

Acyclic Phase-Type Distributions in Fault Trees

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Computer and Communication Systems

Reliability analysis

What is the likelihood of *system failure*?
given
the likelihood of *component failure*?

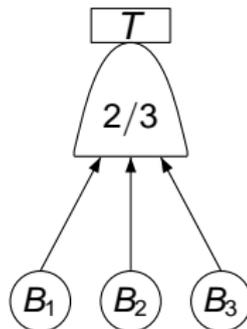
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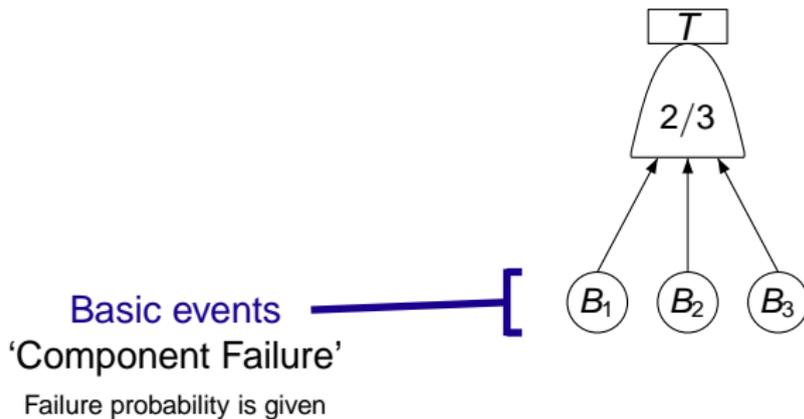
Outline

- 1 Theory
 - Fault Trees
 - Phase-Type distributions
- 2 Practice
 - Dynamic Fault Trees
 - Case study
- 3 Conclusion

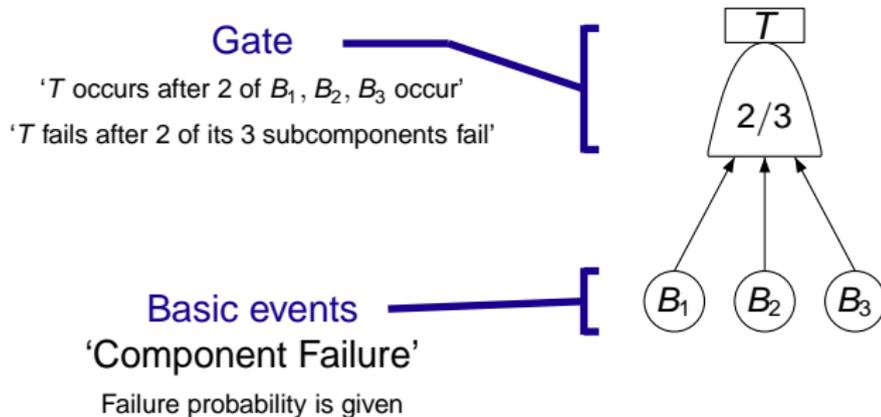
Fault Trees - from component failure to system failure



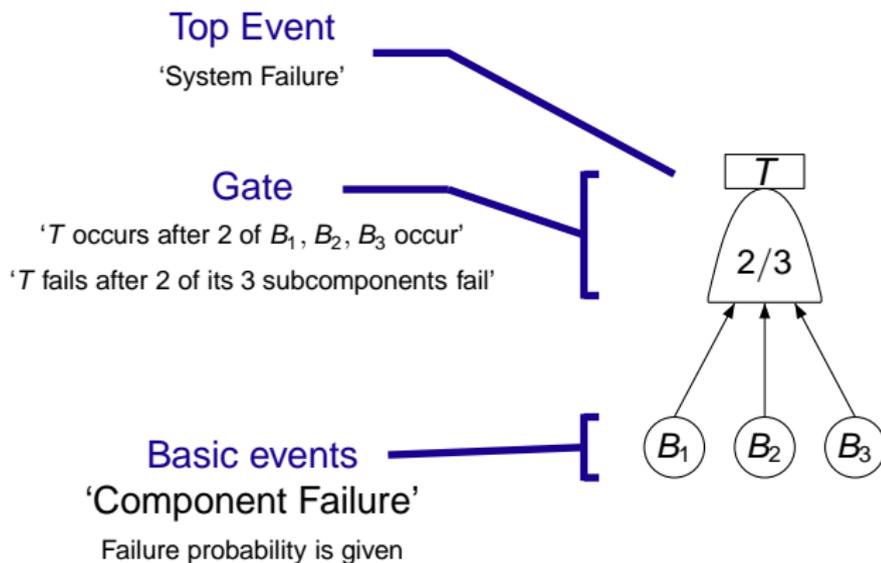
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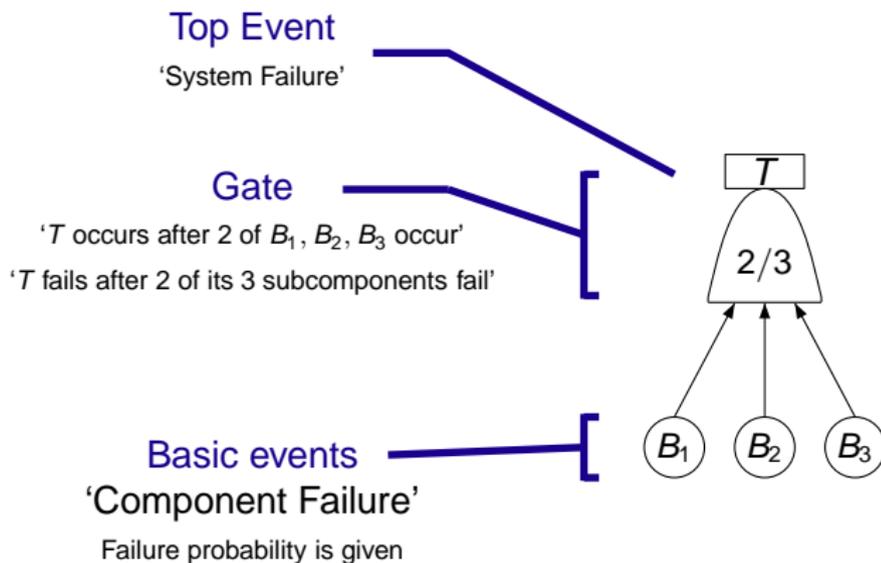
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But actually...

