## Idaho Community-Engaged Resilience for Energy-Water Systems Dynamic Risk Assessment Overview & Primer

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#### **I-CREWS Modeling Group Workshop**



#### Goals for the session

Introducing the concepts of risk and risk assessment

Integrating dynamic risk within the context of the project description

Stimulate discussion for May workshop





# Research Component 2 – Model Dynamic Existing & Alternative E-W Configurations Team

#### Leads -

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#### Team members -

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New hires!



## Research Component 2 – Model Dynamic Existing and Alternative E-W Configurations Objectives

- Objective 2.1 Build datasets to develop the Machine Learning (ML) modeling platform.
- Objective 2.2 Build ML models suitable for use in the testbed regions.
- Objective 2.3 Quantify dynamic risk.

This component aims to increase capacity in computational modeling and machine learning, advancing our understanding of –

- (i) scales of data needed to effectively model the risks and losses of multiple interacting stressors on E-W systems, and
- (ii) stressor conditions, tradeoffs and feedback decisions leading to state-shifts.



## **Objective 2.3 – Quantify dynamic risk**

Objective 2.3: Quantify dynamic risk.								
Detern	Determine risk to be assessed and relevant variables contributing to risk.							
	Year 1	Year 2	Year 3	Year 4	Year 5			
2.3.A	Conduct literature review on similar risk problems.  Coordinate with Characterize and Alternative Futures teams.	SETS variables and stressors based on Char- acterization team results established.	'Status quo' futures (to compare) defined with Alternative Futures team.	Risk to be assessed determined.  Food for thought: Suggested discussion topic for May workshop.	Risk definition reiterated.			



## **Objective 2.3 – Quantify dynamic risk**

Objective 2.3: Quantify dynamic risk.								
Derive	Derive or apply relevant models to characterize risk.							
	Year 1	Year 2	Year 3	Year 4	Year 5			
2.3.B	Conduct literature review for any relevant dynamic risk modeling techniques and metrics that could be applied and/or modified.	Explore how SETS variables change with changes to stressors.	Explore how to integrate ML results.	Integrate ML results into SETS variables andstressors.  Dynamic behavior of stressors defined.	Knowledge Holder Inputs coupled and aligned with Alternative Futures team.			



## **Objective 2.3 – Quantify dynamic risk**

#### Objective 2.3: Quantify dynamic risk.

#### Assess risk.

	Year 1	ear 1 Year 2 Year 3		Year 4	Year 5
2.3.C	Explore incorporating risk within the context of resilience or related construct with all teams.	Couldn't think of anything new for Year 2.	E-W Resilience Indicators applied.	Conduct sensitivity analysis of risk/resilience to changes in SETS variables and stressors if needed.	Risk assessed.







#### Risk has been around for a long time

... the appearance of disease in human populations is influenced by the quality of air, water, and food; the topography of the land; and general living habits.

- Hippocrates; Air, Water and Places



### Risk is the possibility of loss due to a hazard

- A hazard is -
  - ...a dangerous factor
  - ...a person or event that induces a dangerous factor

A hazard is an existing or potential condition that can cause injury, illness, or death; damage to, or loss of equipment, property, finances. . .



## Risk is the probability of a hazard occurring multiplied by its consequences

$$Risk = f \times C \tag{1}$$

Frequency and probablility are interchangeable

Consequence and severity are interchangeable

Probability  $\rightarrow$  How often a hazard could occur... once per facility lifetime, three failures per month, etc.

Consequence  $\to$  Expected result of a hazard...degree of injury, property/financial damage, latent cancers, etc.

 $\mbox{Mitigation} \rightarrow \mbox{Action taken to eliminate or reduce risk identified}$ 



## Risk is the expected value of an undesirable event

$$E[X] \equiv \int x f(x) dx \tag{2}$$

$$E[X] \equiv \sum_{i} x_{i} f(x_{i}) \tag{3}$$

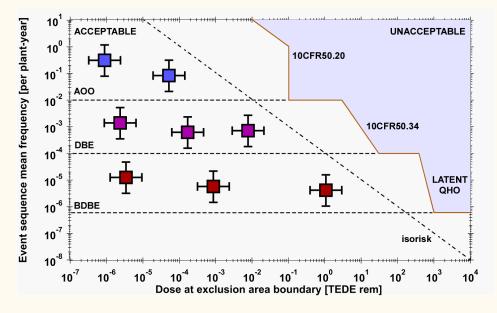
$$f(x) = \lambda e^{-\lambda x} x > 0$$

$$E[X] = \int_0^\infty x \lambda e^{-\lambda x} = -\frac{1+\lambda x}{\lambda} e^{-\lambda x} \Big|_0^\infty = \frac{1}{\lambda}$$

$$E[DICE] = \frac{1}{6} \cdot 1 + \frac{1}{6} \cdot 2 + \frac{1}{6} \cdot 3 + \frac{1}{6} \cdot 4 + \frac{1}{6} \cdot 5 + \frac{1}{6} \cdot 6 = 3.5$$



#### We can visualize risk with the Farmer's chart [1]





[1] Farmer, F. R., 1967. Reactor Safety and Siting: A Proposed Risk Criterion. Nuclear Safety 8, 539

## Risk assessment involves three essential questions [2]

What can go wrong?

