White. 2012. "Principles and Guidelines for Good Practice in Indigenous Engagement in Water Planning." *Journal of Hydrology* 474: 57–65. <https://doi.org/10.1016/j.jhydrol.2011.12.015>.
- Jones, A. Q., S. E. Majowicz, V. L. Edge, M. K. Thomas, L. MacDougall, M. Fyfe, S. Atashband, and S. J. Kovacs. 2007. "Drinking Water Consumption Patterns in British Columbia: An Investigation of Associations with Demographic Factors and Acute Gastrointestinal Illness." *The Science of the Total Environment* 388 (1–3): 54–65. <https://doi.org/10.1016/j.scitotenv.2007.08.028>.
- Jones, Andria Q., Catherine E. Dewey, Kathryn Doré, Shannon E. Majowicz, Scott A. McEwen, David Waltner-Toews, Spencer J. Henson, and Eric Mathews. 2005. "Public Perception of Drinking Water from Private Water Supplies: Focus Group Analyses." *BMC Public Health* 5: 129. <https://doi.org/10.1186/1471-2458-5-129>.
- Lin, L.I. 2000. "A Note on the Concordance Correlation Coefficient (Corrections)." *Biometrics* 56 (1): 324–25. <https://doi.org/10.1111/j.0006-341X.2000.00324.x>.
- Liu, Kevin Fong-Rey, Che-Fan Lu, Cheng-Wu Chen, and Yung-Shuen Shen. 2012. "Applying Bayesian Belief Networks to Health Risk Assessment." *Stochastic Environmental Research and Risk Assessment* 26 (3): 451–65. <https://doi.org/10.1007/s00477-011-0470-z>.
- Lunn, David, Christopher Jackson, Nicky Best, Andrew Thomas, and David Spiegelhalter. 2013. *The BUGS Book: A Practical Introduction to Bayesian Analysis*. CRC Press, Taylor & Francis Group. Boca Raton, FL.
- Lunn, David, David Spiegelhalter, Andrew Thomas, and Nicky Best. 2009. "The BUGS Project: Evolution, Critique and Future Directions." *Statistics in Medicine* 28 (25): 3049–67. <https://doi.org/10.1002/sim.3680>.
- Maathuis, Harm. 2000. "Review and Comparison of Regional Groundwater Quality Data in Saskatchewan." Review and Comparison of Regional Groundwater Quality Data in Saskatchewan. Saskatoon, SK.
- . 2008a. "Groundwater Quality Databases in Saskatchewan." *Saskatchewan Watershed Authority*. Saskatoon, SK.
- . 2008b. "Quality of Natural Groundwaters in Saskatchewan." Saskatoon, SK.
- Markon, Marie-Pierre L., and Louise Lemyre. 2013. "Public Reactions to Risk Messages Communicating Different Sources of Uncertainty: An Experimental Test." *Human and Ecological Risk Assessment: An International Journal* 19 (4): 1102–26. <https://doi.org/10.1080/10807039.2012.702015>.
- Maxwell, Reed M., Susan D. Pelmulder, Andrew F. B. Tompson, and William E. Kastenberg. 1998. "On the Development of a New Methodology for Groundwater-Driven Health Risk Assessment." *Water Resources Research* 34 (4): 833–47.
- Morris, Lucinda, Steve Wilson, and Walton Kelly. 2016. "Methods of Conducting Effective Outreach to Private Well Owners - A Literature Review and Model Approach." *Journal of*



- Water and Health* 14 (2): 167–82. <https://doi.org/10.2166/wh.2015.081>.
- Nsiah-Kumi, Phyllis A. 2008. “Communicating Effectively with Vulnerable Populations during Water Contamination Events.” *Journal of Water and Health* 6 (S1): 63–75. <https://doi.org/10.2166/wh.2008.041>.
- Ochoo, Benjamin, James Valcour, and Atanu Sarkar. 2017. “Association between Perceptions of Public Drinking Water Quality and Actual Drinking Water Quality: A Community-Based Exploratory Study in Newfoundland (Canada).” *Environmental Research* 159: 435–43. <https://doi.org/10.1016/j.envres.2017.08.019>.
- Onufrak, Stephen J., Sohyun Park, Joseph R. Sharkey, Caitlin Merlo, Wesley R. Dean, and Bettylou Sherry. 2014. “Perceptions of Tap Water and School Water Fountains and Association With Intake of Plain Water and Sugar-Sweetened Beverages.” *Journal of School Health* 84 (3): 195–204. <https://doi.org/10.1111/josh.12138>.
- Onufrak, Stephen, Sohyun Park, Joseph R. Sharkey, and Bettylou Sherry. 2012. “The Relationship of Perceptions of Tap Water Safety with Intake of Sugar-Sweetened Beverages and Plain Water among US Adults.” *Public Health Nutrition* 17 (1): 1–7. <https://doi.org/10.1017/S1368980012004600>.
- Orgill, Jennifer, Ameer Shaheed, Joe Brown, and Marc Jeuland. 2013. “Water Quality Perceptions and Willingness to Pay for Clean Water in Peri-Urban Cambodian Communities.” *Journal of Water and Health* 11 (3): 489–506. <https://doi.org/10.2166/wh.2013.212>.
- Owen, A. J., J. S. Colbourne, C. R. I. Clayton, and C. Fife-Schaw. 1999. “A Mental Model’s Approach to Customer Perception of Drinking-Water Supply and Quality.” *Water and Environment Journal* 13 (4): 241–44. <https://doi.org/10.1111/j.1747-6593.1999.tb01041.x>.
- Pastore, Massimiliano. 2017. “Estimation of Overlapping in Empirical Distributions, R Package Version 1.5.0.”
- Patrick, Robert J. 2011. “Uneven Access to Safe Drinking Water for First Nations in Canada: Connecting Health and Place through Source Water Protection.” *Health & Place* 17 (1): 386–89. <https://doi.org/10.1016/j.healthplace.2010.10.005>.
- Plummer, Martyn, Nicky Best, Kate Cowles, Karen Vines, Deepayan Sarkar, Douglas Bates, Russell Almond, and Arni Magnusson. 2016. “Package ‘Coda’: Output Analysis and Diagnostics for MCMC.”
- Pons, Wendy, Ian Young, Jenifer Truong, Andria Jones-Bitton, Scott McEwen, Katarina Pintar, and Andrew Papadopoulos. 2015. “A Systematic Review of Waterborne Disease Outbreaks Associated with Small Non-Community Drinking Water Systems in Canada and the United States.” Edited by Paula V Morais. *PLOS ONE* 10 (10): 1–17. <https://doi.org/10.1371/journal.pone.0141646>.
- Rice, E.W., R.B. Baird, A.D. Eaton, and L.S. Clesceri. 2012. *Standard Methods for the Examination of Water and Wastewater*. Edited by E.W. Rice, R.B. Baird, A.D. Eaton, and L.S. Clesceri. American Public Health Association, American Water Works Association,

Water Environment Federation.

