



Risk & Safety in Engineering



Contents of today's lecture

- Introduction to Bayesian Probabilistic Nets (BPN)
- Causality as a support in reasoning
- Basic theory of BPN with discrete states
- Risk analysis and decision making using BPN
- Large scale risk management using Geographic Information Systems (GIS) and BPN

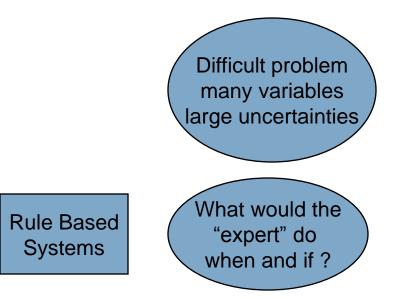
Introduction to Bayesian Probabilistic Nets (BPN)

As stated many times previously

Risk analysis supports decision making subject to uncertainty

Bayesian Probabilistic Nets (or Networks) (BPN) or Bayesian Belief Networks (BBN)

 were developed during the last decade for purposes of decision making in artificial intelligence engineering



Not possible to treat uncertainties consistently !

Inconsistent decisions – "Dutch Books"

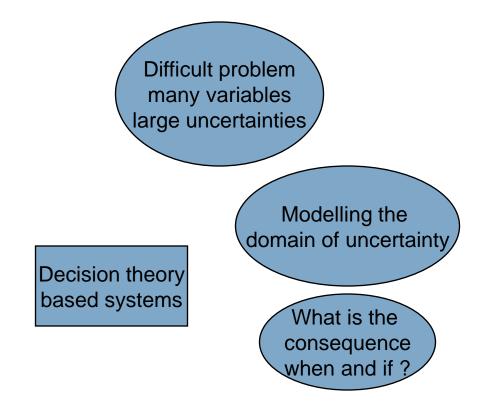
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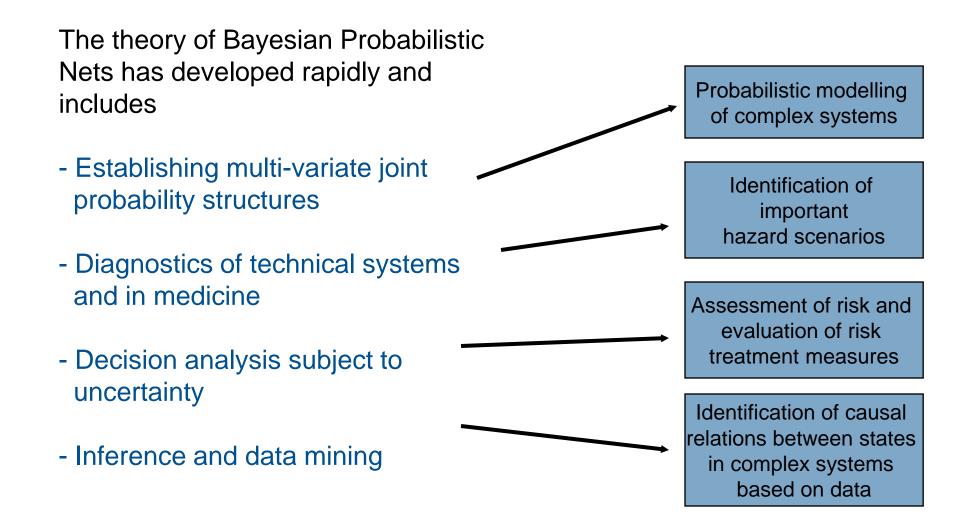
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Possible to treat uncertainties consistently !

- Supporting the expert in decision making !
- Not replacing the expert !

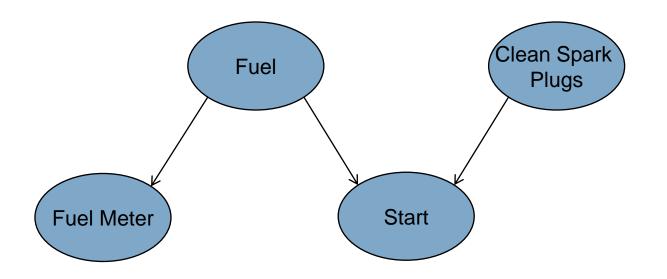
Introduction to Bayesian Probabilistic Nets (BPN)