A Fault Tree is a Boolean Function

Some terminology

- The state of the basic events is a random boolean vector $\vec{B} = (B_1, \dots, B_n)$,
- The fault tree is a function f from $\{0, 1\}^n$ to $\{0, 1\}$. And now,

$$P(T = 1) = P(f(\vec{B}) = 1).$$

- This problem can be solved efficiently with binary decision diagrams.
- We only consider *coherent* fault trees where events are irrevocable.

Truth table of f

B_1	B_2	B_3	T
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Fault Trees with Time

Adding Time...

- State of BEs at time t is a stochastic process $\vec{B}^{(t)} = (B_1^{(t)}, \dots, B_n^{(t)})$,
- $P(B_1^{(t)} = 1)$ is the probability that event B_1 has occurred on or before time-point t , and
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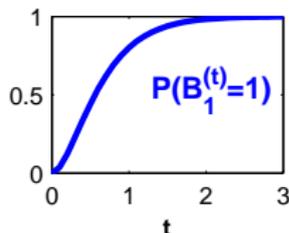
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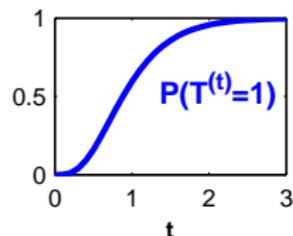
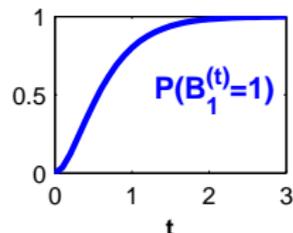
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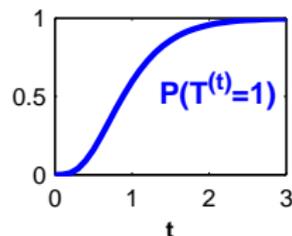
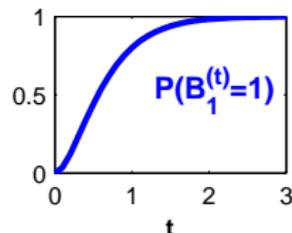
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PH distribution overview

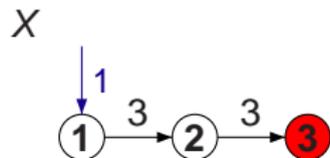
Properties

- A PH-distribution is represented by a CTMC with a **single absorbing state**,
- Matrix characterization,
- For a random variable Z PH-distributed with representation X we have,

$$P(Z \leq t) = P(X^{(t)} = 3) = \vec{\alpha} e^{Qt} \vec{\omega}.$$

- Infinitely many different representations,
- Acyclic PH-distributions (APH),
- For FTs:

$$P(B_1^{(t)} = 1) = P(X^{(t)} = 3).$$



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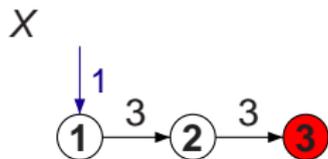
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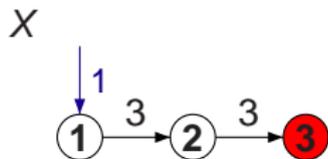
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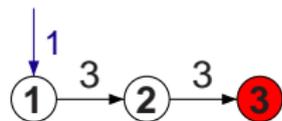
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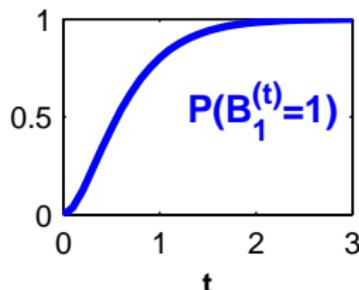
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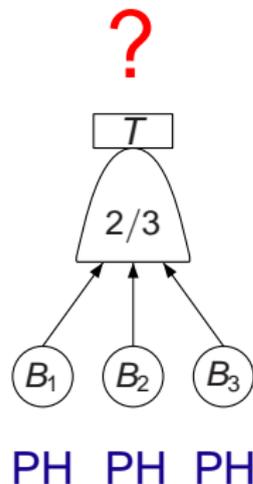
FT and PH

Theorem

The top event of a coherent fault tree with PH-distributed basic events is itself PH-distributed.

Corollary

The top event of a coherent fault tree with APH-distributed basic events is itself APH-distributed.



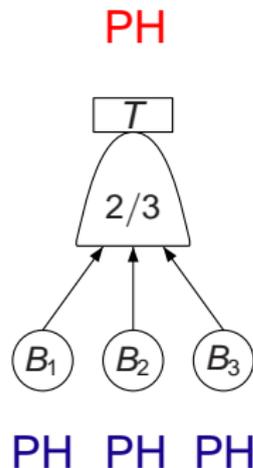
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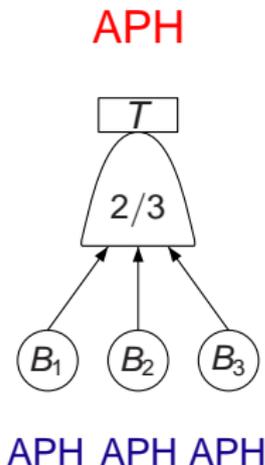
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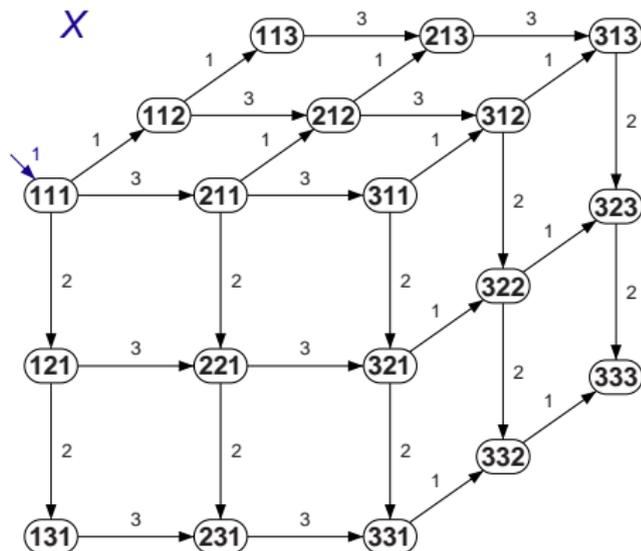
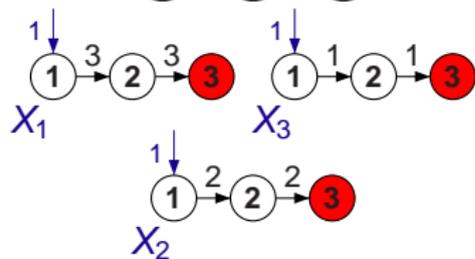
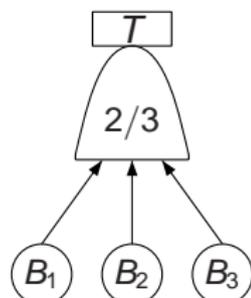


Proof by Construction

Constructing a representation for the top event

- For a coherent FT with n basic events,
- Parallel composition of representations: X
Initial distribution $\vec{\alpha} = \vec{\alpha}_1 \otimes \dots \otimes \vec{\alpha}_n$
Generator matrix $\mathbf{Q} = \mathbf{Q}_1 \oplus \dots \oplus \mathbf{Q}_n$.
- Mark occurrence of basic events. Per state a boolean vector $\vec{b} \in \{0, 1\}^n$,
- Group states by $f(\vec{b})$: Two sets S_0 and S_1 ,
- Collapse S_1 to a single state (Note: S_1 is absorbing),
- The resulting CTMC Y represents the PH distribution of the top event of the FT.