- Richardson, G. Mark. 1997. "Compendium of Canadian Human Exposure Factors for Risk Assessment." Ottawa, Ont.
- Ritter, Len, Keith Solomon, Paul Sibley, Ken Hall, Patricia Keen, Gevan Mattu, and Beth Linton. 2002. "Sources, Pathways, and Relative Risks of Contaminants in Surface Water and Groundwater: A Perspective Prepared for the Walkerton Inquiry." *Journal of Toxicology and Environmental Health, Part A* 65 (1): 1–142. <https://doi.org/10.1080/152873902753338572>.
- Román, Rubén, Mercé Comas, Lorena Hoffmeister, and Xavier Castells. 2007. "Determining the Lifetime Density Function Using a Continuous Approach." *Journal of Epidemiology & Community Health* 61: 923–25. <https://doi.org/10.1136/jech.2006.052639>.
- RStudio Team. 2016. "RStudio: Integrated Development Environment for R." Boston, MA: RStudio, Inc.
- Ryan, Louise. 2003. "Epidemiologically Based Environmental Risk Assessment." *Statistical Science* 18 (4): 466–80. <https://doi.org/10.1214/ss/1081443230>.
- Saskatchewan Health. 2014. "Detailed Regional Health Authority Population Totals." Covered Population Report. 2014. <https://opendata.ehealthsask.ca/MicroStrategyPublic/asp/Main.aspx>.
- Schwarzenbach, René P., Thomas Egli, Thomas B. Hofstetter, Urs von Gunten, and Bernhard Wehrli. 2010. "Global Water Pollution and Human Health." *Annual Review of Environment and Resources* 35: 109–36. <https://doi.org/10.1146/annurev-environ-100809-125342>.
- Serre, M. L., A. Kolovos, G. Christakos, and K. Modis. 2003. "An Application of the Holistochastic Human Exposure Methodology to Naturally Occurring Arsenic in Bangladesh Drinking Water." *Risk Analysis* 23 (3): 515–28. <https://doi.org/10.1111/1539-6924.t01-1-00332>.
- Shaheed, Ameer, Jennifer Orgill, Maggie A. Montgomery, Marc A. Jeuland, and Joe Brown. 2014. "Why 'Improved' Water Sources Are Not Always Safe." *Bulletin of the World Health Organization* 92 (4): 283–89. <https://doi.org/10.2471/BLT.13.119594>.
- Simpson, Hugh. 2010. "Promoting the Management and Protection of Private Water Wells." *Journal of Toxicology and Environmental Health, Part A* 67 (20–22): 1679–1704. <https://doi.org/10.1080/15287390490492296>.
- Slovic, Paul. 1987. "Perception of Risk." *Science* 236 (4799): 280–85. <https://doi.org/10.1126/science.3563507>.
- Spence, Nicholas, and Dan Walters. 2012. "'Is It Safe?' Risk Perception and Drinking Water in a Vulnerable Population." *International Indigenous Policy Journal* 3 (3): 1–23. <https://doi.org/10.18584/iipj.2012.3.3.9>.
- Spiegelhalter, David, Andrew Thomas, Nicky Best, and Dave Lunn. 2012. "OpenBUGS User

- Manual, Version 3.2.2.” Open Source.
- Statistics Canada. 1996. “Groundwater: Groundwater Use.” Quarterly Estimates of the Population of Canada, the Provinces and the Territories. 1996. <https://www.ec.gc.ca/eau-water/default.asp?lang=En&n=300688DC-1#sub5>.
- Suter II, Glenn W., Theo Vermeire, Wayne R. Munns Jr., and Jun Sekizawa. 2005. “An Integrated Framework for Health and Ecological Risk Assessment.” *Toxicology and Applied Pharmacology* 207 (2): 611–16. <https://doi.org/10.1016/j.taap.2005.01.051>.
- Syme, Geoffrey J., and K. D. Williams. 1993. “The Psychology of Drinking Water Quality: An Exploratory Study.” *Water Resources Research* 29 (12): 4003–10. <https://doi.org/10.1029/93WR01933>.
- Tighe, Matthew, Carmel A. Pollino, and Susan C. Wilson. 2013. “Bayesian Networks as a Screening Tool for Exposure Assessment.” *Journal of Environmental Management* 123: 68–76. <https://doi.org/10.1016/j.jenvman.2013.03.018>.
- Turgeon, Steve, Manuel J. Rodriguez, Marius Thériault, and Patrick Levallois. 2004. “Perception of Drinking Water in the Quebec City Region (Canada): The Influence of Water Quality and Consumer Location in the Distribution System.” *Journal of Environmental Management* 70 (4): 363–73. <https://doi.org/10.1016/j.jenvman.2003.12.014>.
- United States Environmental Protection Agency (US EPA). 1994. “Method 200.8: Determination of Trace Elements in Waters and Wastes by Inductively Coupled Plasma - Mass Spectrometry.” *U. S. Environmental Protection Agency*. Vol. 4.
- . 2001. “Process for Conducting Probabilistic Risk Assessment.” In *Risk Assessment Guidance for Superfund: Volume 111 - Part A*, 3rd ed., 3:1–34. Washington, DC: Office of Emergency and Remedial Response, U.S. Environmental Protection Agency.
- Villanueva, Cristina M., Manolis Kogevinas, Sylvaine Cordier, Michael R. Templeton, Roel Vermeulen, John R. Nuckols, Mark J. Nieuwenhuijsen, and Patrick Levallois. 2013. “Assessing Exposure and Health Consequences of Chemicals in Drinking Water: Current State of Knowledge and Research Needs.” *Environmental Health Perspectives* 122 (3): 213–21. <https://doi.org/10.1289/ehp.1206229>.
- Wang, Suiling, and Catherine N Mulligan. 2006. “Occurrence of Arsenic Contamination in Canada: Sources, Behavior and Distribution.” *The Science of the Total Environment* 366 (2–3): 701–21. <https://doi.org/10.1016/j.scitotenv.2005.09.005>.
- Water Security Agency (WSA). 2017. “Saskatchewan’s Drinking Water Quality Standards and Objectives ( Summarized ).” *Water Security Agency, Government of Saskatchewan*.
- Wescoat, James L., Lisa Headington, and Rebecca Theobald. 2007. “Water and Poverty in the United States.” *Geoforum* 38 (5): 801–14. <https://doi.org/10.1016/j.geoforum.2006.08.007>.
- Wilks, M. F., N. Roth, L. Aicher, M. Faust, P. Papadaki, A. Marchis, M. Calliera, et al. 2015. “White Paper on the Promotion of an Integrated Risk Assessment Concept in European Regulatory Frameworks for Chemicals.” *Science of The Total Environment* 521–522 (1):

