SimPRA: A Simulation-Based Probabilistic Risk Assessment Framework for Dynamic Systems

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SimPRA — Simulation-based Probabilistic Risk Assessment Overview





Knowledge Capture Simulation Planner Functions



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							Generate Pla	an	View Plan	
Interdependencies				a SIMPRA Planner Requirement						

Hierarchical State Space Planner Model





Simulation Model Building (Probabilistic)





Scheduler- Simulator Interactions





SimPRA planner



 Captures high level engineering knowledge to provide high level scenarios for guiding the simulation.

improves low probability high consequence scenario generation

helps simulation to converge to real probabilities faster

 Groups the scenarios to generate a complete picture of event sequences.

ESD scenario representation for risk analysts

• Provides an environment that progressively improves the high level model over time.

SimPRA

Planning Example





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Binary vs Multi-state Planner

Type of Engineering Knowledge	Captured by					
	Binary planner	Multi-state planner				
System elements and hierarchy	Structure Tree	Structure Tree				
Elements' states and operational modes	Assumed binary (work or fail)	Structure Tree				
Functionalities/ Activities/Events provided/Acted upon by elements	Functionalities for System level only	Functionality Tree				
The relationship between functionalities and sub- functionalities/Activities and events	-	Functionality Tree				
The allocation (assignment) of functionalities among components	Mapping between Functionalities and Structural Trees	Mapping between Functional and Structural Trees				
The interplay between functionalities and states of the system	State Transition Diagram	State Transition Diagrams				
The interplay between functionalities and states of the subsystems/ components	Assumed only one transition from work to fail state	State Transition Graphs				
The relationship between the functionality of the system with the state of the subsystems and components	Mapping between Functionality Tree and Structure Tree	Transition Rules				
Time dependencies	-	Transition Rules				
Conditionality of the functionalities on the state of the other elements	_	Transition Rules				
Importance of the elements to risk assessment	-	Transition Rules				
Boundary conditions	Deducted from the Mapping between Functional and Structural Trees	Qualitative Reasoning Tree 10				

Planner

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3D







Example of a Generated Plan (Event Sequence Diagram)



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Plan Updating



When a predefined number of simulation-runs are completed I. For components:

- checks every instance of a state change in the detailed scenarios
- if there is an event related to the component that is called between the changes of state, that event will be considered as the cause of the state transition for that component.
- If there is no event between state changes, then the previous event will be considered as the source of change for subsystems:

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Simulation Log		Updater Output				
#System:s1 > #Subsystem1:ss1 > #Comp1:c11 > #Comp2:c21 > !Comp1:event1 > !Comp2:event2 > #Comp2:c22 > #Comp1:c12 > #Comp1:c13 >#Subsystem1:ss2 > #System:s2	\rightarrow	1: #Comp1:c11> !Comp1:event1> #Comp1:c12 2: #Comp1:c12> !Comp1:event1> #Comp1:c13 3: #Comp2:c21> !Comp2:event2> #Comp2:c22 4: #Subsystem1:ss1> @Comp1:c13 AND @Comp2:c22> #Subsystem1:ss2 5: #System:s1> @Subsystem1:ss2> #System:s2				

Planner SimPRA Int



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Holdup tank results

			Probability		SD	# Sequences		
		case 1	case 2	case 3		case 1	case 2	case 3
	Success	9.90E-01	9.88E-01	9.92E-01	2.00E-03	103	97	73
With Plan	Dry-out 9.73E-03		1.18E-02	7.88E-03	1.96E-03	350	353	371
	Overflow	9.02E-05	1.64E-04	1.39E-04	3.75E-05	47	50	56
	Success	9.44E-01	9.53E-01	8.89E-01	3.46E-02	448	441	453
Without Plan	Dry-out	5.57E-02	3.63E-02	9.87E-02	3.19E-02	27	35	26
	Overflow	4.88E-06	1.05E-02	1.25E-02	6.71E-03	25	24	21

•Low probability – High consequence scenarios are generated more often

•Since low prob scenarios get a place holder, simulation converges faster with plan

PSAM 8 Benchmark Problem



	Component	Failure Mode	Effect						
:	PPU	Fails to start on demand	Assembly failure						
		Failure to operate							
2		Failure to shutdown on demand							
3	Ion Engine A	Fails to start on demand	Loss of redundancy						
		Failure to operate							
	Ion Engine B	Fails to start on demand	Assembly failure						
2		Failure to operate							
	Propellant Valve A	Failure to open on demand	Loss of Ion Engine A						
		Failure to close on demand	System failure						
_		External leakage							
$\overline{2}$	Propellant Valve B	Failure to open on demand	Loss of Ion Engine B						
2		Failure to close on demand	System failure						
5		External leakage							
	Propellant tank	External leakage	System failure						
	Propellant distribution lines	External leakage	System failure						



Group Si ze	Group Conditional Failure Probability [%]
2	8.0
3	4.0
4	2.0
5	1.0 18









SimPRA Simulation

(500 runs)

Quantitative biasing (biased sampling) Qualitative biasing (planning) Dynamic Intelligent biasing (e.g., entropy based) Monte Carlo Simulation (10000 runs)

Primary Contributions



- A new method for capturing different types of engineering knowledge to automatically generate high level dynamic risk scenarios and
 - guide DPRA simulation
 - supply classical PRA techniques with generalized event sequence diagrams
 - a way to summarize simulation results for risk management
- As an integral element within the SimPRA framework, the planner has been shown to improve convergence and coverage of risk scenarios
- Computer implementation



Benchmark problem results

Name	Exact/ Approx. solution	Binary/ Multi state	Expandable	Low Prob. High Cons. Scenarios	Complex Systems	Common Cause	Demand- Based/ Time- Based	Model Complexity	Problem Solved
MC	Approximate	Binary but multi-state is also possible	Not known	No	Yes	Yes	Both	High	Yes [E-1]
DFM	Analytical	Binary	Yes	Yes	No	Not shown	Both	Can't get too complex	No
DFT	Exact	Binary	Yes	Yes	No	Not shown	Time based only	Not easy to develop	Yes but way too far of other solutions [E-13]
AO-MC	Approximate	Multi state	Yes but only horizontally	No	Yes	Yes	Both	Complex	Yes[E-1]
SAPHIRE	Exact	Multi state	In some cases in a static form	Yes	No	Yes	Both	Very hard to model	No
FT/ET/Markov	Analytical approach, approximate solutions	Multi state	No	Yes	No	Yes	Both	Not easy to develop	Yes but out of range solutions [E-3]
SimPRA	Approximate	Multi state	Yes, both horizontally and vertically	Yes	Yes	Yes	Both	Complex	Yes [E-1]
DES (TIGER)	Approximate	Multi state	Not known	Not known	Yes	Yes with difficulty	Time based. Demand based with difficulty	Not too complex	Yes [E-1] 21

SimP

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Evaluation

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