Causality as a support in reasoning

Causality and Reasoning

Causal networks are graphical representations of causally interrelated events



Causality as a support in reasoning

Causality and Reasoning

In our daily lives we reason on the basis of causal relations

Consider the following situation:

You are the owner of two new and almost identical bridges 1 and 2 made of concrete produced on site (small factory)

- Tests performed on bridge 1 indicates that the quality of the executed concrete used in bridge 1 is bad The question is :

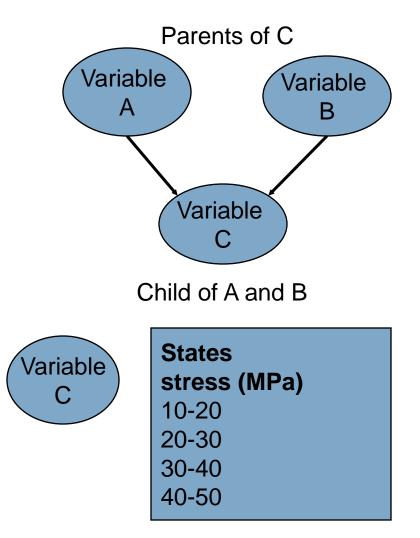
– What is the quality of the executed concrete of bridge 2?

Causality as a support in reasoning

Causal Networks and BPNs

Formally speaking :

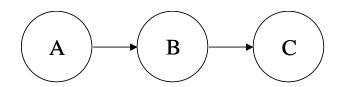
- a directed graph representing the causal interrelation between uncertain events
- interrelations expressed in terms of "family relations"
- a variable can have any number of discrete states or a continuous state space



Networks can be categorized in accordance with their configuration

For serially connected networks :

Information may be passed only if the states of the connecting variables are unknown



Serially connected network

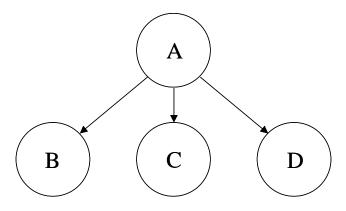
B depends on A, C depends on B

If the state of B is known with certainty variable A and variable C become independent

A and C are *d*-connected given B

For diverging networks :

Information about any of the child variables will influence the uncertainty of the states of the other children as long as the state of the variable A is unknown



Diverging network

B, C and D depend on A

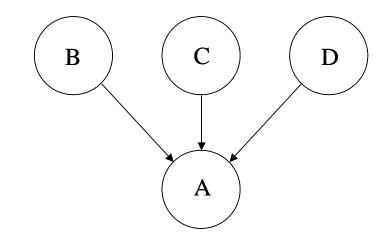
B, C and D are *d*-connected given A

For converging networks :

The parent variables remain independent as long as the state of the child variable is unknown

Given the state of the child variable or any of the parent variables, all the parent variables become dependent

This is called conditional dependence



Converging network

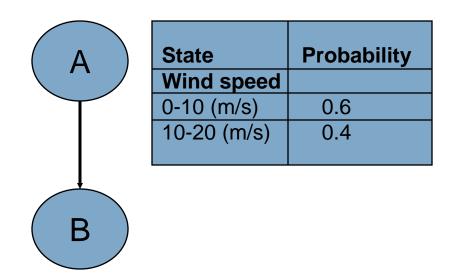
A depends on B, C and D

B, C and D are independent as long as the state of A is unknown

Given the state of A the variables B, C and D become dependent

BPNs are sometimes referred to as directed acyclic graphs (DAG).

The states of each variable is allocated a conditional probability structure.



State Wind force	Probability Wind speed (m/s)		
	0-10	<u>10-20</u>	
50-60 (kN)	0.6	0.3	
60-70 (kN)	0.4	0.7	

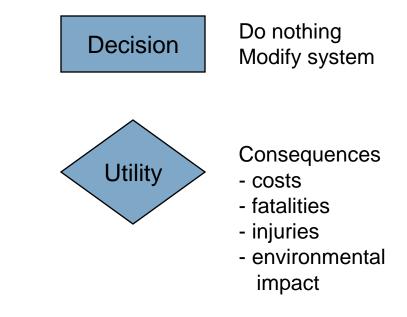
Bayesian Probabilistic Nets

BPNs may in addition to uncertain variables include

- decision nodes
- utility nodes

Decision nodes contain the various actions which may be decided

Utility nodes prescribe the consequences given the state of the variables and the decisions