- 211–18. <https://doi.org/10.1016/j.scitotenv.2015.03.065>.
- World Bank. 2015. “The World Bank: Data.” The World Bank: Data. 2015. <http://data.worldbank.org/indicator/SP.RUR.TOTL.ZS>.
- World Health Organization (WHO). 2011. “Guidelines for Drinking-Water Quality. 4th Ed.” 2011. [http://www.who.int/water\\_sanitation\\_health/publications/2011/dwq\\_guidelines/en/](http://www.who.int/water_sanitation_health/publications/2011/dwq_guidelines/en/).
- World Health Organization (WHO), and UNICEF. 2015. *25 Years Progress on Sanitation and Drinking Water: 2015 Update and MDG Assessment*. New York, NY: United Nation’s Children’s Fund and World Health Organisation. Geneva, Switzerland: New York, NY: United Nation’s Children’s Fund and World Health Organisation.
- Wright, Carlee J., Jan M. Sargeant, Victoria L. Edge, James D. Ford, Khosrow Farahbakhsh, Inez Shiwak, Charlie Flowers, Allan C. Gordon, and Sherilee L. Harper. 2018. “How Are Perceptions Associated with Water Consumption in Canadian Inuit? A Cross-Sectional Survey in Rigolet, Labrador.” *Science of the Total Environment* 618: 369–78. <https://doi.org/10.1016/j.scitotenv.2017.10.255>.
- Zargar, Amin, Roberta Dyck, M. Shafiqul Islam, Asish Mohapatra, and Rehan Sadiq. 2014. “Data Fusion Methods for Human Health Risk Assessment: Review and Application.” *Human and Ecological Risk Assessment: An International Journal* 20 (3): 807–38. <https://doi.org/10.1080/10807039.2012.746145>.
- Zheng, Yan, and Joseph D. Ayotte. 2015. “At the Crossroads: Hazard Assessment and Reduction of Health Risks from Arsenic in Private Well Waters of the Northeastern United States and Atlantic Canada.” *Science of the Total Environment* 505: 1237–47. <https://doi.org/10.1016/j.scitotenv.2014.10.089>.

## 4 Conclusion

### 4.1 Introduction

Almost half of the world's population is geographically rural with limited access to regulated water sources that provide safe drinking water through monitoring and treatment (WHO/UNICEF 2015). Initiated by the United Nations, the Millennium Development Goals (MDGs) were created to address this gap in access to safe drinking water and initiate increased access to improved sources increased global access by 15% (WHO/UNICEF 2015). However, access to improved water sources does not guarantee safe drinking water due to the lack of monitoring and reporting of water quality (Shaheed et al. 2014). For example, despite achieving 100% access to improved drinking water in North America there remains exposure to drinking water hazards associated with unregulated water including private wells (Charrois 2010; Spence and Walters 2012; Corkal, Schutzman, and Hilliard 2004; Fox et al. 2016; Jones et al. 2006).

In addition to the hazards present, rural populations are vulnerable due to their distance from services associated with urban centers, and the subsequent potential for lack of education and resources to manage drinking water (Nsiah-Kumi 2008; Wescoat, Headington, and Theobald 2007; Zheng and Ayotte 2015). The absence of regulatory water monitoring and treatment increases the likelihood that water-use decisions will be formed heuristically based on their perceived risk (Chen et al. 2012; Hynds, Misstear, and Gill 2013; Martz 1983; Maxwell et al. 1998; Orgill et al. 2013; Patrick 2011).

Therefore, characterizing risk perception can inform exposure associated with consumption of unregulated drinking water, quantify uncertainty associated with human health risk, and better inform risk management and communication (Chowdhury, Champagne, and McLellan 2009). Understanding the perceptions and risks of rural populations dependent on unregulated water sources supports community-based drinking water management decisions to ensure safe and sustainable drinking water. Lastly, it is essential to advance research in the field of holistic human health risk assessment by integrating non-traditional factors to determine their influence on the measure and interpretation of risk.

This thesis contributes knowledge to the field of human health risk assessment by identifying the lack of research addressing the human health risks associated with consumption of unregulated drinking water, and a failure of the field to explore new methods of determining risk. This thesis advanced the field of HHRA and drinking water management by showing how risk perception influences consumption (exposure), and human health risks. In addition to the research, the findings of this thesis supports communities and provides an example of how their perceptions can influence our qualitative understanding of risk, and provide them with an improved perspective from which to manage and communicate risk. This was accomplished by conducting a scoping review of recent human health risk assessment methods applied in the literature, and developing a holistic human health risk assessment that integrated risk perception case studied on two rural communities dependent on unregulated drinking water in Saskatchewan, Canada.

Based on the scoping review and applied HHRA case studies, this thesis set out to:

- 1) review the literature and characterize the methods of HHRA applied to rural communities dependent on unregulated drinking water, and to use this information to inform the field of HHRA and the second objective of this research; and
- 2) conduct a community-based participatory observational case study using Bayesian risk assessment methods to develop a holistic human health risk assessment that integrates a non-traditional factor such as risk perception to improve accuracy, and support risk communication and management.

The field of HHRA lacks a current review of the literature that summarizes the applied methods of human health risk assessment as it pertains to rural populations dependent on unregulated drinking water. This thesis exposes a lack of HHRA research dedicated to rural populations dependent on unregulated water in spite of the global concern regarding access to safe drinking water. Contributing to this apparent lack of dedicated research and in agreement with Pons et al. (2015), this thesis suggests studies are often deficient in effectively identifying and defining the population and receptors of concern. In addition, studies failed to specify if water sources were regulated or not, and were not transparent regarding uncertainty and limitations. The absence of this critical information inhibits research by not allowing researchers to source relevant studies and build on existing research.

Despite the benefits associated with Bayesian and probabilistic methods (Bridges 2003a; Burns et al. 2014; Serre et al. 2003; Zargar et al. 2014), the findings from this thesis indicate the majority of HHRA on unregulated and unspecified drinking water apply deterministic methods. These results suggest that the field of HHRA may be delayed in the adoption of methods that allow for the inclusion of various data types and the quantification of uncertainty to support the integration of non-traditional factors (e.g. behaviour) and holistic HHRA. The use of probabilistic and Bayesian methods of risk assessment can move the field of HHRA forward by: 1) characterizing risk with probability density functions, 2) quantifying uncertainty, 3) identifying gaps in the data

where additional data is required, and 3) integrating non-traditional factors influencing risk (Bridges 2003b; Serre et al. 2003; Zargar et al. 2014; Liu et al. 2012; Wilks et al. 2015).

This research provides a measure of the contribution and influence that risk perception has on exposure and human health risk associated with consumption of unregulated groundwater. Furthermore, it provides an example of how a non-traditional factor, which influences behaviour, can be quantified to characterize risk specific to the characteristics of a community. Communities lacking resources, education, and access to regulated water it is important to understand their perception of risk to ensure effective risk communication and management strategies that support the use of safe and sustainable drinking water.

The community-based research approach and integration of risk perception in HHRA lends to increased participation, cooperation, and value for the opinions of community members that may improve trust and minimize health risks as suggested by Doria (2010). This approach also made it easier to show community members and leaders the discrepancies between their heuristically developed perceptions of risk and the actual risk associated with their drinking water source. Understanding the risk and uncertainty, individuals or communities will have the information required to take responsibility and improve decision-making as suggested by (Markon and Lemyre 2013).