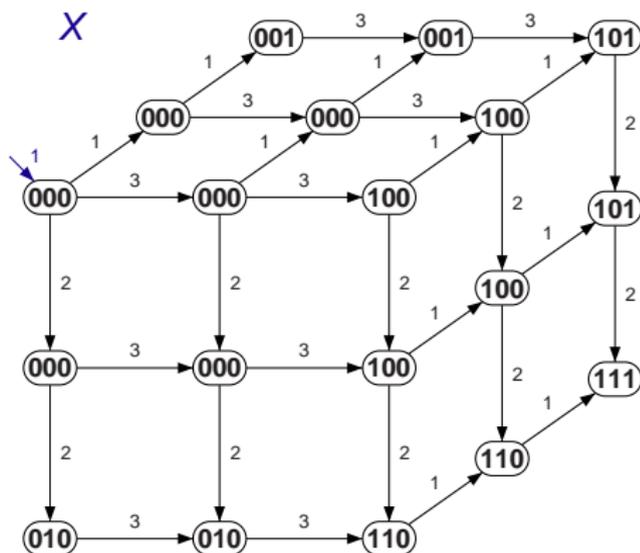
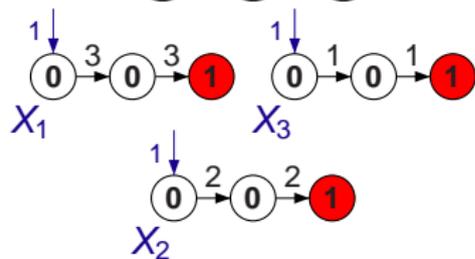
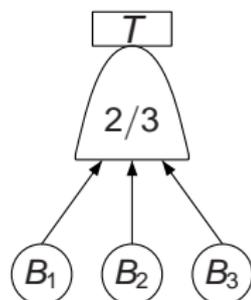
Construction example: parallel composition



The connection

$$P(X_1^{(t)} = 3 \wedge X_2^{(t)} = 1 \wedge X_3^{(t)} = 2) = P(X^{(t)} = 312)$$

Construction examples: From states to events

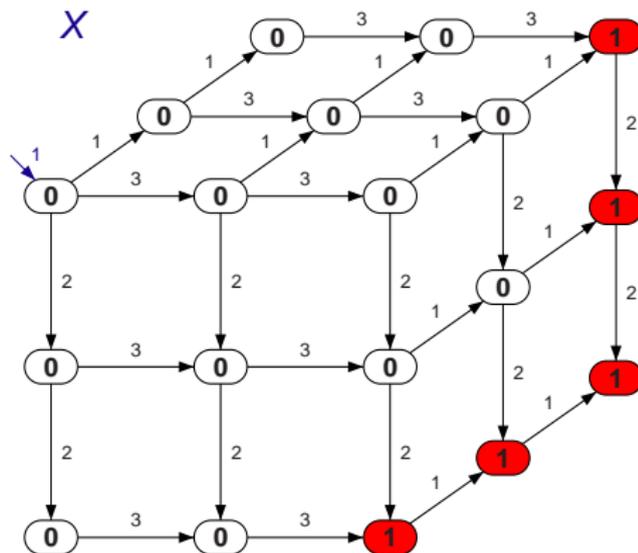
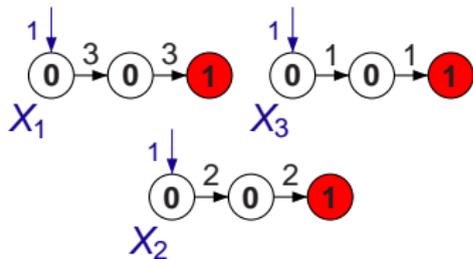


The connection

$$P(\vec{B}^{(t)} = (100)) = P(X_1^{(t)} \in S_1 \wedge X_2^{(t)} \in S_0 \wedge X_3^{(t)} \in S_0) = P(X^{(t)} \in S_{100})$$

