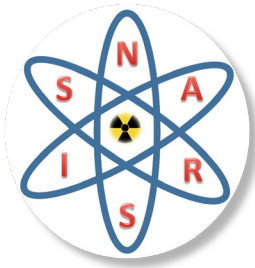




# Application of Bayesian Networks in Multi-Hazard Safety Assessment of Nuclear Power Plants

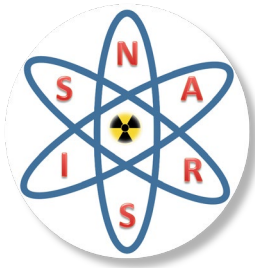
**Varennya Kumar D. Mohan, Philip Vardon, James Daniell, Pierre Gehl, Andreas Schafer, Pieter van Gelder, Venkat Natarajan, Cor Molenaar, Evelyne Foerster, and Florence Ragon**





# Nuclear Power Plant Probabilistic Safety Assessment

- Low probability events are a key consideration in multi-hazard safety assessments of nuclear power plants (NPPs)
- Cascading effects from hazards and associated event sequences could potentially have a significant impact on risk estimates.
- The Bayesian network (BN) can act as a framework to consider aforementioned statistical dependencies between various hazards in multi-risk analyses of nuclear power plants.



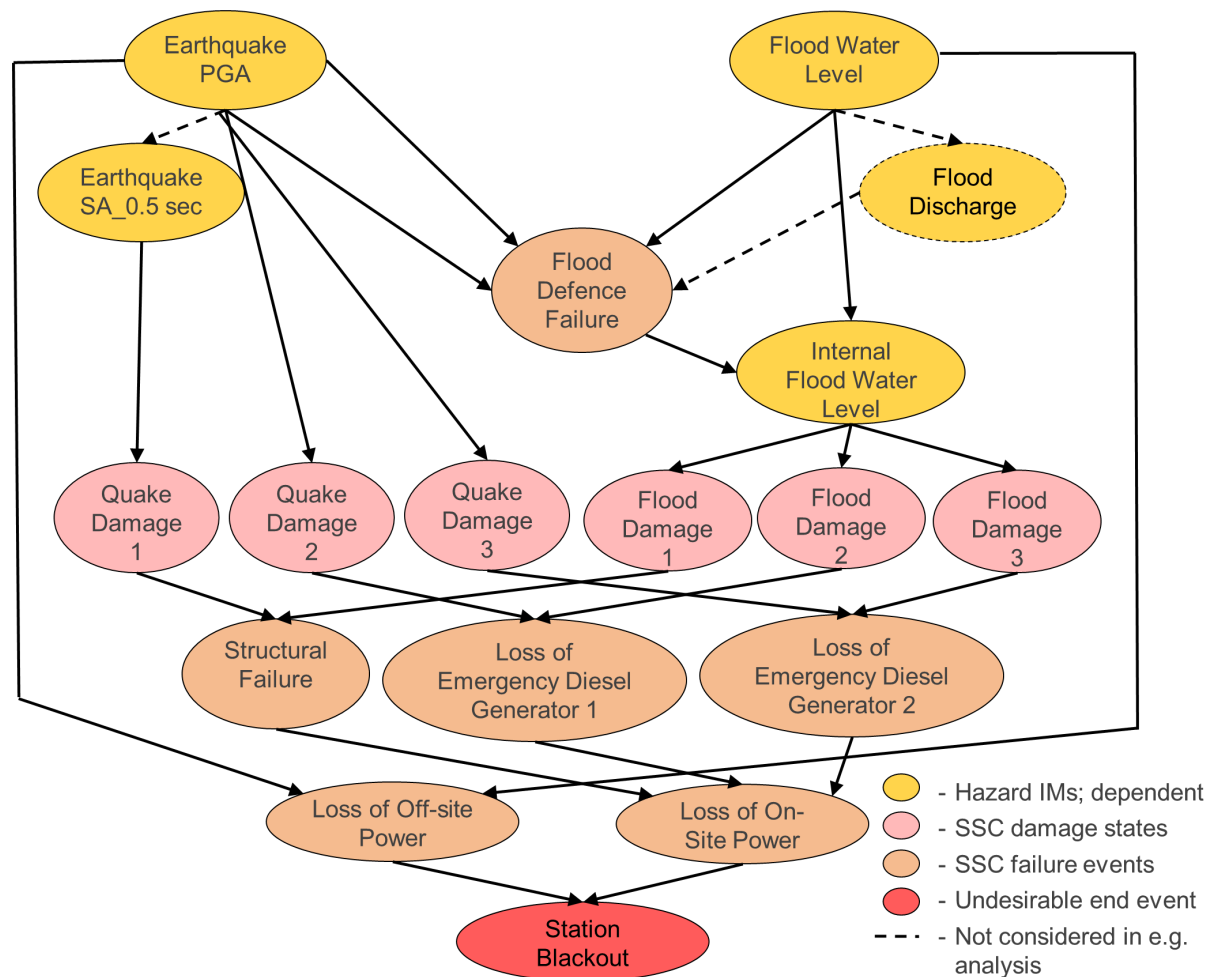
# BNs in NPP Risk - Example Implementation