#### 4.2 Future Research

Global rural populations face potential health risks related to water quality hazards associated with unregulated source water. Evolution and improvement in the approach and application of HHRA methods are necessary for a better understanding of the human health risks, and improved risk communication and management in rural populations. Based on conclusions of this thesis future research should:

- Ensure HHRA and drinking water research adequately describes the exposure population and source water to improve the detection of relevant literature to support future research and support the development and application of new approaches and methods.
- Use a holistic approach to HHRA by integrating data from different sources and types when non-traditional factors are suspected of influencing the human health risk.
- Determine the relationship between risk perception and water consumption to attempt to verify and quantify the exposure as it relates to the national average used by Health Canada.
- Determine the effectiveness of risk communication and management strategies based on a holistic and integrated HHRA.

#### 4.3 Limitations

It is possible that literature relevant to rural communities dependent on unregulated drinking water sources was missed in the scoping review process. Screening and full-text review stages

could be subject to interpretation error, and the exclusion of the regulated water sources limited our ability to compare characteristics of unregulated vs. regulated water sources; however, this comparison was not the focus of the scoping review.

Characterization of risk perception and calculated exposure and human health risk associated with arsenic, and consumption of unregulated drinking water for the case-studied communities cannot be assumed for other communities; however, the opportunity exists to apply the model and the probability density functions associated with the model variables to other communities.

A detailed list of limitations specific to each manuscript can be referenced in Chapter 2 and Chapter 3.

#### 4.4 Conclusion

Human health risk assessments applied to rural populations dependent on unregulated drinking water are poorly represented in the literature despite almost half of the global population being rural. For these vulnerable communities, taking a holistic approach to human health risk assessment necessitates the use of probabilistic or Bayesian methods to integrate non-traditional factors influencing risk. Using the methods developed in this thesis, risk perception can be quantified to determine its influence on exposure and human health risk associated with consumption of unregulated drinking water. This approach can be used to improve risk communication and management specific to the needs of communities and support the exploration of non-traditional factors and their influence on the characterization of risk.



## 4.5 References

- Bridges, Jim. 2003a. "Human Health and Environmental Risk Assessment: The Need for a More Harmonised and Integrated Approach." *Chemosphere* 52 (9): 1347–51. doi:10.1016/S0045-6535(03)00469-7.
- . 2003b. "Human Health and Environmental Risk Assessment: The Need for a More Harmonised and Integrated Approach." *Chemosphere* 52 (9): 1347–51. doi:10.1016/S0045-6535(03)00469-7.
- Burns, Carol J., J. Michael Wright, Jennifer B. Pierson, Thomas F. Bateson, Igor Burstyn, Daniel A. Goldstein, James E. Klaunig, et al. 2014. "Evaluating Uncertainty to Strengthen Epidemiologic Data for Use in Human Health Risk Assessments." *Environmental Health Perspectives* 122 (11): 1160–65. doi:10.1289/ehp.1308062.
- Charrois, J. W. A. 2010. "Private Drinking Water Supplies: Challenges for Public Health." *Canadian Medical Association Journal* 182 (10): 1061–64. doi:10.1503/cmaj.090956.
- Chen, Hanyi, Yaying Zhang, Linlin Ma, Fangmin Liu, Weiwei Zheng, Qinfeng Shen, Hongmei Zhang, et al. 2012. "Change of Water Consumption and Its Potential Influential Factors in Shanghai: A Cross-Sectional Study." *BMC Public Health* 12 (1). BMC Public Health: 450. doi:10.1186/1471-2458-12-450.
- Chowdhury, Shakhawat, Pascale Champagne, and P. James McLellan. 2009. "Uncertainty Characterization Approaches for Risk Assessment of DBPs in Drinking Water: A Review." *Journal of Environmental Management* 90 (5). Elsevier Ltd: 1680–91. doi:10.1016/j.jenvman.2008.12.014.
- Corkal, Darrell, W. C. Schutzman, and Clint R. Hilliard. 2004. "Rural Water Safety From the On-Farm Tap." *Journal of Toxicology and Environmental Health, Part A* 67 (20–22): 1619–42. doi:10.1080/15287390490491918.
- Fox, Mary A., Keeve E. Nachman, Breeana Anderson, Juleen Lam, and Beth Resnick. 2016. "Meeting the Public Health Challenge of Protecting Private Wells: Proceedings and Recommendations from an Expert Panel Workshop." *Science of the Total Environment* 554–555. The Authors: 113–18. doi:10.1016/j.scitotenv.2016.02.128.
- Hynds, Paul D., Bruce D. Misstear, and Laurence W. Gill. 2013. "Unregulated Private Wells in the Republic of Ireland: Consumer Awareness, Source Susceptibility and Protective Actions." *Journal of Environmental Management* 127 (September). Elsevier Ltd: 278–88. doi:10.1016/j.jenvman.2013.05.025.
- Jones, A. Q., C. E. Dewey, K. Dore, S. E. Majowicz, S. A. Mcewen, and D. Waltner-Toews. 2006. "Drinking Water Consumption Patterns of Residents in a Canadian Community." *Journal of Water and Health* 4 (1): 125–38. doi:10.2166/wh.2005.001.
- Liu, Kevin Fong-Rey, Che-Fan Lu, Cheng-Wu Chen, and Yung-Shuen Shen. 2012. "Applying Bayesian Belief Networks to Health Risk Assessment." *Stochastic Environmental Research and Risk Assessment* 26 (3): 451–65. doi:10.1007/s00477-011-0470-z.

- Markon, Marie-Pierre L., and Louise Lemyre. 2013. "Public Reactions to Risk Messages Communicating Different Sources of Uncertainty: An Experimental Test." *Human and Ecological Risk Assessment: An International Journal* 19 (4): 1102–26. doi:10.1080/10807039.2012.702015.
- Martz, Diane J. F. 1983. "Variations in Resident Appraisals of Groundwater Quality on Saskatchewan Farms." *Thesis*. University of Saskatchewan.
- Maxwell, Reed M., Susan D. Pelmulder, Andrew F. B. Tompson, and William E. Kastenberg. 1998. "On the Development of a New Methodology for Groundwater-Driven Health Risk Assessment." *Water Resources Research* 34 (4): 833–47.
- Nsiah-Kumi, Phyllis A. 2008. "Communicating Effectively with Vulnerable Populations during Water Contamination Events." *Journal of Water and Health* 6 (S1): s63. doi:10.2166/wh.2008.041.
- Orgill, Jennifer, Ameer Shaheed, Joe Brown, and Marc Jeuland. 2013. "Water Quality Perceptions and Willingness to Pay for Clean Water in Peri-Urban Cambodian Communities." *Journal of Water and Health* 11 (3): 489–506. doi:10.2166/wh.2013.212.
- Patrick, Robert J. 2011. "Uneven Access to Safe Drinking Water for First Nations in Canada: Connecting Health and Place through Source Water Protection." *Health & Place* 17 (1). Elsevier: 386–89. doi:10.1016/j.healthplace.2010.10.005.
- Pons, Wendy, Ian Young, Jenifer Truong, Andria Jones-Bitton, Scott McEwen, Katarina Pintar, and Andrew Papadopoulos. 2015. "A Systematic Review of Waterborne Disease Outbreaks Associated with Small Non-Community Drinking Water Systems in Canada and the United States." Edited by Paula V Morais. *PLOS ONE* 10 (10): e0141646. doi:10.1371/journal.pone.0141646.
- Serre, M. L., A. Kolovos, G. Christakos, and K. Modis. 2003. "An Application of the Holistochastic Human Exposure Methodology to Naturally Occurring Arsenic in Bangladesh Drinking Water." *Risk Analysis* 23 (3): 515–28. doi:10.1111/1539-6924.t01-1-00332.
- Shaheed, Ameer, Jennifer Orgill, Maggie A. Montgomery, Marc A. Jeuland, and Joe Brown. 2014. "Why 'improved' Water Sources Are Not Always Safe." *Bulletin of the World Health Organization* 92 (4): 283–89. doi:10.2471/BLT.13.119594.
- Spence, Nicholas, and Dan Walters. 2012. "'Is It Safe?' Risk Perception and Drinking Water in a Vulnerable Population." *International Indigenous Policy Journal* 3 (3): 9. doi:10.18584/iipj.2012.3.3.9.
- Wescoat, James L., Lisa Headington, and Rebecca Theobald. 2007. "Water and Poverty in the United States." *Geoforum* 38 (5): 801–14. doi:10.1016/j.geoforum.2006.08.007.
- WHO/UNICEF. 2015. "25 Years Progress on Sanitation and Drinking Water: 2015 Update and MDG Assessment." *New York, NY: United Nation's Children's Fund and World Health Organisation*. [http://www.who.int/water\\_sanitation\\_health/publications/jmp-2015-update/en/](http://www.who.int/water_sanitation_health/publications/jmp-2015-update/en/).