Construction: Apply function f

B_1	B_2	B_3	T
0	0	0	0
0	0	1	0
0	1	0	0
1	0	0	0
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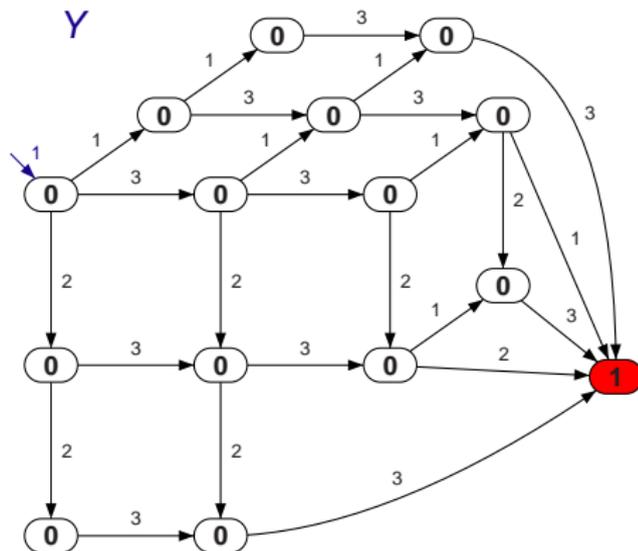
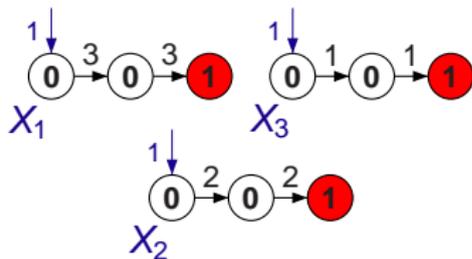


The connection

$$P(T^{(t)} = 1) = P(f(\vec{B}^{(t)}) = 1) = P(X^{(t)} \in S_1)$$

Construction: Collapse set '1'

B_1	B_2	B_3	T
0	0	0	0
0	0	1	0
0	1	0	0
1	0	0	0
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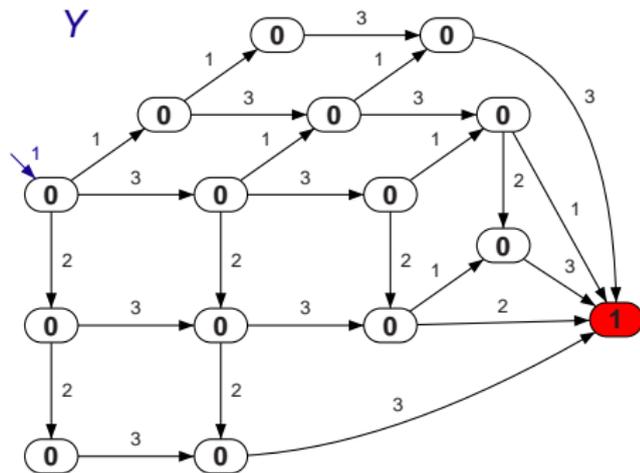
The connection

$$P(T^{(t)} = 1) = P(X^{(t)} \in S_1) = P(Y^{(t)} = 1)$$

Minimal Representation for APH

A better representation

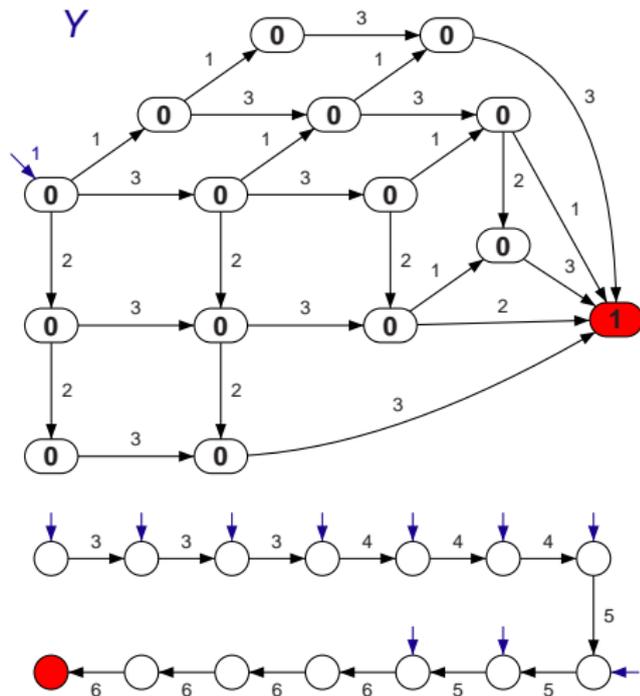
- Y has 21 states and 45 transitions,
- Smallest representation: 14 states and 13 transitions,
- For APH: find smallest representation with APHMIN.