How likely is it to happen?

What are the consequences?

[2] Kaplan, S. et al., 1981. On The Quantitative Definition of Risk. Risk Analysis 1, 11 (Kaplan & Garrick is so important that the paper is hosted on the NRC website.)





#### Risk assessment is a retrospective process

Developed by the US Space Program in 50s and 60s

Reactor safety study WASH-1400 was first real use of Probabilistic Risk Assessment (PRA) analysis and techniques

But only after about 75 NPPs designed, built, operating; providing considerable data cohorts

PRA came to prominence after Three Mile Island in 1979 as NRC standard

Severe Accident Risks: An Assessment for Five U.S. Nuclear Power Plants (1990) was first Level III full scope PRA (use of Farmer)

#### Risk assessment involves substantial breadth and depth

Requires diverse expertise – Highly dependent on detailed system analysis

Start by identifying hazards

Then characterize all identified hazards combined to complete a task, process, etc.

Multiple hazards have varying risk

Risk assessment is deterministic and reductionistic and highly data driven

Sophsticated modeling is usually required







### A risk matrix is a high level assessment tool

There are many different types of risk matrices

Severity	Probability						
Severity	Frequent	Likely	Occasional	Seldom	Unlikely		
Catastrophic	E	Е	Н	Н	М		
Critical	E	E	Н	М	L		
Moderate	Н	М	M	L	L		
Marginal	Н	М	L	L	L		
Negligible	М	L	L	L	L		
E – Extreme Ri	sk	M – Moderate Risk					
H – High Risk		L – Low Risk					

What is important is that the team agrees on the terminology and risk definition

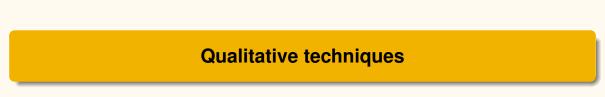


## Risk matrix for cyberattacks on a Nuclear Power Plant [3]

**Table.** Preliminary Hazards Analysis results for selected Nuclear Power Plant systems. The product of Accessibility and Impact nominally reflects the vulnerability of the system as to whether the reactor will be tripped (shut down) or if the operator can take corrective or mitigating actions to recover operability.

Impact	Accessibility						
impact	Frequent	Likely	Occasional	Seldom	Unlikely		
Catastrophic					Reactor controls		
Critical		Spent fuel pool Boron monitoring					
Moderate	Exciter		Steam generator Condenser				
Marginal	al Cooling water systems						
Negligible							





## Qualitative analysis yields meaningful results if you know what you're doing

Preliminary Hazards Analysis (PHA) early to identify hazards

Failure Modes & Effects Analysis (FMEA) to determine high level frequency and consequences; nominal risk measure

Fault trees are top down, deductive failure analysis tool to decompose hazards

Event trees identify accident sequences and consequences

Hazard & Operability Analysis (HAZOP) analyzes how hazards affect system operations





#### Uncertainties come in different flavors but are essential in risk [4]

There are things we know we know

Then there's things we know that we do not know

But there's still things that we don't know we don't know

It is this latter space (unknown-unknowns) that is the most dangerous to risk assessment

[4] Der Kiureghian, A. et al., 2009. Aleatory or epistemic? Does it matter? Structural Safety 31, 105



#### **Aleatory uncertainty is statistical**

Random variations and chance outcomes in the physical world, natural randomness in a process

If a parameter sometimes has one value and sometimes has another values



#### **Epistemic uncertainty is systematic**

Lack of knowledge about the physical world, scientific uncertainty in the model or the process

If a parameter always has either one value or another, but we are not sure which



## **Ethics**

### There is an ethical theory basis for risk

Based on universal rules and principles by Descartes (1596–1650)

Rights ethics by John Locke (not the guy from the Island) (1632–1704)

Duties ethics by Immanuel Kant (1724–1804)

Utilitarianism by Jeremy Bentham and John Stuart Mill (1748–1832),(1806–1873)

'Greatest good for greatest number of people'

The 'Red Wedding' on Game of Thrones is an example of utilitarianism for the greatest benefit to the Lannisters



