Impact Assessment, Risk Assessment and Artificial Intelligence



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Introduction

Risks affecting development of Projects are driven by Financial, Technical, Environmental, Social and Governance factors

Context in BC:

- Only 1 out of 2 mining projects with approved EACs are built and operate; and 1 out of 4 is built in the predicted timeline (Collard,R. et.al., 2024)
- There are cases of mining projects generating residual impacts in a materially different way than predicted in the Environmental Assessments
- Environmental Assessment Act of 2018 requires that risks and uncertainties associated to effects of a project be considered in every assessment

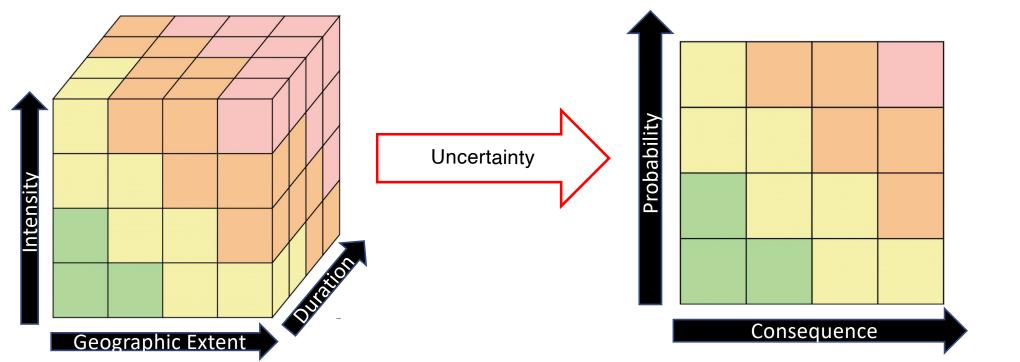
Can Artificial Intelligence assist in assessing the risk of residual effects manifesting themselves in a worse manner than predicted in the Environmental Assessment?



Impact Analysis and Risk Analysis

Impact Assessment 3D Matrix



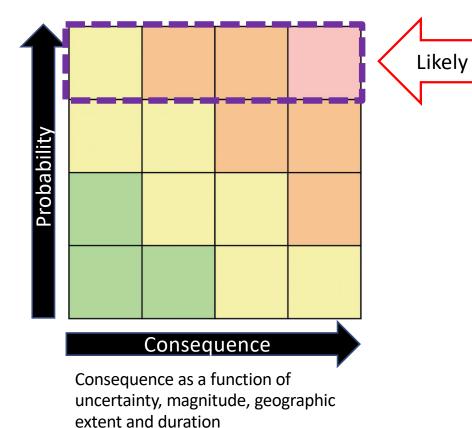


The significance of a residual effect is driven by its magnitude, geographic extent and duration (Paredes, IAIA 2022)

Risk is the effect of uncertainty on achieving objectives (ISO 14001)



Risk is a Function of Probability and Consequence



Probability

Residual effects are considered likely; because the environmental assessment process discards interactions without the potential to generate effects

Residual effects could generate more severe consequences than anticipated

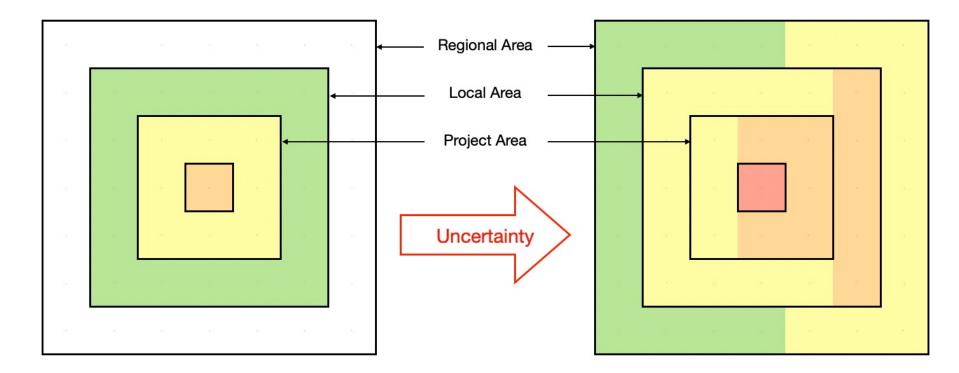
The higher the uncertainty, the higher the risk