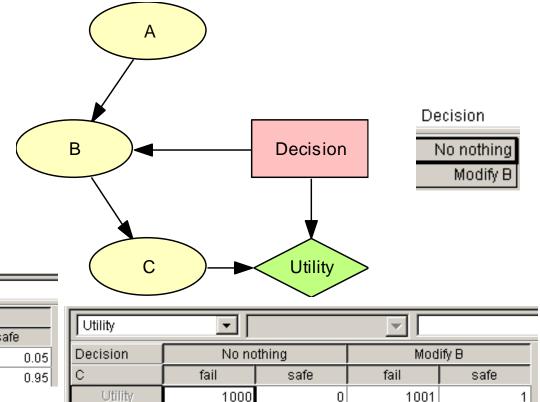


Bayesian Probabilistic Nets

The probabilities assigned to the states of a variable may be conditional on the decisions.

The utility may be given as a function of the states of the variables and the decisions.

в	- L	abelled.	•	
Decision	No nothing		Modify B	
A	fail	safe	fail	safe
fail	0.5	0.2	0.3	0.
safe	0.5	0.8	0.7	0.

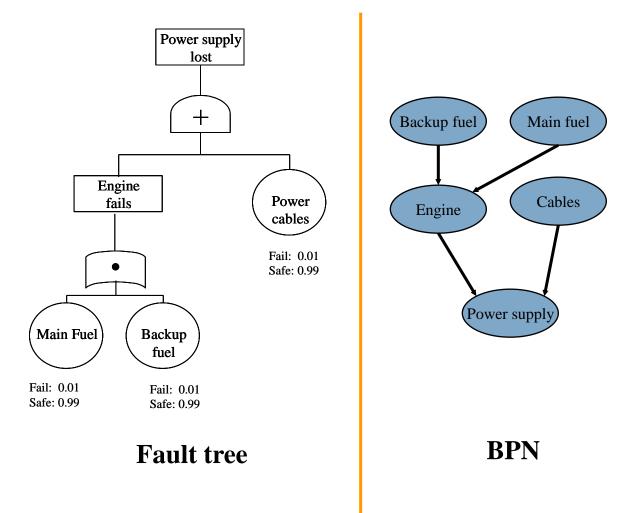


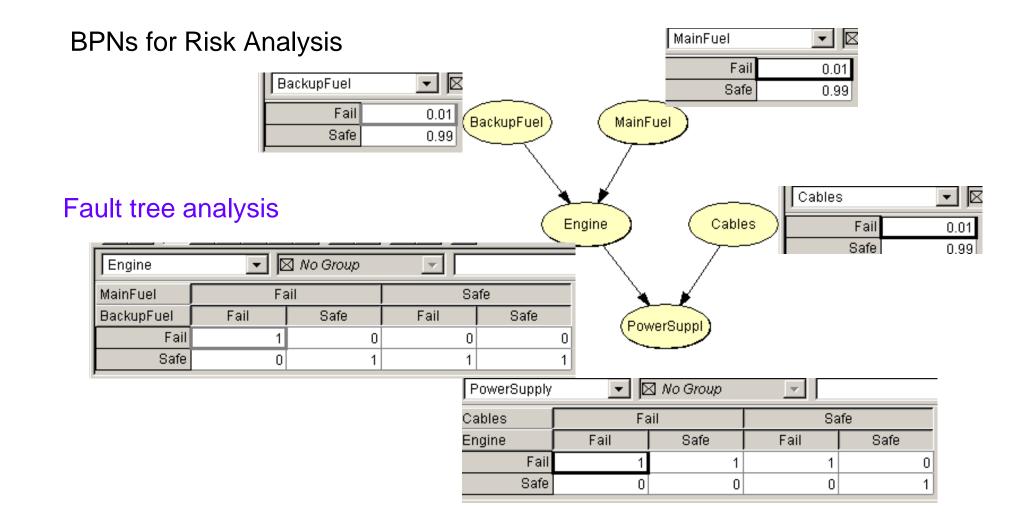
BPNs for Risk Analysis

BPNs may readily substitute fault trees, event trees, cause consequence charts and decision trees in risk analysis

No problems with common cause failures when using BPN.