#### What are some drawbacks of utilitarianism?

Only the greatest good, as a singular body and not distributed among people

Difficulty in quantifying the greatest good

Anthropocentric

Utilitarianism judges by consequences rather than actions

Low probability-high consequence events carry as much weight as high probability-low consequence events





#### What else can be used?

Justice ethics by Rawls (1971)

Each person is to have an equal right to equal basic liberties

Social and economic inequalities are to the greatest benefit of the least-advantaged

Treating everyone equally is a challenge

Environmental justice is gaining acceptance as an alternative

Environmental justice is being applied for the consent-based siting of spent nuclear fuel



#### Other seminal Probabilistic Risk Assessment cases

The Hyatt Horror [5]

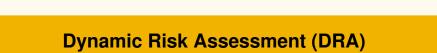
The Pinto Case [6]

WASH1400 documents

Other contemporary cases in media

- [5] Pfatteicher, S. K. A., 2000. The Hyatt Horror: Failure and Responsibility in American Engineering. Journal of Performance of Constructed Facilities 14, 62
- [6] De George, R. T., 1981. Ethical Responsibilities of Engineers in Large Organizations: The Pinto Case. Business & Professional Ethics Journal 1, 1





### Quite simply, DRA refers to time dependent frequencies

$$Risk(t) = f(t) \times C$$
 (4)

Proposed initially as a way to model complex accident scenarios with higher fidelity [7]

[7] Siu, N., 1994. Risk assessment for dynamic system: An overview. Reliability Engineering & System Safety 43, 43



# Typically, some phenomenon is modeled in time and integrated with stochastic models to evolve system risk dynamically [8]

Hazards are advanced in time using appropriate models

Continuous – Integral, differential equations

Discrete - Monte Carlo, Markov

Graphical – Petri nets

Machine Learning - Natural language processing, deep mapping

Bayesian analysis can also be applied to update posterior probabilities in time

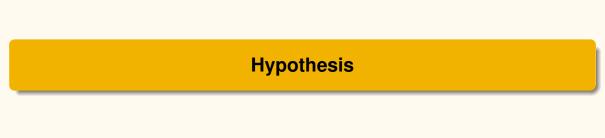
Use of 'bow-tie' analysis to combine event trees and fault trees in time

Fuzzy set theory also can be applied [9]

[8] Aldemir, T., 2013. A survey of dynamic methodologies for probabilistic safety assessment of nuclear power plants. Annals of Nuclear Energy 52, 113

[9] Redfoot, E. K., et al., 2022. Applying analytic hierarchy process to industrial process design in a nuclear renewable hybrid energy system. Progress in Nuclear Energy 145. 104083





### The hazard (stressor) evolves system component response (SETS) over time

Focus on time dependence allows risk to be re-assessed as new data is obtained

Here, we can integrate with ML as we obtain new data from Characterization

Specify system states in time (Alternative Futures?) and defining the transitions between states

Similar to Discrete Event Simulation (DES) [10] – Each state evolves in time as needed, rather than the entire architecture, thereby allowing a more flexible definition of risk.

Can dynamic risk be derived from consequences? –  $Risk(t) = f \times C(t)$ ? (5)

Food for thought: Suggested discussion topic for May workshop.







## A comprehensive review on dynamic risk analysis methodologies [11]

Hazard identification is the first key part of a risk analysis

Dynamic Logical Analytical Methodology – Markov modelling time-dependent state transition matrix

Dynamic Event Tree Analysis – Models dynamic variation of state space

Food for thought: Possible mechanism for Alternative Futures

Bow Tie model – Qualitatively and quantitatively represents a complete scenario from initial hazard to final consequences

Artificial Neural Network – Updated real time frequencies through a back propagation

[11] Raveendran, A., et al., 2022. A comprehensive review on dynamic risk analysis methodologies. Journal of Loss Prevention in the Process Industries 76, 104734



## Dynamic Bayesian network-based operational risk assessment for industrial water pipeline leakage [12]

Applied dyanamic Bayesian network to update frequencies for hydraulic failures

Uses fault trees for hazards where Bayesian updating is used to model frequency

Dry out scenario example - Less flow, protection failure, valve failure

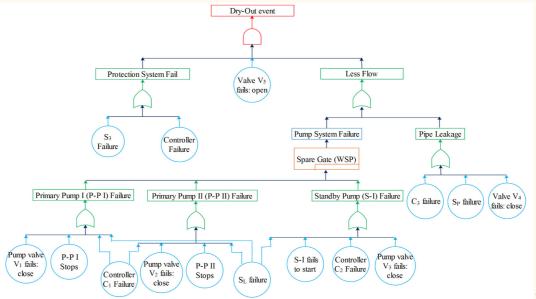
Fault tree is updated at discete time intervals based on failure scenarios