- Wilks, M. F., N. Roth, L. Aicher, M. Faust, P. Papadaki, A. Marchis, M. Calliera, et al. 2015. "White Paper on the Promotion of an Integrated Risk Assessment Concept in European Regulatory Frameworks for Chemicals." *Science of The Total Environment* 521–522 (1). Elsevier B.V.: 211–18. doi:10.1016/j.scitotenv.2015.03.065.
- Zargar, Amin, Roberta Dyck, M. Shafiqul Islam, Asish Mohapatra, and Rehan Sadiq. 2014. "Data Fusion Methods for Human Health Risk Assessment: Review and Application." *Human and Ecological Risk Assessment: An International Journal* 20 (3): 807–38. doi:10.1080/10807039.2012.746145.
- Zheng, Yan, and Joseph D. Ayotte. 2015. "At the Crossroads: Hazard Assessment and Reduction of Health Risks from Arsenic in Private Well Waters of the Northeastern United States and Atlantic Canada." *Science of the Total Environment* 505. Elsevier B.V.: 1237–47. doi:10.1016/j.scitotenv.2014.10.089.

## 5 Supplementary Materials

### 5.1 Database Search Terms and Results

Search History – May 8, 2014

Database: **Ovid MEDLINE(R)** <1946 to April Week 5 2014>

Search Strategy:

- 
- 1 (risk adj2 (assessment\* or analys\*)).mp. [mp=title, abstract, original title, name of substance word, subject heading word, keyword heading word, protocol supplementary concept word, rare disease supplementary concept word, unique identifier] (199988)
  - 2 exp Risk Assessment/ or risk assessment.mp. (189599)
  - 3 1 or 2 (201925)
  - 4 water.mp. or exp Water/ (562529)
  - 5 groundwater.mp. or exp Groundwater/ (10264)
  - 6 4 or 5 (564267)
  - 7 exp Health/ or health.mp. (1835151)
  - 8 3 and 6 and 7 (2603)
  - 9 limit 8 to (english language and yr="2000 -Current") (2218)

\*\*\*\*\*

Database: **Ovid MEDLINE(R)** <1946 to April Week 5 2014>

Search Strategy:

- 
- 1 (risk adj2 (assessment\* or analys\*)).mp. [mp=title, abstract, original title, name of substance word, subject heading word, keyword heading word, protocol supplementary concept word, rare disease supplementary concept word, unique identifier] (199988)
  - 2 exp Risk Assessment/ or risk assesment.mp. (175843)
  - 3 1 or 2 (201926)
  - 4 water.mp. or exp Water/ (562529)
  - 5 groundwater.mp. or exp Groundwater/ (10264)
  - 6 4 or 5 (564267)
  - 7 exp Health/ or health.mp. (1835151)
  - 8 3 and 6 and 7 (2603)
  - 9 limit 8 to (english language and yr="2000 -Current") (2218)

\*\*\*\*\*

Database: **Embase Classic+Embase** <1947 to 2014 May 07>

Search Strategy:

- 1 (risk adj2 (assessment\* or analys\*)).mp. [mp=title, abstract, subject headings, heading word, drug trade name, original title, device manufacturer, drug manufacturer, device trade name, keyword] (394871)
- 2 risk assessment.mp. or exp risk assessment/ (343254)
- 3 1 or 2 (394871)
- 4 water.mp. or exp water/ (823914)
- 5 groundwater.mp. or exp ground water/ (21259)
- 6 4 or 5 (825148)
- 7 health.mp. or exp health/ (2676374)
- 8 3 and 6 and 7 (4358)
- 9 limit 8 to (english language and yr="2000 -Current") (3509)

\*\*\*\*\*

Database: **Global Health**

Search Strategy:

- 
- 1 (risk adj2 (assessment\* or analys\*)).mp. [mp=abstract, title, original title, broad terms, heading words] (26366)
  - 2 risk assessment.mp. or exp risk assessment/ (22755)
  - 3 1 or 2 (26366)
  - 4 exp water/ or water.mp. (81998)
  - 5 groundwater.mp. or exp groundwater/ (3695)
  - 6 4 or 5 (82096)
  - 7 health.mp. or exp health/ (276768)
  - 8 3 and 6 and 7 (1811)
  - 9 limit 8 to (english language and yr="2000 -Current") (1631)

\*\*\*\*\*

## Scopus

Search Strategy from ProQuest

May 08 2014 15:11

Set#

Searched for

Databases

Results

S1

all(risk NEAR/2 assessment\* OR risk NEAR/2 analys\*) AND all((water OR groundwater)) AND all(health)

ProQuest Public Health

2590°

S2

(all(risk NEAR/2 assessment\* OR risk NEAR/2 analys\*) AND all((water OR groundwater)) AND all(health)) AND la.exact("English") AND pd(>20000101)

ProQuest Public Health

2538°

S3

((all(risk NEAR/2 assessment\* OR risk NEAR/2 analys\*) AND all((water OR groundwater))  
AND all(health)) AND la.exact("English")) NOT stype.exact("Newspapers") AND  
pd(>20000101)

ProQuest Public Health

2105°

° Duplicates are removed from your search and from your result count.