Minimal Representation for APH

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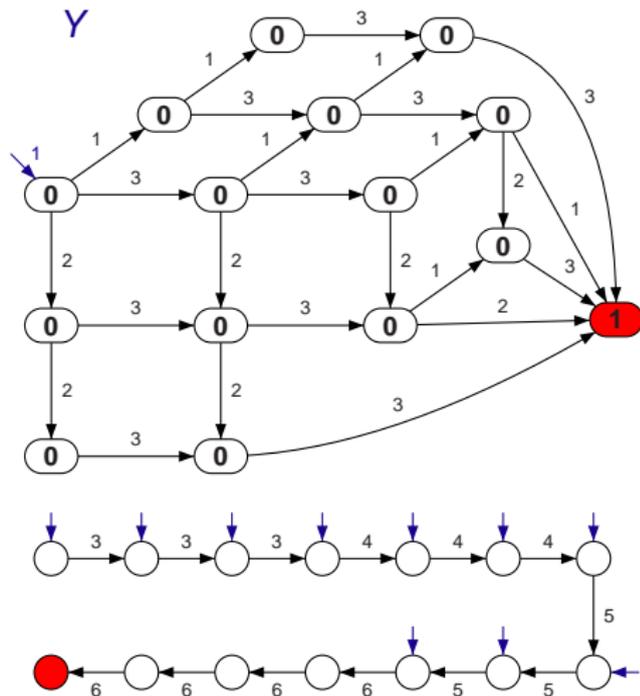
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APHMIN Sketch

Computing the 'minimal' representation

- 1 Convert the APH representation to bidiagonal form,
- 2 Consider the Laplace-Stieltjens transform,
- 3 For each state:
 - 1 Check for states that are linear combinations of other states,
 - 2 Compute a new initial 'distribution',
 - 3 Check if the new initial 'distribution' is a distribution.

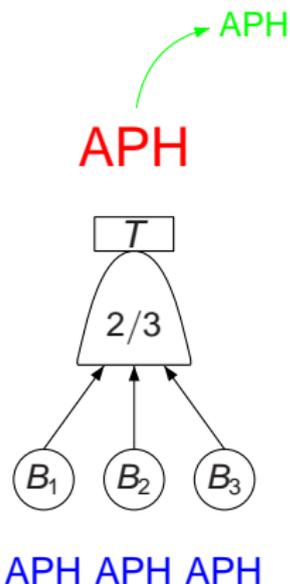
So far

We have seen...

- Describe BEs with APH distributions,
- Top event is also APH distributed, and
- APH representations can be minimized.

But...

- Solving a FT is anyway easy!
- Perhaps FTs are too simple...



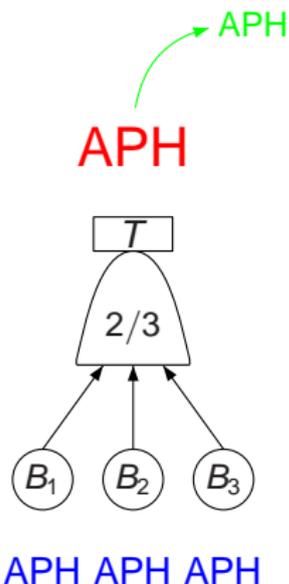
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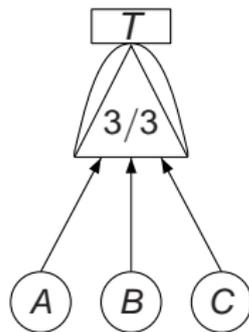
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Bringing order to FTs

Dynamic Fault Trees

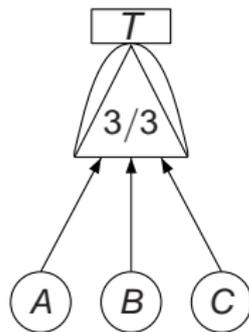
- Adds three gates to FTs,
- Example: T only occurs if A , B and C happen in the correct order,
- A DFT is not a boolean function!
- We must construct a representation of the distribution of event T .
- FT construction does not work.



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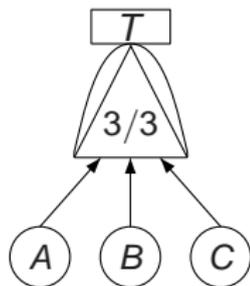


Compositional Aggregation

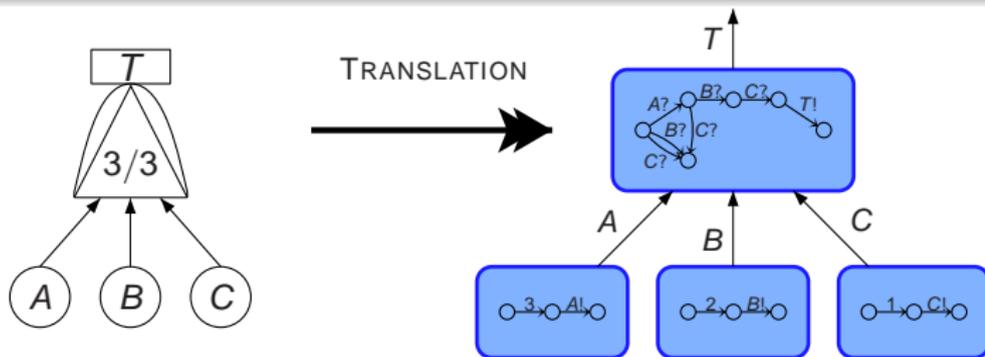
Generating the state space step by step

- 1 Translate syntactic elements to interactive models,
- 2 Select a subset of interactive models,
- 3 Compose them,
- 4 Minimize the result,
- 5 More than one model left: goto 2.