Projecting failures in time can aid in managerial decision-making

[12] Abdelhafidh, M., et al., 2023. Dynamic Bayesian network-based operational risk assessment for industrial water pipeline leakage. Computers & Industrial Engineering 183, 109466



#### Fault tree for dry-out failure





**Figure.** Fault trees should be precise, clear, and detailed. This could use improvements.

### Dynamic risk of coastal flood and driving factors [13]

Combination of urban growth and sea level rise drive significant increases in coastal flood risk

Time dependent component to risk is sea level rise and urban growth

Markov model predicts future land use

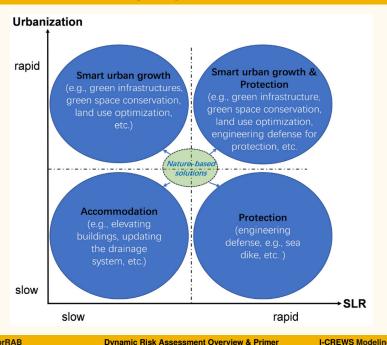
Risk defined as physical damage, economic loss due to coastal flooding

Risk informs urban planning strategies

[13] Xu, L., et al., 2021. Dynamic risk of coastal flood and driving factors. Journal of Cleaner Production 321, 129039



#### Policy alternatives for adapting to coastal flood risk





## Dynamic risk analysis of hydrogen gas leakage using Bow-tie technique and Bayesian Network [14]

Bow-tie and Bayesian Network used to assess risk of hydrogen gas leak from chlorination unit

Bow-tie was used to combine fault trees and event trees

Bayesian network updated posterior & conditional probabilities in the bow-tie in time and can carry out deductive reasoning on the bow tie to update probability

Food for thought: Possible application of ML.

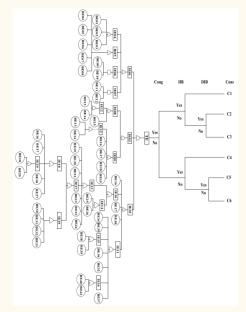
Decrease in flow rate leading to fire or explosion highest risk

[14] Borgheipour, H., et al., 2021. Dynamic risk analysis of hydrogen gas leakage using Bow-tie technique and Bayesian network. International Journal of

**Environmental Science and Technology 18, 3613** 



### Bow-tie diagram of hydrogen gas leakage from chlorination unit





**Figure.** Bow tie combines fault & event trees by the top event (TE).



#### What can we learn from DRA relevant to I-CREWS?

DRA is widely applicable to many topics

Based on fundamental risk principles

Difficult to find example directly relevant to I-CREWS strategic goals

Upside – Demonstrates the novelty of I-CREWS

The 'dynamic' part of risk assessment is strictly focused on frequency analysis

We should agree on what risks we will assess and an overall analytical framework

Food for thought: Is a new definition of DRA needed?



#### References

- 1. Farmer, F. R., 1967. Reactor Safety and Siting: A Proposed Risk Criterion. Nuclear Safety 8, 539.
- 2. Kaplan, S. et al., 1981. On The Quantitative Definition of Risk. Risk Analysis 1, 11.
- 3. Root, S. J., et al., 2023. Cyber Hardening of Nuclear Power Plants. Progress in Nuclear Energy 162, 104742.
- 4. Der Kiureghian, A. et al., 2009. Aleatory or epistemic? Does it matter? Structural Safety 31, 105.
- 5. Pfatteicher, S. K. A., 2000. The Hyatt Horror: Failure and Responsibility in American Engineering. Journal of Performance of Constructed Facilities 14, 62.
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- 9. Aldemir, T., 2013. A survey of dynamic methodologies for probabilistic safety assessment of nuclear power plants. Annals of Nuclear Energy 52, 113.
- 10. Lee, J. et al., 2019. Use of discrete event simulation for material throughput. Nuclear Engineering and Design 345, 183.
- 11. Raveendran, A., et al., 2022. A comprehensive review on dynamic risk analysis methodologies. Journal of Loss Prevention in the Process Industries 76, 104734.
- 12. Abdelhafidh, M., et al., 2023. Dynamic Bayesian network-based operational risk assessment for industrial water pipeline leakage. Computers & Industrial Engineering 183, 109466.
- 13. Xu, L., et al., 2021. Dynamic risk of coastal flood and driving factors. Journal of Cleaner Production 321, 129039.
- 14. Borgheipour, H., et al., 2021. Dynamic risk analysis of hydrogen gas leakage using Bow-tie technique and Bayesian network. International Journal of Environmental Science and Technology 18, 3613.



