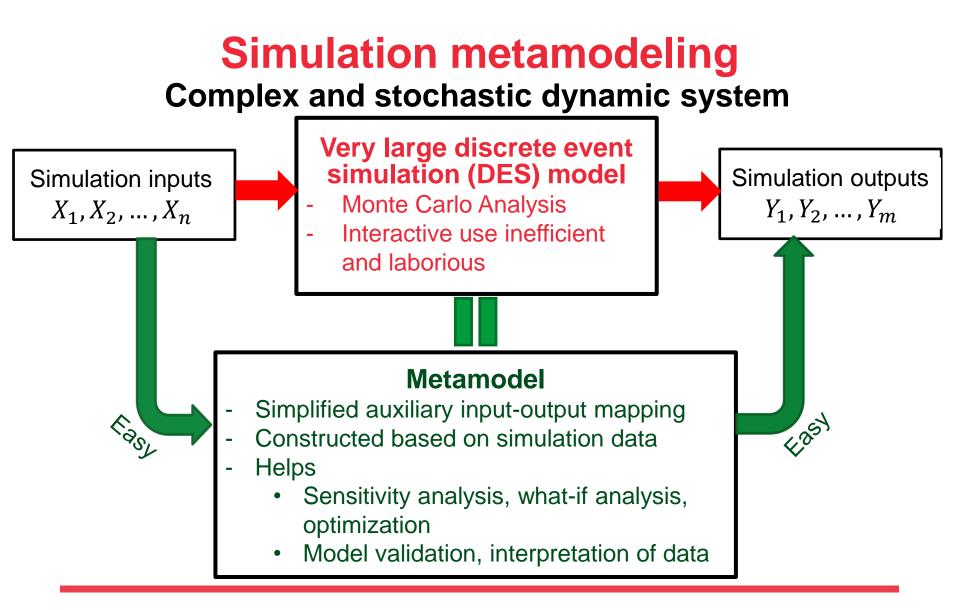




Simulation Metamodeling Using Dynamic Bayesian Networks with Multiple Time Scales

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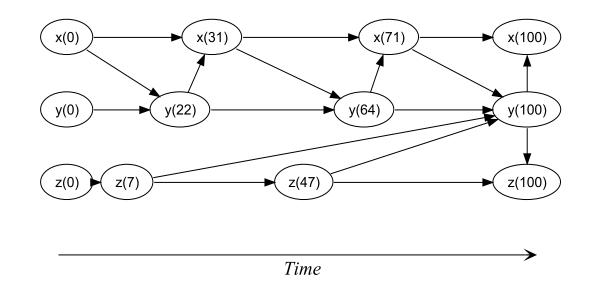






Dynamic Bayesian Network (DBN)

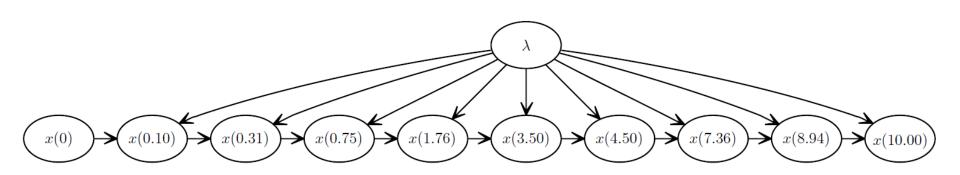
- Joint probability distribution of a sequence of discrete valued random variables
- Dynamic variables
 - Nodes
 - Specific time instants
- Dependencies
 - Arcs
 - Conditional probability tables



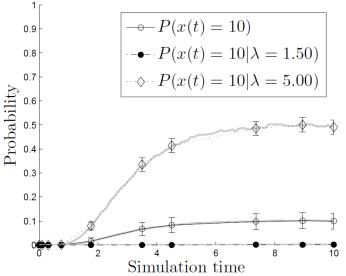




DBN metamodels



- Variables of DES models and DBNs equated
- Joint distribution of all variables
- Key time instants for state variables
- Construction is challenging







Construction of DBN Metamodels

- 1) Selection of variables
- 2) Collecting simulation data
- 3) Optimal selection of time instants
- 4) Determination of network structure
- 5) Estimation of probability tables
- 6) Inclusion of simulation parameters
- 7) Validation

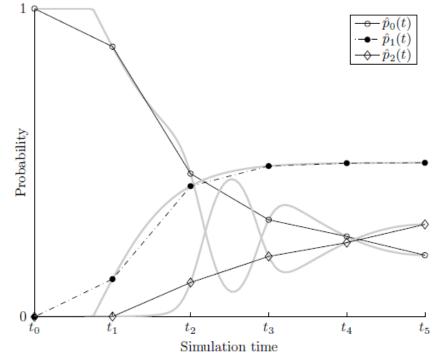




Optimal Selection of Time Instants

- Probability curves $p_i(t) = P(x(t) = i)$ estimated from simulation data
- DBN gives probabilities at discrete times $T = \{t_0, \dots, t_f\}$
- Piecewise linear interpolation

$$\hat{p}_{j}(t) = p_{j}(t_{-}) + \frac{p_{j}(t_{+}) - p_{j}(t_{-})}{t_{+} - t_{-}}$$
$$t_{-} = \max\{v \in T | v \le t\}$$
$$t_{+} = \min\{v \in T | v \ge t\}$$