## ➤ Simplified BN implementation in Nuclear PSA

- ☐ Simplified scenario used to link hazards, fragilities and NPP end event - Station Blackout - SBO
- ☐ Risk integration procedure developed to move from multi-hazard analysis to final risk estimate



# Example Implementation Scenario - SBO

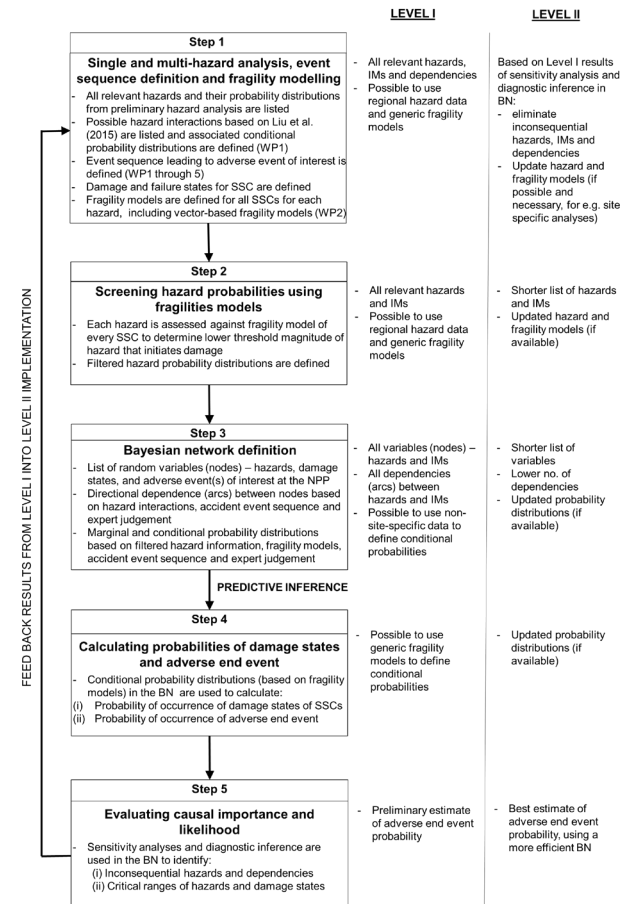




# Risk Integration

## ➤ NARSIS RISK INTEGRATION METHODOLOGY

- ❑ Step-wise, 2-Level procedure
- ❑ Multiple hazards
- ❑ Multiple Intensity Measures
- ❑ Vector-Based Fragilities
- ❑ Sensitivity analysis used to make BN efficient
  - Remove inconsequential hazards and dependencies
- ❑ Diagnostic inference used to fine-tune hazards/fragilities





# NARSIS RISK INTEGRATION – Step 1

## LEVEL I

## LEVEL II

### Step 1

#### **Single and multi-hazard analysis, event sequence definition and fragility modelling**

- All relevant hazards and their probability distributions from preliminary hazard analysis are listed
- Possible hazard interactions based on Liu et al. (2015) are listed and associated conditional probability distributions are defined (WP1)
- Event sequence leading to adverse event of interest is defined (WP1 through 5)
- Damage and failure states for SSC are defined
- Fragility models are defined for all SSCs for each hazard, including vector-based fragility models (WP2)

- All relevant hazards, IMs and dependencies
- Possible to use regional hazard data and generic fragility models

Based on Level I results of sensitivity analysis and diagnostic inference in BN:

- eliminate inconsequential hazards, IMs and dependencies
- Update hazard and fragility models (if possible and necessary, for e.g. site specific analyses)

### Step 2



# NARSIS RISK INTEGRATION – Step 2



## Step 2

### Screening hazard probabilities using fragilities models

- Each hazard is assessed against fragility model of every SSC to determine lower threshold magnitude of hazard that initiates damage
- Filtered hazard probability distributions are defined



