



Rapid Stochastic Assessment of Post-hazard Connectivity and Flow Capacity of Lifeline Networks



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Needs for Rapid Stochastic Assessment and Challenges

Stochastic assessment in DSS

▪ Needs for efficient stochastic assessment for lifeline networks

- Hazard **mitigation** planning (long-term)
 - can have many scenarios or cases
 - time-variant analysis (deterioration, aging, etc.)
- Quick hazard **responses** (short-term)
 - proximity of damaged components through probabilistic inference
 - immediate responses to minimize socio-economic losses
- **Integration** with optimization, information technology, monitoring system

▪ Computational challenges

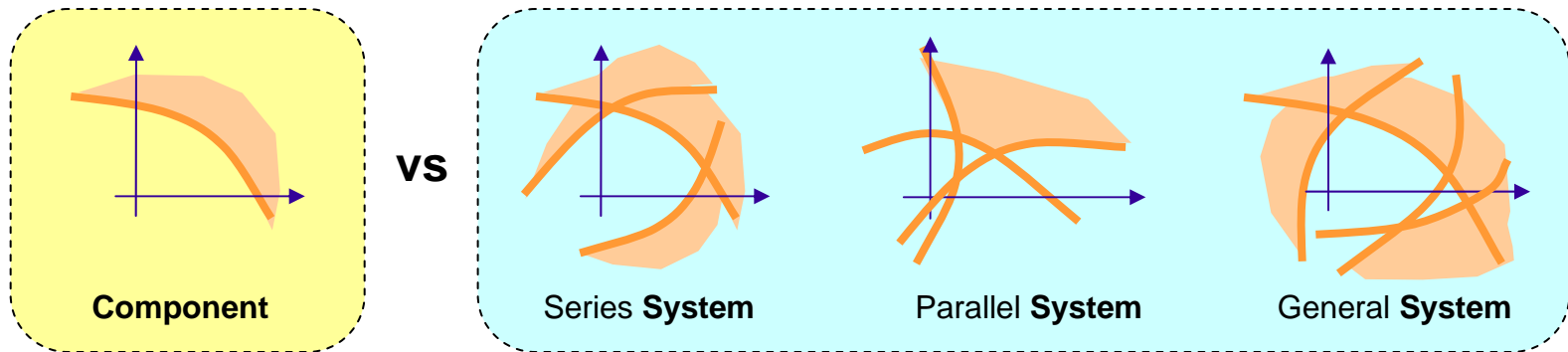
- **Complexity** of lifeline networks (size, system effect, etc.)
- **Spatial correlation** of hazard intensities (further increase the system size)
- **Importance** measures (component, uncertainty, etc.)
- Parameter **sensitivities**
- **Integration** with network flow (capacity and/or demand) analysis
- **Time-variant** analysis (deterioration)

It's system reliability problem!

* Song, J. and A. Der Kiureghian (2003). Bounds on system reliability by linear programming, *Journal of Engineering Mechanics*, 129(6), 627-636.

■ System reliability analysis (SRA)

- System event E_{sys} : logical function of component events, E_i , $i = 1, \dots, n$