NOTE: When proquest search run, the numbers come out differently. However, once the last page of results is loaded, the final numbers to change to those above and the export contains 2105 records. The initial results are shown below for completeness:

Set#

Searched for

Databases

Results

S5

((all(risk NEAR/2 assessment\* OR risk NEAR/2 analys\*) AND all((water OR groundwater))  
AND all(health)) NOT stype.exact("Newspapers") AND pd(>20000101)

ProQuest Public Health

## 5.2 Full-Text Review Categorization

<b>THEMES</b>	<b>CATEGORY</b>	<b>DEFINITION/EXAMPLE (if applicable)</b>
<b>Publication Type</b> (choose one)	Journal	Peer reviewed journal
	Conference	Conference document not published
	Paper/Proceeding	
	Thesis	Masters/PhD
	Non-peer reviewed article	Government, public document, opinion paper, etc.
	Other (describe)	Other category of publication
<b>What is the publication year?</b>	Year published	Year of publication
<b>Does the journal/article fit into one of these categories?</b> (choose all	Human Health, Health and Social Sciences, Social Sciences, Toxicology, Epidemiology, Agriculture,	Based on journal title, scope of journal, and/or content of the paper

that apply)	Engineering, Medicine, Environmental/Resource Management	
	Unspecified	Unable to determine the research category
	Other (describe)	Other research field
<b>What is the application of the HHRA?</b> (choose all that apply)	Hypothetical/Theoretical	Method paper, randomly generated data, etc.
	Observational/Field study	Field data is collected or historical data used in 'real life' context
	Unspecified	Unable to determine the application
	Other (describe)	Other application of the HHRA
<b>What is the scope of the HHRA?</b> (choose all that apply)	Integrated Risk Assessment (wide scope)	Ecological & human assessment of risk which may include socio-economic components (Bridges 2003; Sekizawa and Tanabe 2005; WHO/IPCS 2001)
	Human Health Risk Assessment	Only human health risk assessment conducted
	Holistic	Considers non-traditional factors that may influence overall risk; includes non-traditional data integration (Arquette et al. 2002; Bridges 2003; Serre et al. 2003). Does not include the mention of non-traditional factors or interpretation of risk relative to non-traditional data but rather data that contributes quantitatively to the overall determination of risk.
	Other (describe)	Other risk assessment scope was used
<b>How is the study described by the authors?</b> (choose all that apply)	Human Health Risk Assessment	"...is the process to estimate the nature and probability of adverse health effects in humans who may be exposed to chemicals in contaminated environmental media, now or in the future." (United States Environmental Protection Agency (US EPA 2015)
	Risk Assessment	"The probabilities and consequences of adverse events are assumed to be produced by physical and natural processes in ways that can be objectively quantified by risk assessment." (Slovic 1999).
	Health (Risk) Assessment	Risk assessment as defined by Ware (1987) with the broad scope of 'health' and all of its dimensions as identified by Ware (1987) - physical, mental, social function, role function, general health perceptions but more than absence of disease but "presence of well-

		being" (Slovic 1999; Ware 1987).
	Not Reported	Authors don't describe the study in any terms
	Other (describe)	Other study description
<b>What method of HHRA was used?</b> (choose one)	Stochastic/Probabilistic	"Risk assessment that uses probability distributions to characterize variability or uncertainty in risk estimates with the outcome described as a probability distribution rather than a single number" (United States Environmental Protection Agency (US EPA 2001). Chowdhury et al. (2009) provide examples of methods.
	Traditional/Deterministic	Outcomes described with a single number (Health Canada 2010)
	Both	Both probabilistic/stochastic and deterministic methods used
	Unspecified	Unable to identify the method used
	Other (describe)	Other method of HHRA used
<b>Was a standard method used?</b> (choose all that apply)	Health Canada, US EPA, WHO	Standard national or international HHRA method
	Unspecified	Unable to determine method used
	Other (describe)	Other method referenced
<b>Geographic Location</b>	Country	State the country
	Undetermined	Unable to identify the country in which the research was conducted
<b>What is the drinking water source?</b> (choose all that apply)	Ground	Well of any type (e.g. shallow, deep, GUDI, hand-dug, drilled, bored, etc.)
	Surface	Lakes, rivers, streams, dugouts
	Rain collection	e.g. Roof top
	Cistern	Water hauled from any of the above sources
	Bottled	e.g. commercial or regulated bottled water (i.e. bottled water from a government or private treatment facility)
	Undetermined	Unable to identify the water source
	Other (describe)	Other drinking water source
<b>What is the drinking water type?</b> (choose all)	Treated	Subject to regulated treatment
	Not-Treated	Private or unregulated/unknown treatment
	Unspecified	Cannot identify if source is treated or not
	Other (describe)	Other drinking water type



that apply)		
<b>What data informed the risk assessment?</b> (choose all that apply)	Water source tested	As outlined in Health Canada's Guidance on peer review of HHRA for federal contaminated sites in Canada (Health Canada 2010b).
	Proxy tested	e.g. bio-indicators
	Predicted/extrapolated	Prediction modeling or extrapolation
	Based on historical data	Not based on current data but pre-existing information
	Unspecified	Cannot identify data type
	Other (describe)	Other data source
<b>How is the community defined?</b> (choose all that apply)	Cultural/Spiritual	FN, Aboriginal, Indigenous, language, ethnicity
	Geographic	Country, city, town, province, etc.
	Topographic	Watershed
	Unspecified	Unable to identify the community
	Other (describe)	Other definition for the community
<b>What is the population of concern?</b> (choose all that apply)	Urban	As defined by the study and the country in which it was conducted. This is the approach the United Nations takes and the World Bank defines 'rural' when comparing different countries (United Nations 2015).
	Rural	Responsible for establishing source water, not receiving centralized, distributed, treated, and regulated water (e.g. farms, villages, hamlets, private well owners, etc).
	Remote	Geographically isolated or too far from urban centres to receive treated, regulated, distributed water.
	Both	Both urban and rural communities studied
	Unspecified/Undefined	Unable to determine or define the population the population accurately the way it is described by the authors
	Other (describe)	Other description of the population
<b>What are the hazards identified?</b> (choose all	Chemical (natural)	e.g. associated with natural geological characteristics to which the water is exposed
	Chemical (anthropogenic)	e.g. human induced, agricultural, industrial, etc.

that apply) *do not interpret, only answer with reported info	Microbiological/Pathogen	bacteria, protozoans, viruses
	Radiation	e.g. radon, uranium
	Undefined	Unable to determine the hazard
	Other (describe)	Other hazard identified
<b>Who are the receptors?</b> (choose all that apply)	Responsible for Source Water	Receptor is responsible for point of use water quality
	First Nations/Aboriginals	Native/Indigenous populations
	Infants, toddler, child, teen, adults, or senior	Age categories or as described in the study
	General Public	Paper states or describes the general population without distinguishing any age group in particular
	Local Residents	People in the area that may be exposed to the hazard
	Local Farmers and their families	Specifically described as farmers and/or their families
	Employees	People exposed through work place
	<b>Any of the above without age identified?</b>	Note if any of the above did not have the specific age or age category defined
	Undefined	Unable to determine the receptors
	Other (describe)	Other receptor identified in the study
<b>What are the exposure pathways?</b> (choose all that apply)	Oral, dermal, inhalation	Exposure pathways as described by Health Canada (Health Canada 2010b)
	Undefined	Unable to determine exposure pathway
<b>Was uncertainty acknowledged?</b> (choose all that apply) *was it at least discussed	Sufficiency of sampling, analytical detection limits, data gaps, QA/QC, seasonal/environmental factors	(Health Canada 2010a) identifies these areas of potential uncertainty for discussion.
	Quality of historical use information to identify chemicals of potential concern	Relevant if exposure was determined using estimated or historical data.
	<b>Was there a section addressing</b>	An explicit section of the paper was dedicated to addressing uncertainty associated with the risk