## Step 3

### LEVEL I

- All relevant hazards and IMs
- Possible to use regional hazard data and generic fragility models

### LEVEL II

- Shorter list of hazards and IMs
- Updated hazard and fragility models (if available)



# NARSIS RISK INTEGRATION – Step 3



## Step 3

### Bayesian network definition

- List of random variables (nodes) – hazards, damage states, and adverse event(s) of interest at the NPP
- Directional dependence (arcs) between nodes based on hazard interactions, accident event sequence and expert judgement
- Marginal and conditional probability distributions based on filtered hazard information, fragility models, accident event sequence and expert judgement



### PREDICTIVE INFERENCE

## Step 4

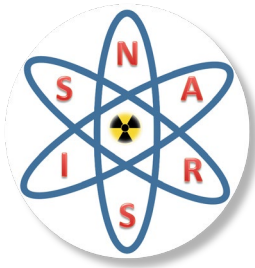
### LEVEL I

- All variables (nodes) – hazards and IMs
- All dependencies (arcs) between hazards and IMs
- Possible to use non-site-specific data to define conditional probabilities

### LEVEL II

- Shorter list of variables
- Lower no. of dependencies
- Updated probability distributions (if available)





# NARSIS RISK INTEGRATION – Step 4

↓  
PREDICTIVE INFERENCE

**Step 4**

## **Calculating probabilities of damage states and adverse end event**

- Conditional probability distributions (based on fragility models) in the BN are used to calculate:
  - (i) Probability of occurrence of damage states of SSCs
  - (ii) Probability of occurrence of adverse end event

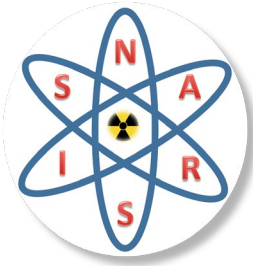
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**Step 5**

### LEVEL I

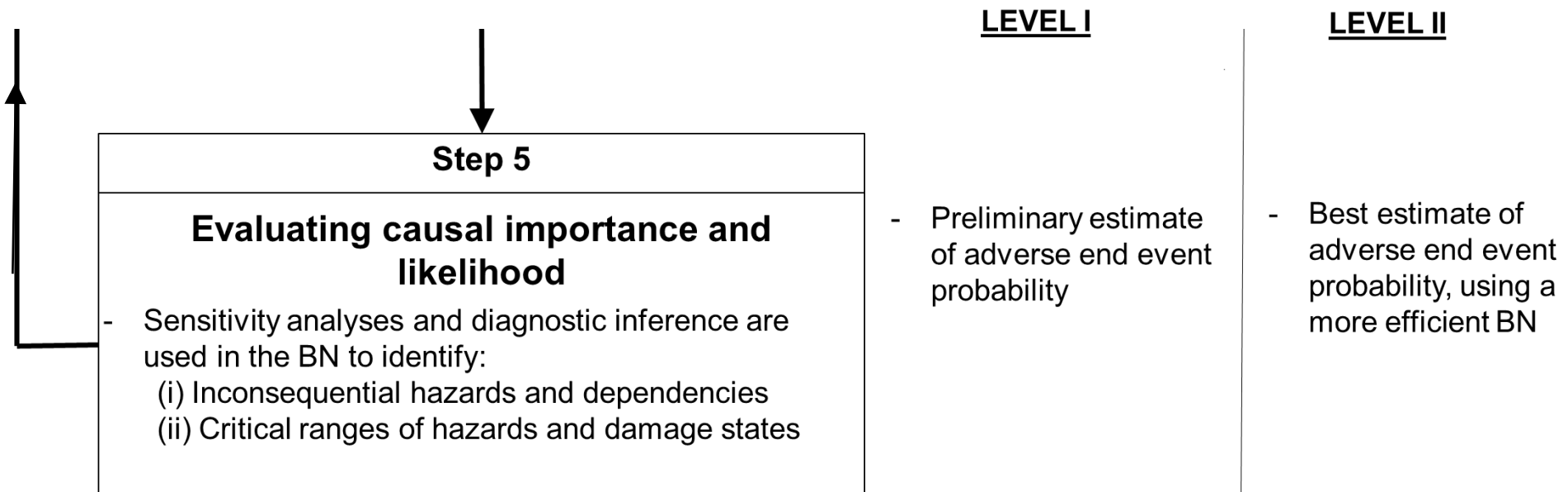
- Possible to use generic fragility models to define conditional probabilities

### LEVEL II

- Updated probability distributions (if available)