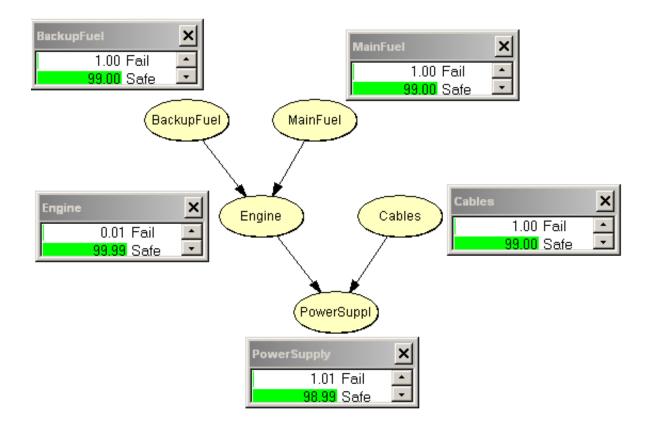
Let us consider a simple power supply example





BPNs for Risk Analysis

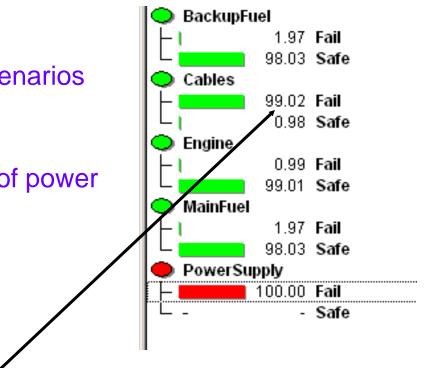
Fault tree analysis



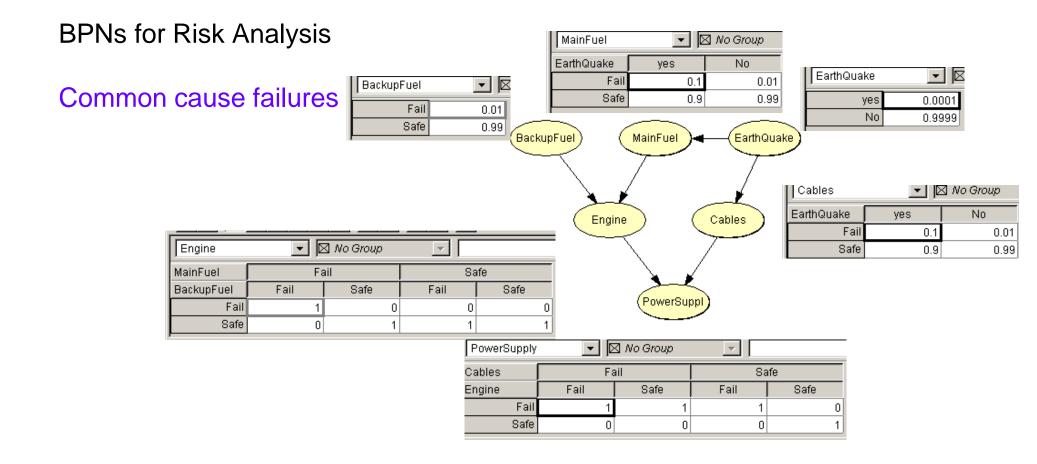
BPNs for Risk Analysis

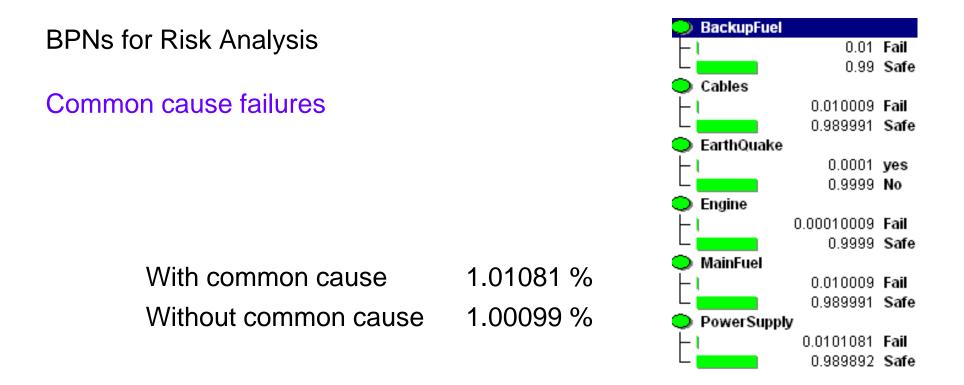
Identification of important hazard scenarios