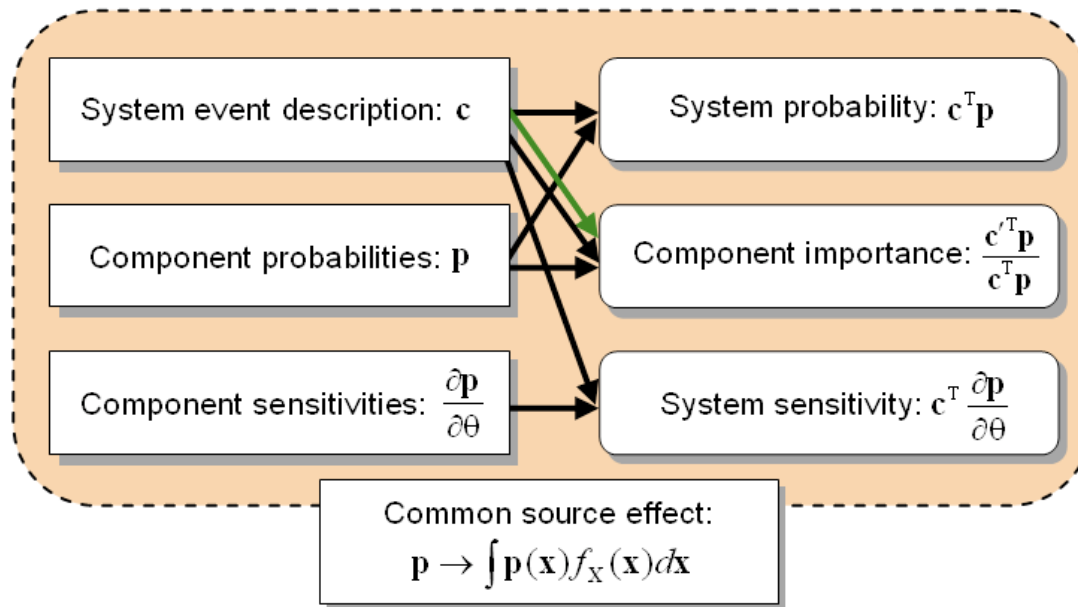
- $P(E_{sys})$: theoretical bounding formulas (Ditlevsen), sampling methods, and direct integration in random variable space

■ Challenges in SRA

- **Complexity** of system reliability problems: large number of components, difficulty in identifying critical cut sets or link sets, computational challenges
- **Statistical dependence** between component states (common source effect)
- **Diversity or lack** of available **information**
- Difficult to **identify important components or parameters** (critical for decision support)

Matrix-based System Reliability (MSR) method

* Song, J. and W.-H. Kang (2009). System reliability and sensitivity under statistical dependence by matrix-based system reliability method, *Structural Safety*, 31(2), 148-156.