Risk is a Function of Probability and Consequence

Consequence

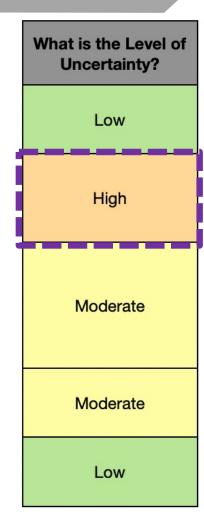
Uncertainty amplifies the residual effect; creating the potential for it to be of higher magnitude; larger geographic extent or longer duration; therefore, generating a more severe consequence





Characterization of Uncertainty

Factors Influencing Uncertainty	Sources of Uncertainty
Knowledge of the Valued Component	Limitations in the understanding of processes, interactions or behaviour. The mitigation is reduced when studies have been conducted during several years to understand the behavious of the VC
Mitigation Effectiveness	Mitigation effectiveness is difficult to predict because of complex systems or human behaviour. The uncertainty is reduced when the results of monitoring programs demostrate that mitigation measures are producing predicted results or achieving expected outcomes on the VC.
Modelling of the Effect	Inadequate, over-simplification, omission of processes. Uncertainty is reduced when appropriate models that chosen to predict the behaviour of the VC and/or conservative assumptions are used to set up modelling scenarios. Uncertainty is reduced when models are calibrated using information collected in the field. Also, uncertainty is mitigated when sensitivity analysis are conducted for modelling scenarios.
Data Used to Feed Models	Limitation is data availability or quality, spatial or temporal resoulution challenges, poorly known model parameters. Uncertainty is reduced when conservative assumptions are used to deal with information gaps
Interpretation of Results	Values or terms are interpreted differently by differently people. Uncertainty is reduced with Peer Review or Senior Review of the results of the assessment





Source: BC EAO Effects Assessment Policy, 2020

Characterization of Uncertainty

Mitigation Effectiveness

- Existing operations can measure mitigation effectiveness based on environmental monitoring
- Environmental assessments rely on available information to estimate effectiveness of proposed mitigation
- Section 30 of BC Environmental Assessment Act requires reporting of mitigation effectiveness
- Global Industry Standard for Tailings Management requires regular reviews of management systems to assure effectiveness

Complex Systems







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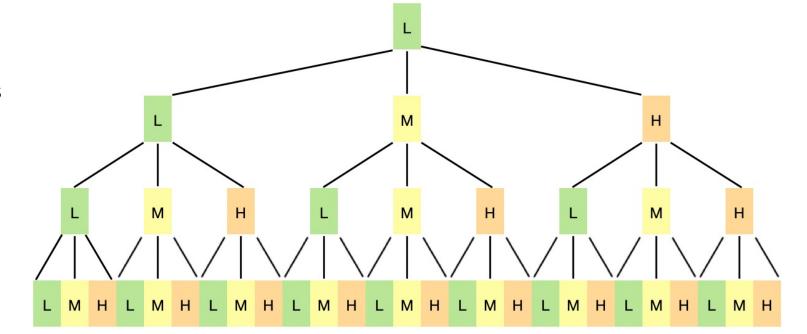
Artificial Intelligence

Machine Learning Algorithms

- Supervised Learning Algorithms
- Unsupervised Learning Algorithms
- Reinforcement Learning Algorithms