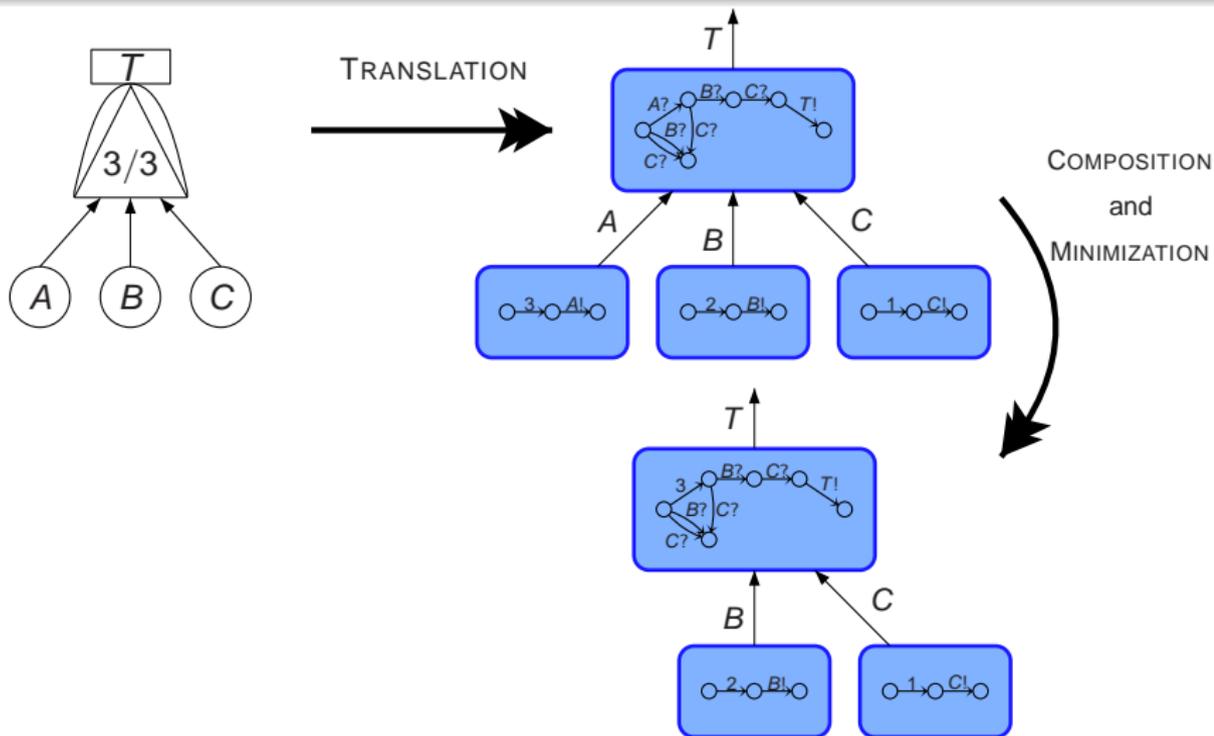
Constructing CTMC with CORAL



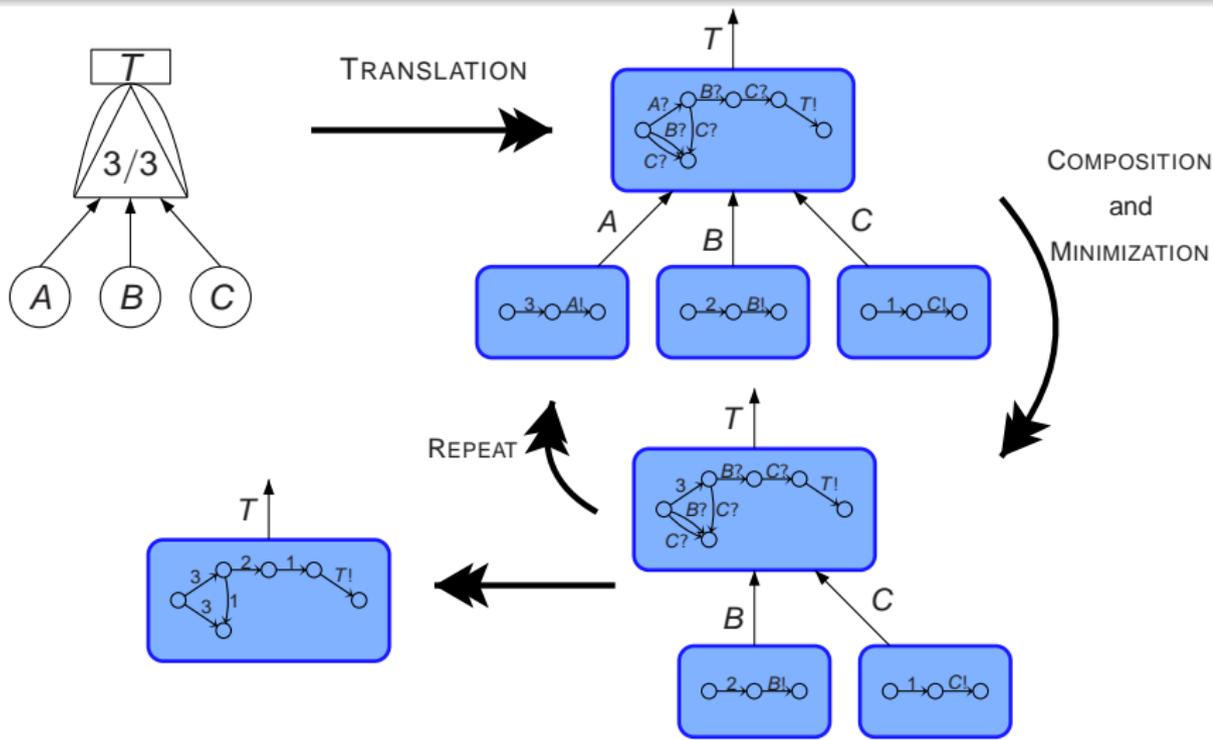
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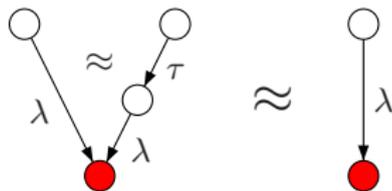


Constructing CTMC with CORAL



How to minimize?

Weak bisimulation

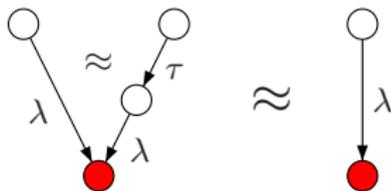


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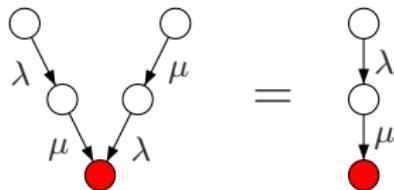
- Equality = same observable transitions,
- Eliminate equivalent states,
- For CTMCs and IOIMCs,
- Partition refinement algorithm.

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APHMIN



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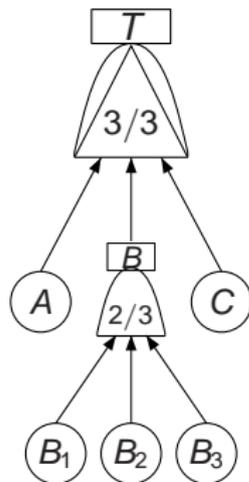
Properties

- Considers Laplace-Stieltjes transform,
- Eliminate states that are linear combinations of other states,
- For APH representations,
- Weaker than weak!

Improving Compositional Aggregation

Which minimization?