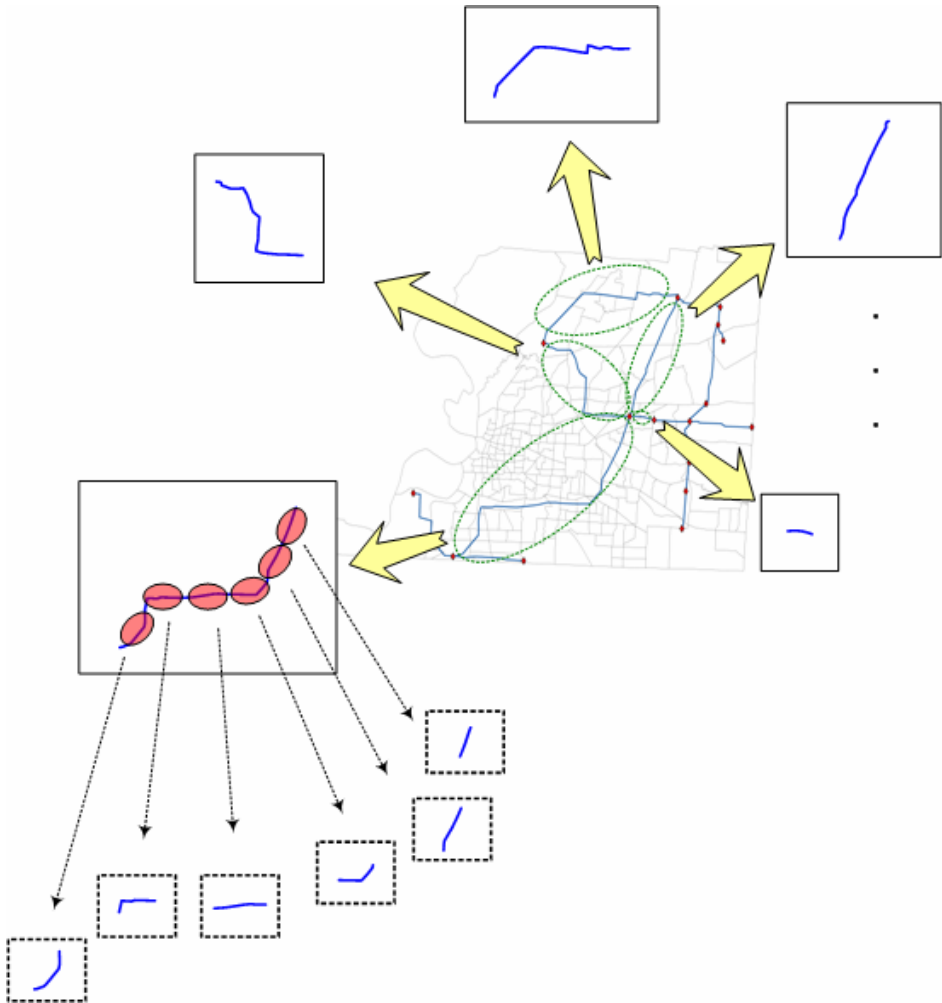


- **Convenient:** matrix-based procedures for \mathbf{c} and \mathbf{p} ; easy SRA calculation (inner product)
- **General:** uniform application to series, parallel, and any general systems
- **Flexible:** inequality-type information; incomplete information ("LP bounds" method)
- **Efficient:** no need to re-compute " \mathbf{p} "; replace " \mathbf{c} " for SRA of a new event
- **Common Source Effect:** can account for statistical dependence between components
- **Decision Support:** parameter sensitivities, component importance measure; inferences

App1: Post-hazard connectivity of gas pipeline network

Multi-scale SRA of lifeline networks

* Song, J., S.-Y. Ok, and L.. Chang (2008). Rapid risk assessment and decision support for urban infrastructure networks by MSR method. *Proc. Inaugural International Conference of the Engineering Mechanics Institute*, May 18~21, Minneapolis, MN.



■ “Divide and Conquer” approach

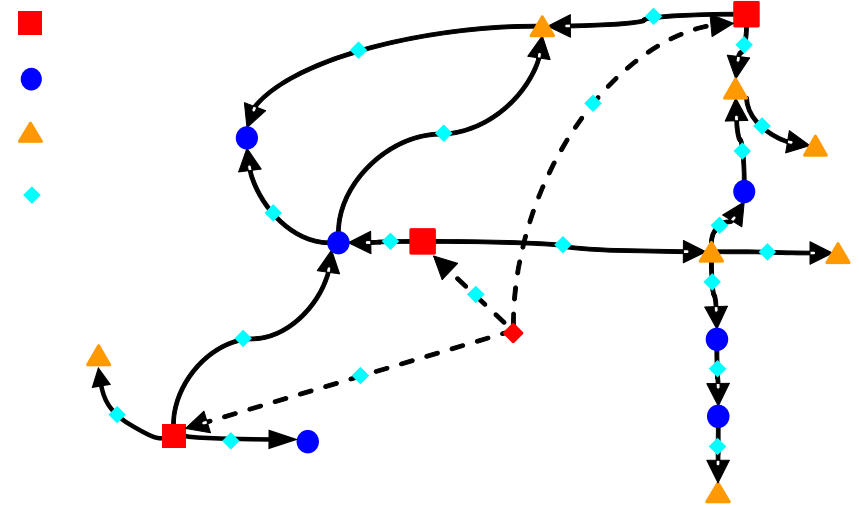
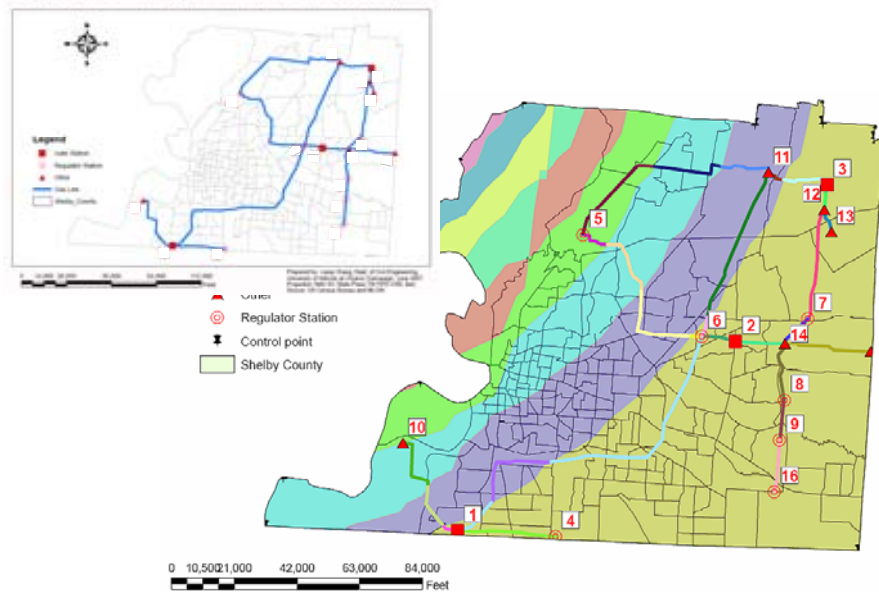
- Lower-scale system reliability analyses are performed for “supercomponents” and followed by higher-scale system reliability analyses
- Proposed to facilitate the use of **LP bounds method** (Song and Der Kiureghian, 2003) for large-size systems
- **MSR method** is a good tool for SRA at multiple scales