Supervised Learning Algorithms

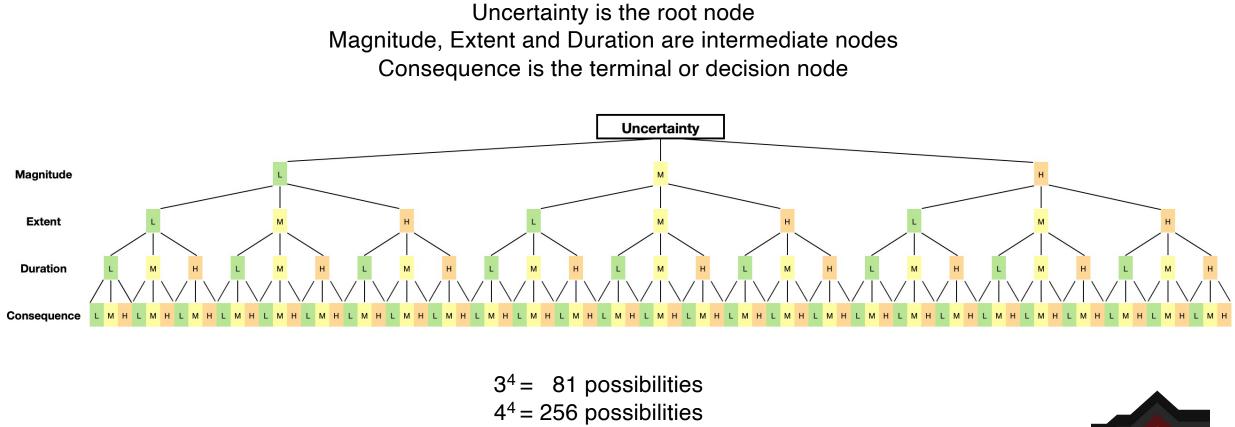
- Linear Regression
- Logistic Regression
- Decision Trees
- Random Forests
- Support Vector Machines
- Neural Networks



A Decision Tree is proposed to derive the value of a **target variable** (i.e. Consequence) by using simple decision rules inferred from the data features (i.e. Uncertainty, Magnitude, Geographic Extent and Duration).



Supervised Learning Algorithms – Decision Trees

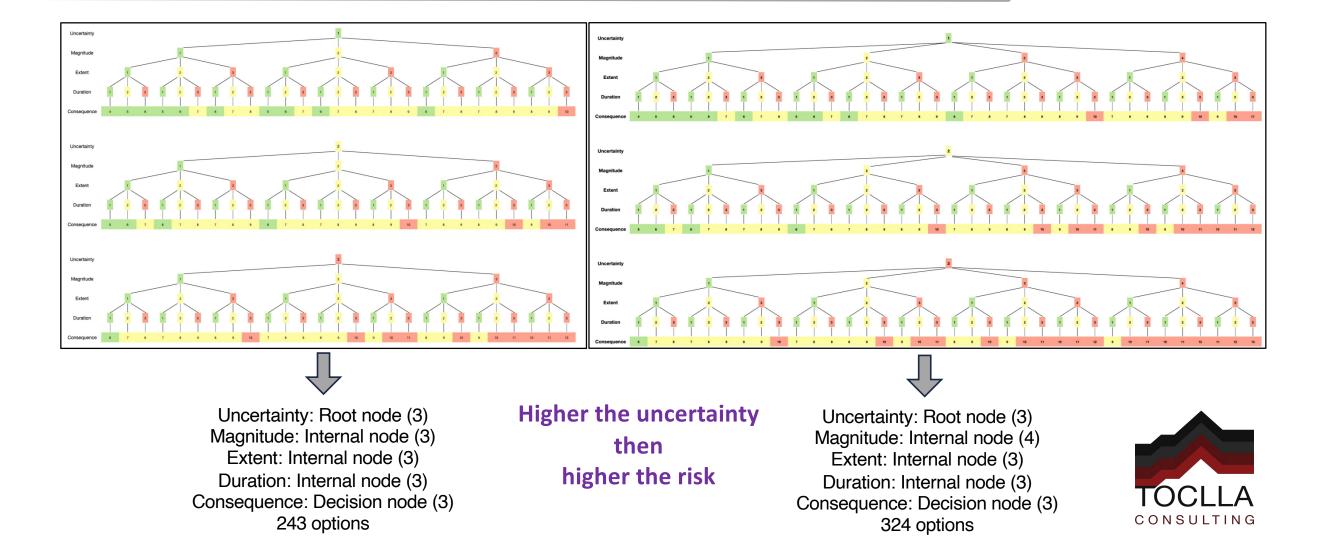


 $5^4 = 625$ possibilities



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Supervised Learning Algorithms – Decision Trees



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Project Cases Used for Comparative Analysis



Assessment Report: Highland Valley Copper Mine Life Extension

WITH RESPECT TO THE APPLICATION BY TECK HIGHLAND VALLEY COPPER PARTNERSHIP TO EXTEND THE LIFE OF THE HIGHLAND VALLEY COPPER MINE

