

### Prioritizing Failure Events in Fault Tree Analysis Using Interval-valued Probability Estimates

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#### **Uncertainty in probability estimates**

- Risk importance measures help prioritize failure events
  - E.g. Fussell-Vesely, Birnbaum
- These are typically computed using crisp probabilities
- Probability estimates can be uncertain
  - Statistical data, simulation models, expert opinions
  - If component fails iid 20 times out of 1000 → 95 % confidence interval of probability is [0.012, 0.031] (Pearson-Klopper method)
- What is the impact of this uncertainty?



# Prioritization of failure events with interval probabilities

- Interval-probabilities define a set of values that the "true" probability can have
  - We use confidence intervals
- A dominates B iff
  - Risk importance of A is at least as great as the risk importance of B for all probabilities within the intervals, and
  - Risk importance of A is strictly greater than the risk importance of B for some probabilities within the interval
- Dominance relation determine an incomplete ordering of the failure events
  - Extends prioritization based in crisp values





#### **Illustrative example (1/2)**

**System** 



Traditional

 $p_j = 0.02 \forall j$ 

- Comp FV
- 1 5,00E-01
- 2 5,00E-01
- 3 5,00E-01
- 4 5,00E-01
- 5 1,96E-02
- 6 1,96E-02
- 7 9,80E-01

#### Interval-probability

Fussell-Vesely dominance  $0.01 \le p_i \le 0.03 \forall j$ 



With a wider interval  $0.01 \le p_j \le 0.04 \forall j$  no component dominates the other





### Illustrative example (2/2)

Interval-probability  $0.01 \le p_i \le 0.03 \forall j$ 

Traditional

 $p_i = 0.02 \forall j$ Comp Birnbaum FV 1 5,00E-01 8,16E-04 5,00E-01 8,16E-04 2 5,00E-01 8,16E-04 3 5,00E-01 8,16E-04 4 5 1,96E-02 6,34E-05 1,96E-02 6,34E-05 6 9,80E-01 1,57E-03 7



With a wider interval  $0.01 \le p_j \le 0.04 \forall j$  no component dominates the other







#### Application on the Residual Heat Removal System (RHRS)

- Medium sized fault tree
  - 31 basic events (BEs)
  - 147 minimal cut sets of 1-3 BEs
  - Each component typically belongs to 1-13 of the MCS
- Probability interval equals the 90 % confidence interval
- Dominances computed using our algorithm
  - Implemented in Mathematica

7



Part of the RHRS fault tree



<mark>Systems</mark> Analysis Laboratory

## RHRS Fussell-Vesely dominances

- Basic events labeled by their conventional FV risk importance ranks
- Dominances define an incomplete order
  - Eg. 5 and 12 are equal in the sense that neither dominates the other, even if with crisp probabilities

 $FV_5 = 5.86E-02$  and  $FV_{12}=5.21E-03$ 



#### Lessons learned from the RHRS case

- Our method computatinally viable
  - RHRS case model (medium size) solved under a minute
- Model data readily available
  - MCS, probability confidence intervals from standard fault tree analysis
- The dominance graph gives an overview of the sensitivity of the priorities
  - In RHRS case relatively few dominances → uncertainty has large impact in priorities
  - Transparent, conservative and justifiable





#### **Further research**

- Derive lower and upper bounds for the relative risk ranking of components
  - Which rankings are attainable?(cf. Salo and Punkka, 2011)
- Apply current method for prioritizing and ranking minimal cut sets
- Apply methods to large fault trees





#### References

- Salo, A., J. Keisler and A. Morton (eds) (2011).
  Portfolio Decision Analysis: Improved Methods for Resource Allocation, Springer, New York
- Salo, A. and A. Punkka (2011). Ranking Intervals and Dominance Relations for Ratio-based Efficiency Analysis Management Science, Vol. 57, No. 1, pp. 200-214.
- Walley, Peter (1990). *Statistical Reasoning with Imprecise Probabilities*. London: Chapman and Hall.