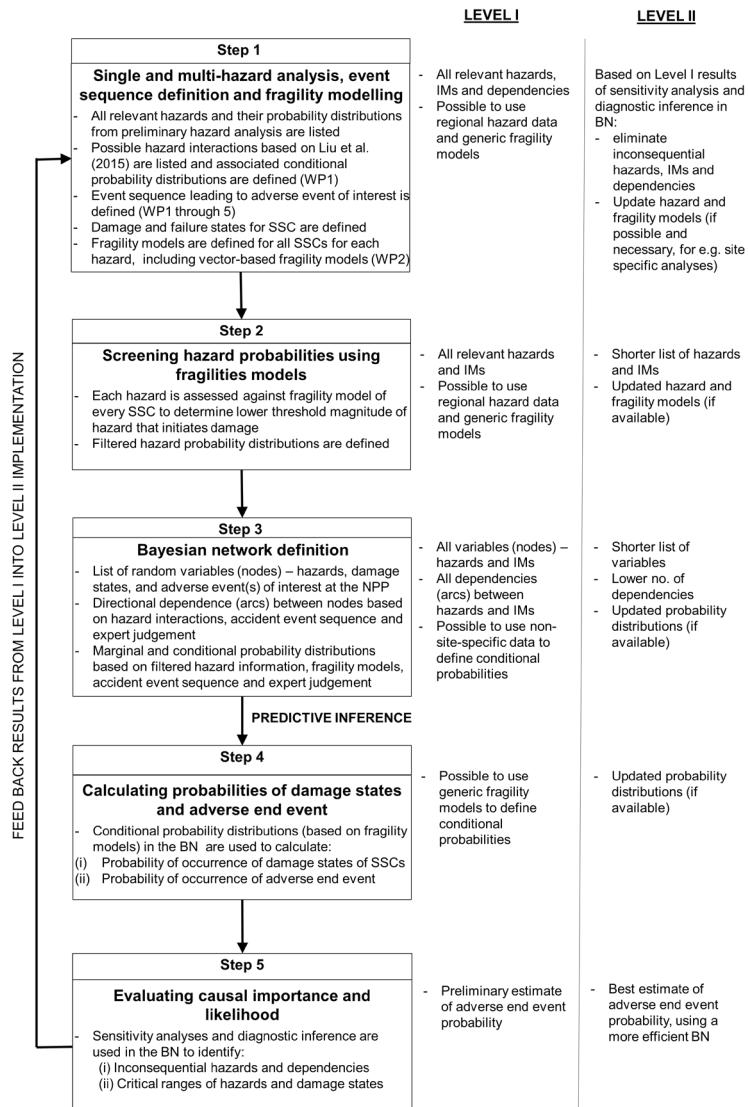


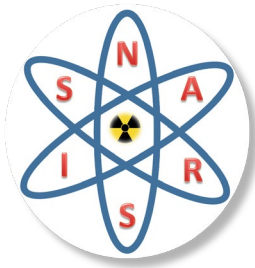
# NARSIS RISK INTEGRATION – Step 5





# NARSIS RISK INTEGRATION – Steps



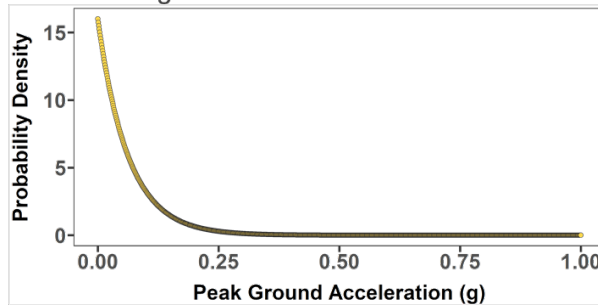


# BNs in NPP Risk

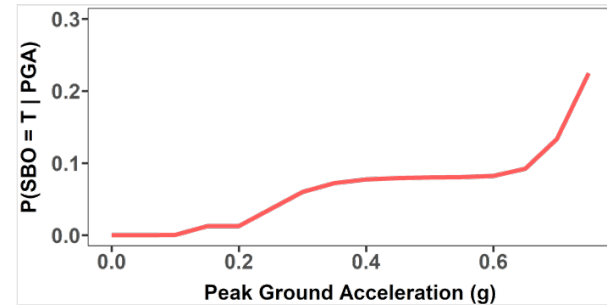
## ➤ Toy BN - Inference

**Predictive Inference** – e.g.  $P(\text{SBO} = \text{T})$  given earthquake PGA

E.g. Prior Distribution for PGA



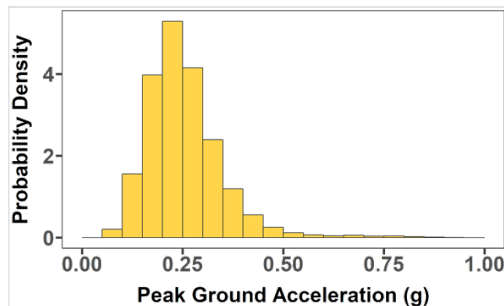
Posterior Distribution for SBO



## Diagnostic Inference

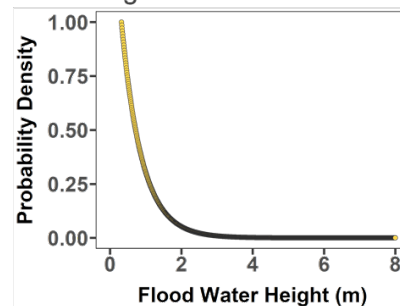
Earthquake PGA given  $\text{SBO} = \text{T}$

Posterior Distribution for PGA

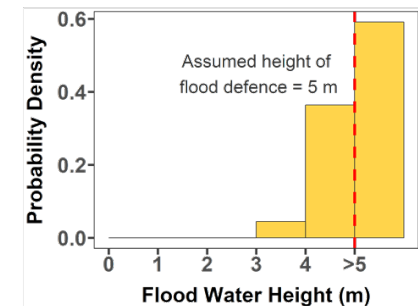


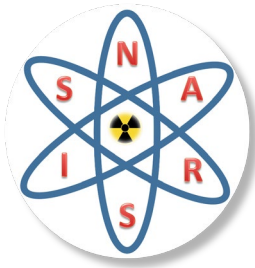