Assessment Report: Cariboo Gold Project

WITH RESPECT TO THE APPLICATION BY OSISKO DEVELOPMENT CORP. FOR AN ENVIRONMENTAL ASSESSMENT CERTIFICATE

WRITTEN BY THE BRITISH COLUMBIA ENVIRONMENTAL ASSESSMENT OFFICE

SEPTEMBER 2023

WRITTEN BY THE ENVIRONMENTAL ASSESSMENT OFFICE ON BEHALF OF THE PROVINCE OF BRITISH COLUMBIA





Assessment Report for Ksi Lisims LNG

WITH RESPECT TO THE APPLICATION BY NISGA'A NATION,

ASSESSMENT ACT (2019) AS A SUBSTITUTED ASSESSMENT

WRITTEN BY

NOVEMBER 12, 2024

ROCKIES LNG LIMITED PARTNERSHIP AND WESTERN LNG LLC FOR

THE ENVIRONMENTAL ASSESSMENT ACT (2018) AND THE IMPACT

AN ENVIRONMENTAL ASSESSMENT CERTIFICATE PURSUANT TO

THE BRITISH COLUMBIA ENVIRONMENTAL ASSESSMENT OFFICE



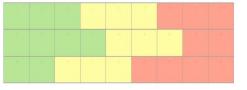
Algorithms and Criteria for Comparative Analysis

Six (6) algorithms were developed:

Two (2) equations to estimate consequence as a function of Uncertainty (U), Magnitude (M), Extension (E) and Duration (D): (A1) $C = \sum (U,M,E,D)$ (A2) $C = \sum (4xU,3xM,2xE,D)$

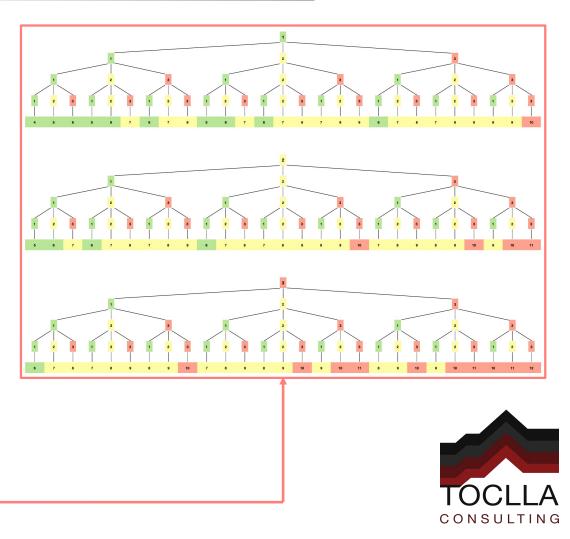
Three (3) sets of criteria were applied to rank consequence:

- (C1) Balanced Risk Tolerance
- (C2) High Risk Tolerance
- (C3) Low Risk Tolerance