Optimization Problem

- Separately for each state variable by solving an optimization problem
 - Piecewise linear approximation sought for the time evolution of probabilities, i.e., a curve fitting problem
 - Minimize error of approximation
 - Break points of the approximation selected as the time instants
- Solved by an algorithm based on dynamic programming
 - Initial dense discretization of the timeline
 - Optimal subsets of time instants determined
 - Different solutions for different numbers of time instants





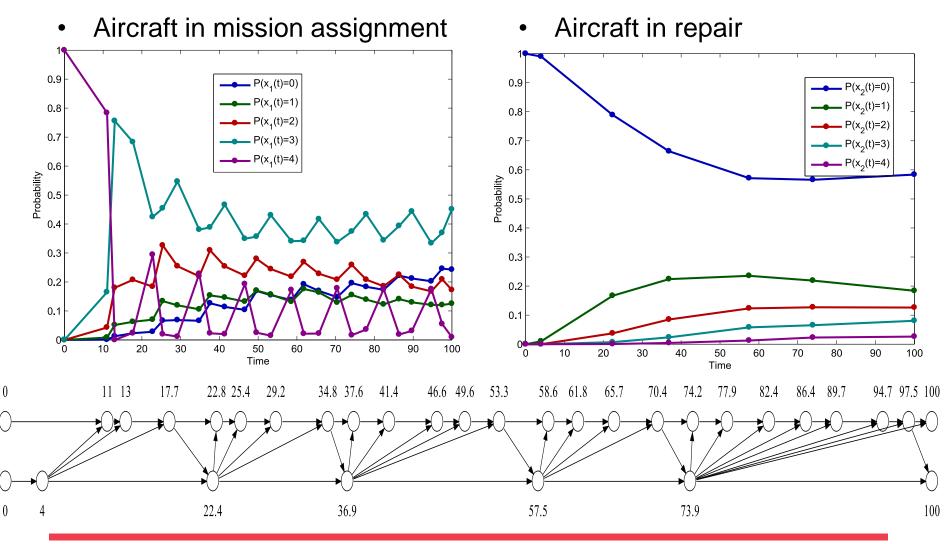
Example: Simulated Operation of Air Base

- Aircraft go through a four stage cycle
 - Mission assignment
 - Mission execution
 - Repair (only if damaged)
 - Service
- Two types of missions
 - Regularly scheduled patrol missions
 - Combat missions, whose generation is random and dependent on an input variable
- Random duration of mission execution, repair, and service





State Variables

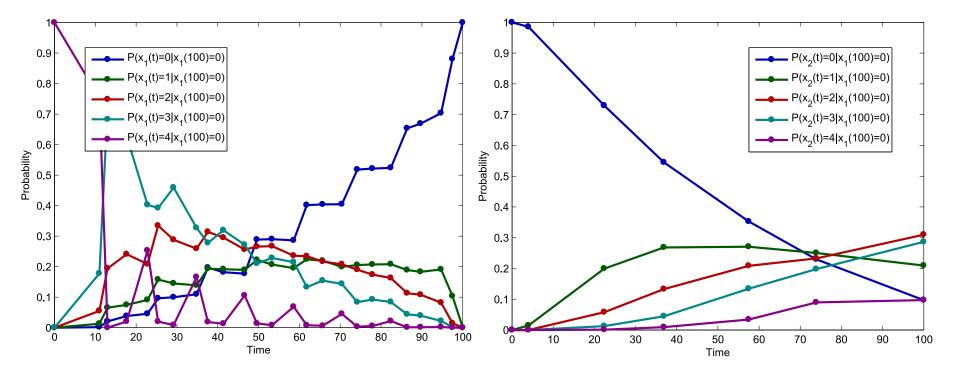






What-if Analysis

• What if no aircraft are available for mission assignment at time instant 100?



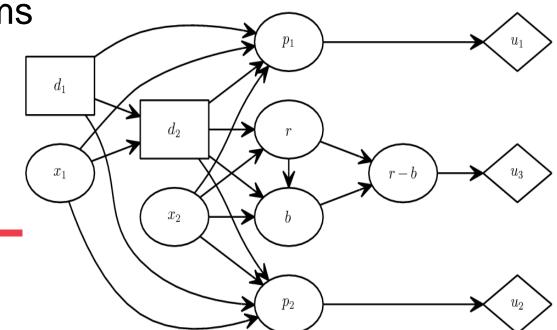




Conclusions

- Dynamic Bayesian networks in simulation metamodeling
 - Time evolution of simulation
 - Simulation parameters as random variables
 - What-if analysis
- Simulation metamodeling using (dynamic) influence diagrams
 - Decision making problems
 - Optimal decision suggestions





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