Flood water level, given  $\text{SBO} = \text{T}$ ,  $\text{PGA} = 0.05\text{g}$

E.g. Prior Distribution



Posterior Distribution

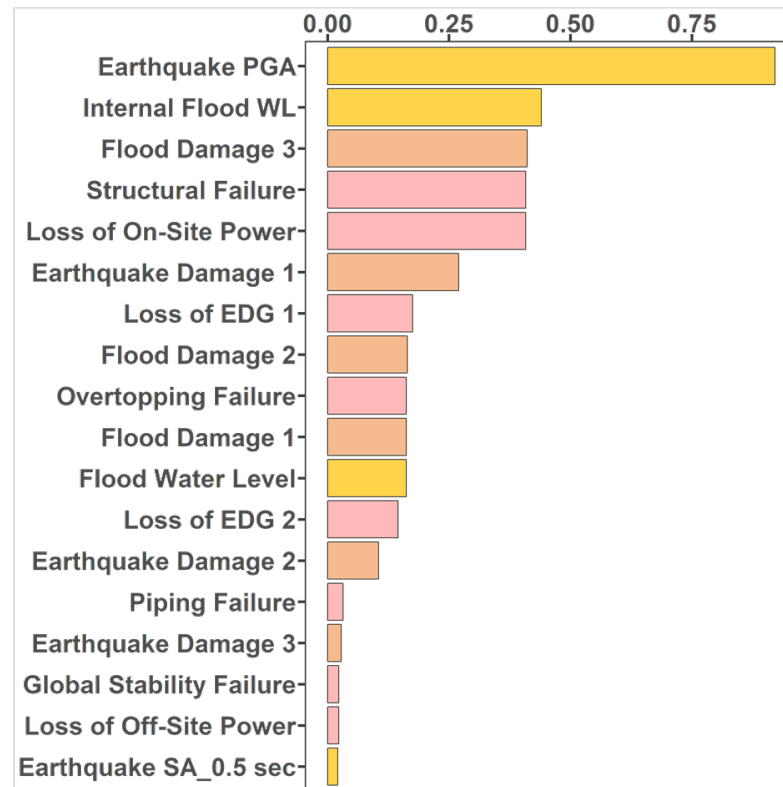




# BNs in NPP Risk

## ➤ Toy BN – Sensitivity Analysis

Change in  $P(\text{SBO} = \text{T})$  for variation between min and max values of other variables





## Summary - BNs for NPP Risk

- Some advantages of BN over conventional methods – e.g. Fault Trees:
- account for dependencies between events
- allow for bi-directional inference: causal and diagnostic
- Allow for discrete or continuous distributions to model random variables
  - ❑ The effect of discretization of continuous variables on the risk estimate can be easily studied
- Quantify uncertainty and track its propagation