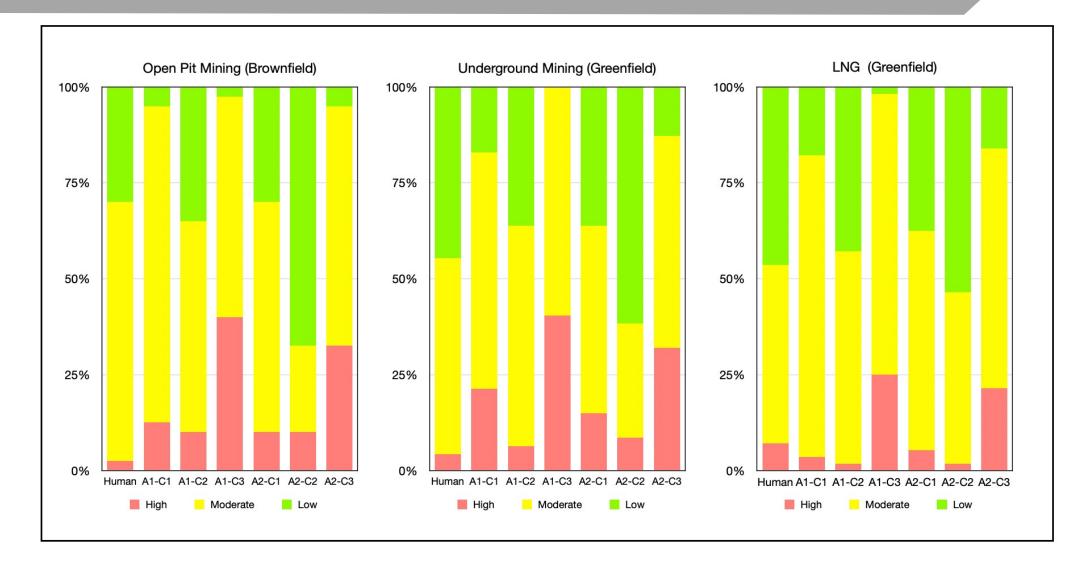


Six (6) scenarios were run for comparative analysis:

1	2	3	4	5	6
A1-C1	A1-C2	A1-C3	A2-C1	A2-C2	A2-C3
	•		•	•	

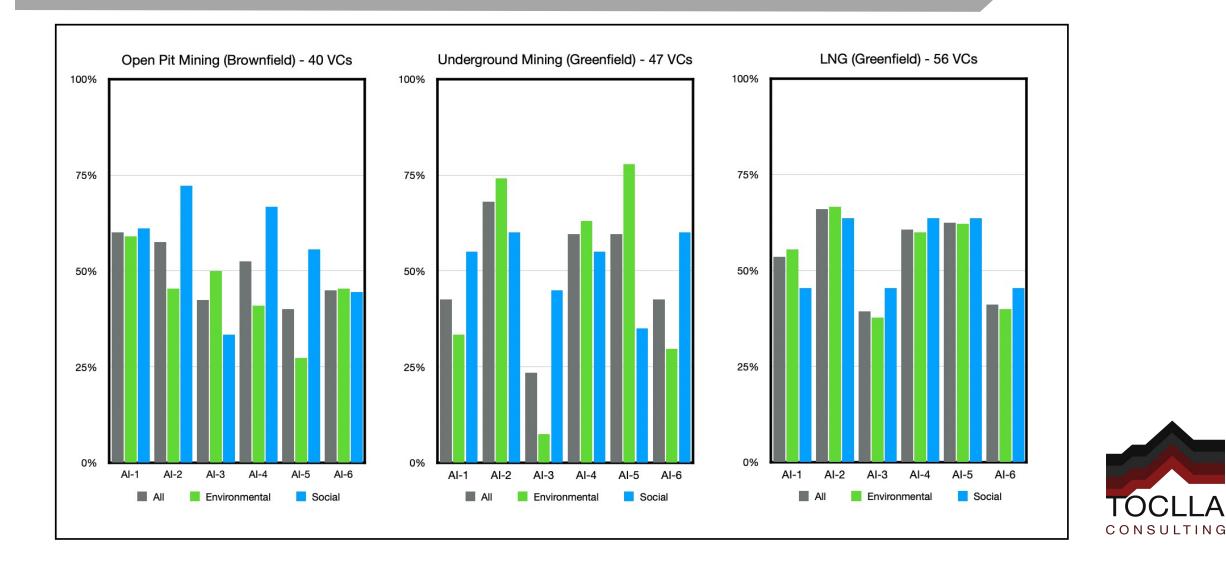


Comparative Analysis Results – Risk Rankings





Comparative Analysis Results – Correlations



Comparative Analysis Results – Open Pit Mining (Brownfield)

Valued Component	Human	AI1-C1	A1-C2	A1-C3	A2-C1	A2-C1	A2-C3
Air Quality		-			-	-	-
Noise			-	-	-	-	
GW Quantity	-				-	-	
GW Quality	-						
GW Quantity (BASA)	-	-			-		
Surface Water Quantity	-						
Surface Water Quality	-						
Surface Water Quantity (BASA)					-		
Fish Health	-	-		-			
Fish Habitat Quantity		-		-	-		-
Fish Habitat Quality	-		-		-	-	
Fish Health (BASA)		-			-		
Wildlife Habitat		-	-		-	-	-
Wildlife Movement		-		-			
Wildlife Mortality		***					
Wildlife Health					-		
Terrestrial Ecosystem Abundance	-	-	-	-	-		-
Terrestrial Ecosystem Productivity	-			-	220		
Wetland Ecosystems		-	-		-	-	
Sensitive Ecosystems	-	-			-		1.00
Listed Ecological Communities	-		-				
Plant Species of Interest	-		-	-	-		-
Health and Emergency Response Services	-		***			-	
Food Security		-	-	-			-
Employment	-		-	-	-		
Housing	-				-		2.55
Public Safety			1.1	-			-
Environmental Determinants of Health	-				-	-	-
Employment			-		-	-	
Economic Benefits	-	-			-		
Housing	-			-			
Social Infrastructure						-	
Traffic				1000 - 100 -		-	100 Sec. 10
Range and Forest Tenures		1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 -					e
Traplines				1000 T			100 CT
Public Use and Access		-					
Visual Resources			-		-		-
Archaeological Sites		-		-			
Heritage Sites	-	-		-			-
Paleontological Sites		-	1.00				-