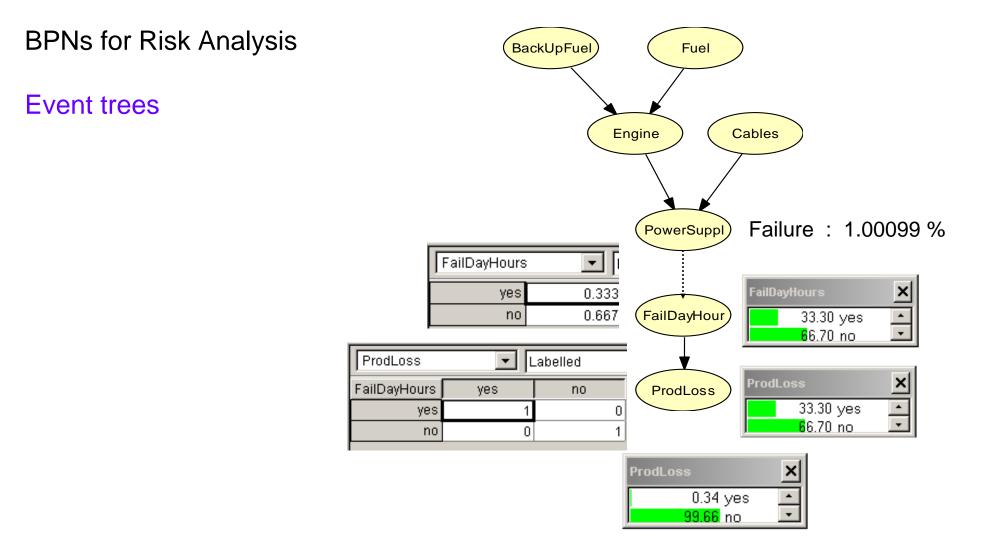
Conditioning on the event of failure of power supply



Most likely scenario is cable failure - this scenario should be detailed further in the modelling







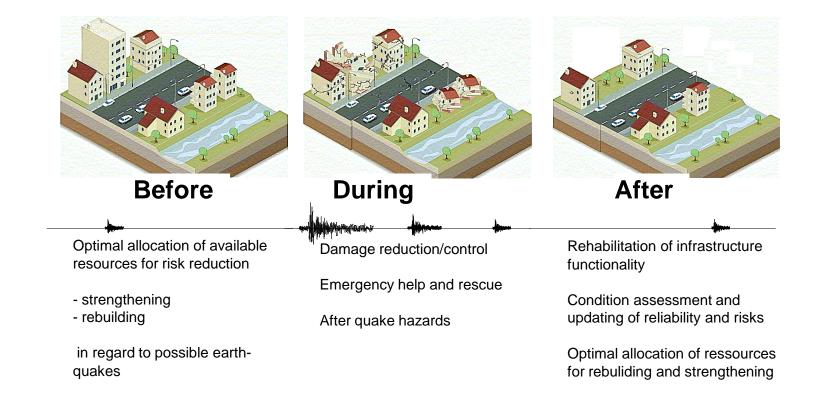
Labelled Cables -BackUpFuel Fuel Modify cables No action **Decision trees** Decision fail 0.001 0.01 safe 0.999 0.99 BackUpFuel 0.01 fail Engine Cables Decision 0.99 safe Decision 🔵 Cables 0.0055 fail Modify cables 0.9945 safe 🔵 Engine No action \vdash 0.0001 fail 0.9999 safe PowerSuppl FailDavHours FailDayHour 0.333 yes 0.667 no 🔵 Fuel 0.01 fail 0.99 safe PowerSupply ProdLoss Costs 0.00559945 fail 0.994401 safe Costs • 🔵 ProdLoss Ŧ \vdash 0.00186462 ves Modify cables No action 0.998135 no Decision Decision ProdLoss yes. no yes no 4.66267 Modify cables Utility 10001 10000 0 33.6297 No action 1

BPNs for Risk Analysis

 Risk management concerning natural hazards often involves large geographical areas