	<b>uncertainty?</b>	assessment.
	Other (describe)	Other source of uncertainty identified
<b>What other factors were acknowledged?</b> (choose all that apply) *discussion only	Risk perception	Perception of water or risk associated with any aspect of drinking water
	Economic	e.g. income levels, etc.
	Social	e.g. education, gender, etc.
	Cultural/Spiritual	e.g. homelands, historical use, generational, etc.
	Undefined	Unable to identify other factors acknowledged in the risk assessment
	<b>Geography</b>	Geography is mentioned as influencing exposure to hazards or identifying receptors
	Other (describe)	Other factor acknowledged in the risk assessment
<b>What other factors were applied in the RA?</b> (choose all that apply) *is represented by data that is included in risk assessment analysis	Risk perception	Perception of water or risk associated with any aspect of drinking water
	Economic	See Economic – What other factors were acknowledged?
	Social	See Education – What other factors were acknowledged?
	Cultural/Spiritual	See Cultural/Spiritual – What other factors were acknowledged?
	<b>Geography</b>	Geography data is used to determine areas of increased risk or comparison of regions
	Undefined	Unable to determine if a factor was applied to the risk assessment
	Other (describe)	Other factor applied in the risk assessment
<b>What were the results of the assessment?</b> (choose all that apply)	Exposure assessment, hazard/toxicology assessment, hazard quotient	As outlined in HC Guidance on peer review of HHRA for federal contaminated sites in Canada (Health Canada 2010b).
	Epidemiological assessment/analysis	Use of epidemiological studies in the evaluation/setting of microbiological guidelines for recreational water, wastewater re-use, and drinking water. As defined by Blumenthal et al. (2001)not Ryan (Ryan 2003) in which epidemiological information informs a full risk assessment.
	Qualitative assessment	Differs from quantitative because conclusions are based on 'hazard qualitative description and potency' not DNELs, and risk characterization is justified not

		calculated (European Chemicals Agency 2012).
	Other (describe)	Other result was provided
<b>Did the journal/article conclude the risk assessment?</b> (choose one)	Yes, quantitatively.	Quantitative result - has a quantified result stating there is a risk
	Yes, qualitatively.	Qualitative result - has a description identifying a risk.
	Yes, both quantitative & qualitative	Both qualitative and quantitative conclusions were made
	No	No conclusion was made by the authors
	Undefined	Cannot determine if there is a conclusion or not
	Other (describe)	Other conclusion was provided
<b>What gaps in the literature are identified?</b>	Literature gaps	List gaps in research as identified by the authors
	Describe literature gaps	

### 5.2.1 References

- Arquette, Mary, Maxine Cole, Katsi Cook, Brenda LaFrance, Margaret Peters, James Ransom, Elvera Sargent, Vivian Smoke, and Arlene Stairs. 2002. "Holistic Risk-Based Environmental Decision Making: A Native Perspective." *Environmental Health Perspectives* 110 (s2): 259–64. doi:10.1289/ehp.02110s2259.
- Blumenthal, Uj, Jm Fleisher, Sa Esrey, and A Peasey. 2001. "Epidemiology – a Tool for the Assessment of Risk." In *Water Quality: Guidelines, Standards and Health.*, edited by Lorna Fewtrell and Jamie Bartram, 135–60. London, UK: IWA Publishing. <http://discovery.ucl.ac.uk/76470/>.
- Bridges, Jim. 2003. "Human Health and Environmental Risk Assessment: The Need for a More Harmonised and Integrated Approach." *Chemosphere* 52 (9): 1347–51. doi:10.1016/S0045-6535(03)00469-7.
- Chowdhury, Shakhawat, Pascale Champagne, and P. James McLellan. 2009. "Uncertainty Characterization Approaches for Risk Assessment of DBPs in Drinking Water: A Review." *Journal of Environmental Management* 90 (5). Elsevier Ltd: 1680–91. doi:10.1016/j.jenvman.2008.12.014.
- European Chemicals Agency. 2012. "Practical Guide 15: How to Undertake a Qualitative Human Health Assessment and Document It in a Chemical Safety Report." Helsinki, Finland. <http://echa.europa.eu/>.
- Health Canada (HC). 2010a. "Guidance on Human Health Preliminary Quantitative Risk

Assessment (PQRA), Version 2.0. Part I.” Ottawa, Ont.

———. 2010b. “Guidance on Peer Review of Human Health Risk Assessments for Federal Contaminates Sites in Canada, Version 2.0.” *Health Canada*. Ottawa, Ont.

Ryan, Louise. 2003. “Epidemiologically Based Environmental Risk Assessment.” *Statistical Science* 18 (4): 466–80. doi:10.1214/ss/1081443230.

Sekizawa, Jun, and Shinsuke Tanabe. 2005. “A Comparison between Integrated Risk Assessment and Classical Health/environmental Assessment: Emerging Beneficial Properties.” *Toxicology and Applied Pharmacology* 207 (2): 617–22. doi:10.1016/j.taap.2005.01.047.

Serre, M. L., A. Kolovos, G. Christakos, and K. Modis. 2003. “An Application of the Holistochastic Human Exposure Methodology to Naturally Occurring Arsenic in Bangladesh Drinking Water.” *Risk Analysis* 23 (3): 515–28. doi:10.1111/1539-6924.t01-1-00332.

Slovic, Paul. 1999. “Trust, Emotion, Sex, Politics, and Science: Surveying the Risk-Assessment Battlefield.” *Risk Analysis* 19 (4): 689–701. doi:10.1111/j.1539-6924.1999.tb00439.x.

United Nations. 2015. “Department of Economic and Social Affairs: Population Division.” <https://esa.un.org/unpd/wup/General/GlossaryDemographicTerms.aspx>.

United States Environmental Protection Agency (US EPA). 2001. “Risk Assessment Guidance for Superfund (RAGS) Volume III - Part A: Process for Conducting Probabilistic Risk Assessment, Appendix B.” Vol. III. <http://www.epa.gov/oswer/riskassessment/rags3adt/>.

———. 2015. “Human Health Risk Assessment.” <https://www.epa.gov/risk/human-health-risk-assessment>.

Ware, J.E. Jr. 1987. “Standards for Validating Health Measures: Definition and Content.” *Journal of Chronic Disease* 40 (6): 173–480.

WHO/IPCS. 2001. “Framework for the Integration of Health and Ecological Risk Assessment.” *Integrated Risk Assessment*. Vol. II. Geneva, Switzerland. doi:WHO/IPCS/IRA/01/12.