Valued Component	Human	AI1-C1	A1-C2	A1-C3	A2-C1	A2-C1	A2-C3
Groundwater Quantity	-	-	-	-	-	-	
Groundwater Quality		-	-	-	-	-	-
GW Quantity (BASA)	-	-	-		-		
Surface Water Quantity			-	-	-		
Surface Water Quality	-						
Wildlife Habitat		-	-			-	

All artificial intelligence runs result in High risks for impacts to surface water and groundwater

Surface water and groundwater assessments were declared to have high levels of uncertainty



Comparative Analysis Results – Underground Mining (Greenfield)

Valued Component	Human	A1-C1	A1-C2	A1-C3	A2-C1	A2-C2	A2-C3
Air Quality (SO2 & CO)		-		-	14 A.		
Air Quality (NO2)		(H)			1	-	
Air Quality (PM)	-		-				
Low Frequency Noise		1.00		-	201		
Noise and Vibration				-	-		
Groundwater Quantity	-				14	-	
Groundwater Quality	0.1	-	2				
Surface Water Quantity	-			-		-	-
Surface Water Quality		240				1	12
Fish Health	-	-	-	-	-		_
Fish Habitat	-	-	-	-	1		-
Aquatic Resources		1000			1.2		
Soil Quantity	2	1.0	2	-	20		1 3
Soil Quality				-	40		
Terrain Stability				-			
Ecosystem Function		-		-			-
Wetland Function		171		-	875		1.50
Plant Communities of Interest		- 1943 1		-			
Plant Species of Interest		-	8	-			
Amphibians	12	846	2	-	12	1	123
Songbirds				-	÷2		-
Raptors		1.00	1 2				
Bats	7.	1.000			125		
Ungulates (Except Caribou)		(H)	1	-			-
Southern Mountain Caribou	-	-	-			-	
Large Carnivores (Including Grizzly Bears)	1. A.		-	-	-		-
Medium Carnivores (including Wolverines)	10	1.070		-	72		1.00
Private Property	-	-	-	-	1.00		
Tenured Land and Resource Use	-	1. The second		-	100		
Access to Contemporary Land and Resource Use	-	243	8	4	820	8	
Conditions to Contemporary Land and Resource Use	-	-	-	-			-
Public Land and Resource Use							
Enjoyment and Experience for Contemporary Use of the Land	-	040					-
Lighting	-	-	-	-	100		-
Visual Resources	-	-	-		-	-	-
Traffic	-			-		8	-
Housing and Accommodation		-	-	-	-	1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 - 1999 -	-
Employment	-			-	1		
Economic Benefits	5		100 B	1.1		1	1.5
Health Infrastructure and Services	-		-		-	-	
Population Health		1.050			1. Jan 1.		1.00
Human Health	- 1						
Ecological Health	-	-		-			1.141
Plants Species of Cultural Significance				-	-		
Traditional Foods		-	-			-	
Indigenous Language and Culture	2.1	1.1.1.1.1.1.1.1.1			120	-	
District of Wells Arts Sector		9 <u>0</u> 9	2	2	121		100

Valued Component	Human	A1-C1	A1-C2	A1-C3	A2-C1	A2-C2	A2-C3
Air Quality (PM)		-			~	~	
Groundwater Quality		-			-	-	
Surface Water Quantity		~		-	~	-	-
Surface Water Quality						5 - 5 B 5	
Aquatic Resources		~	-			-	
Bats	-	-	-		-	-	-
Southern Mountain Caribou		-	-		-	-	
Housing and Accommodation	-	-	-		-	14 - C	-
Health Infrastructure and Services		~		-	-		÷
Human Health	-	-			-	-	
Indigenous Language and Culture		-			-		