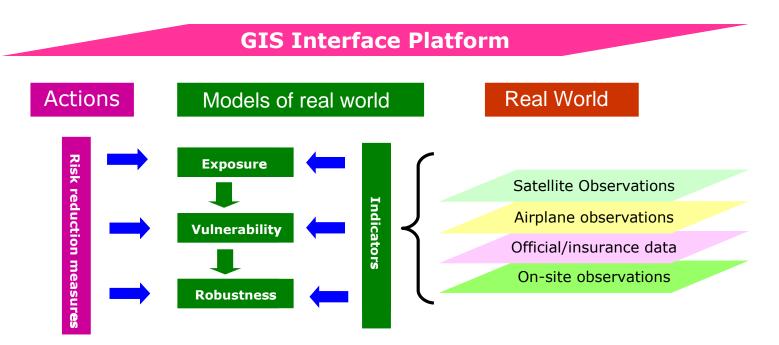


 It is important to be able to provide decision support in the situations before, during and after the occurrence of natural hazards

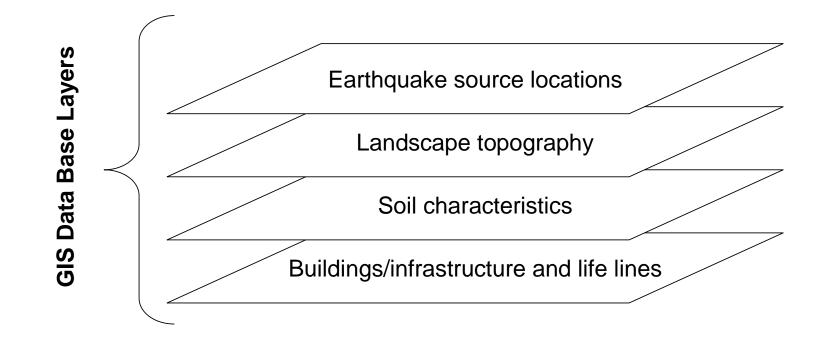


 A general framework for natural hazards risk management using GIS can be visualized as:

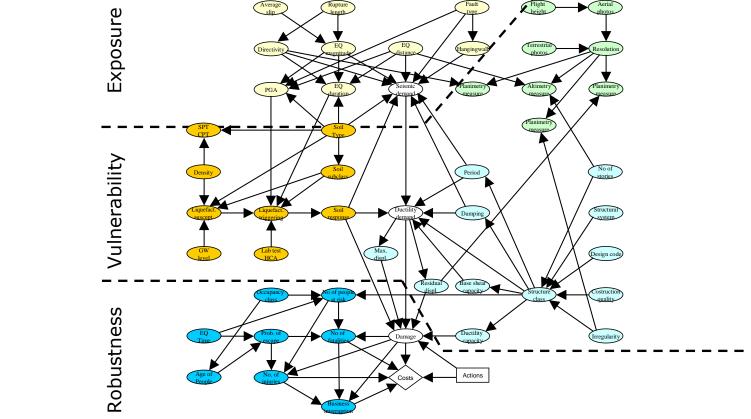
Risk Management



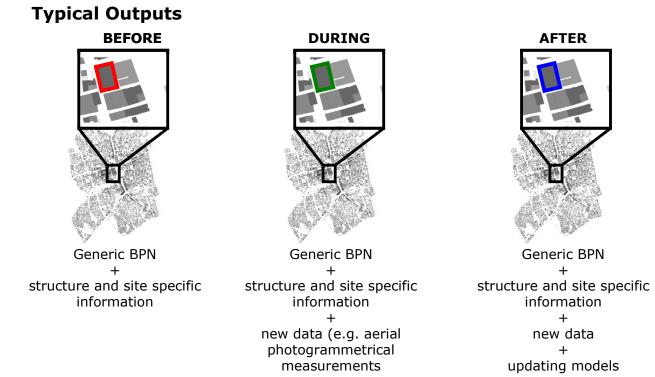
• The GIS database is important as most of the required data generally are spatially distributed for the considered system, e.g. city or region



Utilizing that indicators of exposures (hazards) and consequences (vulnerability and robustness) can efficiently be stored and managed in the GIS data base, BPN risk models may be established and linked to each asset in the considered system



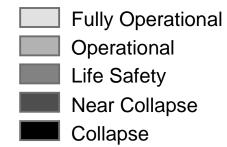
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 The generic BPN risk models linked with the GIS database facilitate the efficient risk assessment for large numbers of buildings and other assets



Damage State



 The generic BPN risk models linked with the GIS database facilitate the efficient risk assessment for large numbers of buildings and other assets



Total Risk [\$]

0 0 - 200'000 200'000 - 400'000 400'000 - 600'000 600'000 - 800'000