### 5.3 Household Survey

Category	Question	Answer
Well User Information	How many adults aged 20 to 59?	Number of adults
	What is your household income?	< \$24,999
		\$25,000 - \$49,999
		\$50,000 - \$99,999
		> \$100,000
		Prefer not to say

	What is the highest level of education in the household?	< Grade 12 (no diploma)
		High school diploma
		Some college (no degree)
		Associate/technical degree
		Bachelor's degree
		Graduate/professional degree
		Prefer not to say
	How long have you lived here?	Years
Water Consumption	How much water do you drink per day?	Liters/Day
Well Use	Do you use your well water for drinking?	Y/N
Alternate Drinking Water Source	Do you have an alternative source of drinking water? (e.g. bottled, hauled)	Y/N
Well Safety	Do you think your well water is safe to drink?	Y/N
Previous Water Sampling <sup>a</sup>	Have you previously tested your well water? For metals?	Y/N
	Have you ever been notified that you should not drink your water?	Y/N
Water Treatment	Do you operate or own any water treatment equipment? (e.g. softener, filters or reverse osmosis)	Y/N

<sup>a</sup> 'Y' was also selected if comments in the survey indicated they had received water quality results exceeding drinking water guidelines but failed to understand those results as a 'notice to not consume'.

#### 5.4 Example of Model Code for Bayesian Human Health Risk Assessment

```

model {
###TIME related variables###
res.t ~dunif(20, 59) #residence time - uniform distribution for adults age 20-59
exp.f ~dunif(0.93, 0.96) #i.e. between 340-350 days per year
lif.exp~dnorm(life.ex,tau) #80 year life expectancy assuming SD = 10
#semi-informative using mean=10 & variance = 0.8 in Parameter Solver
    sigma ~ dgamma(125,12.5) #shape=125, scale=0.08, rate=12.5
    tau <- pow(sigma, -2) #calculated tau prior for the precision on the life.ex.dist
###CHEM.EXP###
chem.c~ dlnorm(LNAs.ugl,tau2) #chemical exposure
    LNAs.ugl~ dnorm(0, 0.00001) #likelihood for mean (LN study mean) & uninformative
prior

```

```

sigma2 ~ dgamma(0.0001,0.0001) #likelihood for SD (LN study SD) & uninformative
prior
tau2 <- pow(sigma2, -2) #calculated tau2 for the precision on the As chemical
distribution
h2o.in ~ dlnorm(d.rate, tau3) #water consumption
d.rate<- intake*t.risk #calculate drinking water rate with HC x P(use|s/ns)
intake~dnorm(0,0.00001) #likelihood for mean on LN HC calculated drinking water rate
sigma3 ~ dgamma(0.0001,0.0001) #likelihood for SD (LN HC SD) & uninformative
prior
tau3 <- pow(sigma3, -2) #calculated tau3 prior for the precision on water intake
#probability of use| the perception they feel it is safe or not safe to drink
t.risk <- ((prop.s*prob.s)+(prop.ns*prob.ns)) #sum of P(use|s)+P(use|ns)
prob.s ~ dbeta(27,1) #proportion that think it is safe and drink = 26
#proportion that think it is is safe and don't drink = 0
prob.ns ~ dbeta(8,12) #proportion that think it is not safe and drink = 7
#proportion that think it is not safe and don't drink = 11
prop.s ~ dbeta(57,19) #proportion that answered safe to drink = 26
#proportion that answered not safe to drink = 18
prop.ns ~ dbeta(19,57) #proportion that answered not safe to drink = 18
#proportion that answered safe to drink = 26

####RECEPTORS####
recp.t~dlnorm(adu.wt, tau4) #receptor weight
adu.wt~dnorm(0,0.00001) #likelihood for mean (LN HC adult body weight) &
uninformative prior
sigma4~dgamma(0.0001,0.0001) #likelihood for SD (LN HC SD) & uninformative prior
tau4<-pow(sigma4, -2) #calculated tau4 prior for the precision on adult weight

####EXPOSURE####
## Dose (mg/kg bw/day) = Cw × IRw × RAFOral × D2 ×D3×D4/ BW × LE ##
exp <- (chem.c* h2o.in * exp.f * res.t) / (recp.t* lif.exp) #exposure in mg or ug/bw.day

####RISK####
risk.adu <- exp/SF.As #risk for carcinogenic substance EDI*(1/slope factor)
p.ILCR <- step(risk.adu - 0.00001) #step fuction to get probability of population over 1:100,000
}

Data
list(life.ex=80, LNAs.ugl=-0.207 , sigma2=1.24 , intake=0.28, sigma3=0.50, adu.wt=4.24,
sigma4=0.20, SF.As = 1800)

Inits
list( res.t=30, h2o.in = 1, d.rate = 1, recp.t = 75, adu.wt=75, sigma = 0.001, sigma3 = 0.001,
sigma2=0.001, sigma4=0.001, res.t=20)

```

## 5.5 Community Tap Water Excursions

Tap Water Quality	Drinking Water <sup>a</sup> Guideline	Number (Percent) Excursions RM184 ( <i>n</i> = 76)	Number (Percent) Excursions BOFN ( <i>n</i> = 45)
<i>Maximum Acceptable Concentration</i>			
Arsenic	10 µg/L	6 (8)	0 (0)
Boron	5000 µg/L	1 (1)	0 (0)
Barium	1000 µg/L	0 (0)	0 (0)
Cadmium	5 µg/L	0 (0)	0 (0)
Total Coliform Bacteria	0 organisms/100 mL	39 (51)	23 (51)
Chromium	50 µg/L	0 (0)	0 (0)
<i>Escherichia coli</i>	0 organisms/100 mL	7 (9)	1 (2)
Flouride	1500 µg/L	0 (0)	0 (0)
Nitrate	45000 µg/L	15 (20)	7 (16)
Lead	10 µg/L	2 (3)	0 (0)
Selenium	10 µg/L	8 (11)	0 (0)
Uranium	20 µg/L	12 (16)	0 (0)
<i>Aesthetic Objectives</i>			
Chloride	250,000 µg/L	4 (5)	0 (0)
Copper	1000 µg/L	0 (0)	1 (2)
Iron	300 µg/L	16 (21)	15 (33)
Hardness	800,000 µg/L	27 (36)	0 (0)
Magnesium	200,000 µg/L	6 (8)	0 (0)
Manganese	50 µg/L	32 (42)	25 (56)
Sodium	300,000 µg/L	15 (20)	1 (2)
pH	6.5-9.0 pH Units	24 (32)	0 (0)
Sulphate	500,000 µg/L	34 (45)	0 (0)
Total Alkalinity	500,000 µg/L	18 (24)	1 (2)
Total Dissolved Solids	1,500,000 µg/L	35 (46)	0 (0)
Zinc	5000 µg/L	0 (0)	0 (0)

<sup>a</sup> compared to Water Security Agency. 2017. *Saskatchewan's Drinking Water Quality Standards and Objectives*. Saskatchewan, Canada. <http://www.saskh20.ca/pdf/epb507.pdf>