Two High risks identified by human analysis were linked to both High and Low levels of uncertainty

High risk tolerance criteria result in the highest correlation with human analysis; while low risk tolerance criteria result in the lowest correlation



Comparative Analysis Results – LNG (Greenfield)

Valued Component	Human	Al1	AI2	AI3	Al4	AI5	Al6
Noise (Construction)	_			_			
Noise (Operations)	_	-		-			
Vibration (Construction)		1.	100				1.2
Air Quality (Construction)				-			1-1
Air Quality (Operations)		-	-	-	-		
Groundwater Quantity							1020
Surface Water Quantity							
Surface Water Quality							
Marine Water quality							
Marine Habitat							
Marine Behaviour (Sensory Disturbances)							1
Marine Benaviour (Sensory Disturbances) Marine Mortality							
Marine Mortainy Marine Northern Abalone (Species at Risk)	-		-	-	-	-	
Marine Northern Adaione (Species at Risk) Marine Fish		-		-			-
		-		-	-		-
Marine Mammals 1 Marine Mammals 2		-			-	-	
		-	-		-	-	
Sea Turtle (Species at Risk)		-					
Phytoplankton Density		-		-			-
Fresh Water Fish Habitat	-	-		-	-		1000 (C)
Fresh Water Fish Health		-	1	-			-
Wildlife Habitat (Direct Loss)	-	-	-		-	-	
Wildlife Habitat (Indirect Loss)	1	-			-	1.000	1.000
Wildlife Movement	-	-	1.00		-	-	
Wildlife Mortality	-	-	-	-	-		0.000
Migratory Birds		22	243		2	12	
Grizzly Bear	-	1	in the second is		-		1949
Small Mammals/Furbearers	-	-	100 Att 1		-		
Amphibians	-		-	-	-		-
Bats	-	1.00					
Western Screech Owl		1.00	-	-	-	-	
Northern Goshawk	-	1.00		-			1.000
Marbled Murrelet	-	1.1.1	250	.	-	1	2-2
Plant Species of Interest (Direct)	-	-		-			-
Ecological Communities of Interest	-	1	1 n=1	-	-		
Wetlands	2	-	-		<u> </u>		1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 - 1948 -
Plant Species of Interest (Indirect)			1.000	-		-	
Ecological Communities of Interest (Indirect)		-		-			1
Wetlands							
Introduction of Invasive Species	-	-	-		-		
Species and Communities of Interests in Wetlands	-			_			-
Soil Acidification							
Lichen							
Marine Navigation	1	-		-	-		-
Marine Fisheries	-	1000	177		-	1000	
Marine Aesthetic Conditions		-		-	-		1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 - 1997 -
Human Health from Air Quality				-			1. 18 4 8 19
Human Health from Noise	1.1.1	-					223
Community Health	10.000	123	100	12 C			8 <u>4</u> 8
Community Wellness	-	_	-	-	-	-	-
Food Security	-			_	-		
Health Services and Infrastructure	_	-		_	-		
Regional Business							
Regional Economy							
Infrastructure and Services Accommodation Availability and Affordability			-				
				-			
Transportation Infrastructure		1.00	873	1.1	2		100

Valued Component	Human	Al1	AI2	AI3	Al4	AI5	Al6
Wetlands							
Introduction of Invasive Species	-	-	-	-	I	-	
Lichen	-		-				
Community Health	-			-			

A higher number of High risks were identified by Human analysis when compared to most Artificial Intelligence outcomes

High risk assigned by Human Analysis were associated to High and Moderate levels of Uncertainty



Conclusions

- Uncertainty can be used as an input for assessing risks derived from residual impacts potentially occurring differently than predicted in Environmental Assessments
- Artificial Intelligence tools can support Human Analysis; to improve consistency and transparency in the assessment of Risk
- Methods for characterization of uncertainty should be advanced; particularly when it relates to the effectiveness of proposed mitigation measures



Let's continue the conversation!

Post questions and comments via chat in the IAIA22 platform.



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