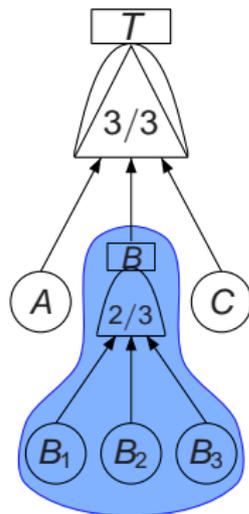
- Compositional aggregation uses minimization,
- For DFTs: weak bisimulation minimization,
- Can we use APHMIN instead?
- For FT-subtrees we can!
- 14 states instead of 21 states.



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Case Study

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- FTTP case study (20 basic events, 21 gates),
- Three variants with 1,2, or 3 FT subtrees,
- Compare normal CA (CORAL) with enhanced CA (APHMIN),
- For FT subtrees: 21 states (CORAL) vs. 14 states (APHMIN).

#	Tool	States	Transitions	Time (s)	Unreliability
1	CORAL	1,672	12,303	10.37	$1.13 \cdot 10^{-7}$
	APHMIN	1,119	7,410	10.42	$1.13 \cdot 10^{-7}$
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- W.E. Vesely, F.F. Goldberg, N.H. Roberts and D.F. Haasl, *Fault Tree Handbook*. United States Nuclear Regulatory Commission, 1981, vol. (NUREG-0492).

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