■ Advantages

- **Multi-scale modeling** of a system – seeing big picture without disregarding the details
- Helps identify **important** components and parameters at **multiple scales**
- **Collaborative** risk management
- Facilitates parallel computing

Example: MLGW gas network

MLGW Gas Transmission System in Memphis and Shelby County, TN

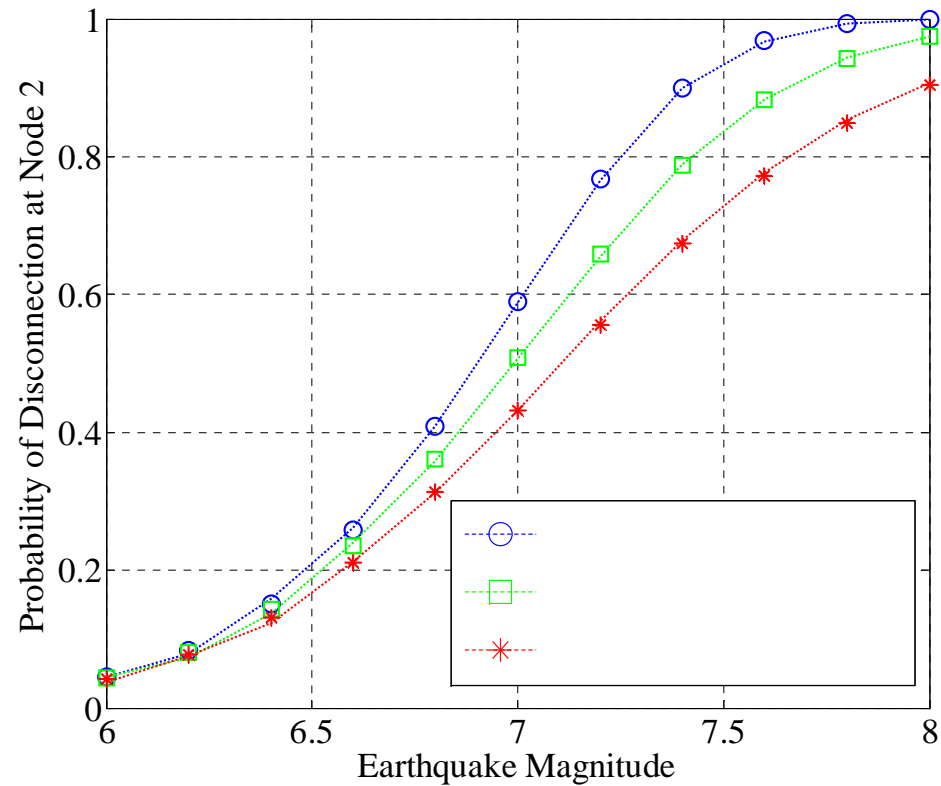
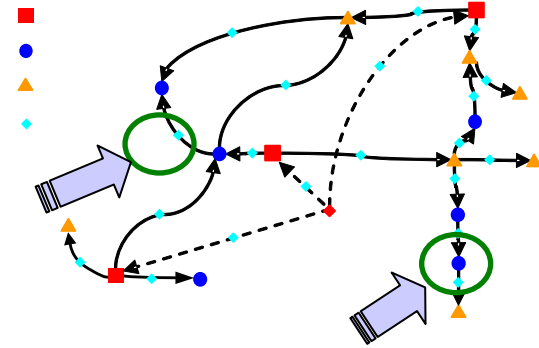
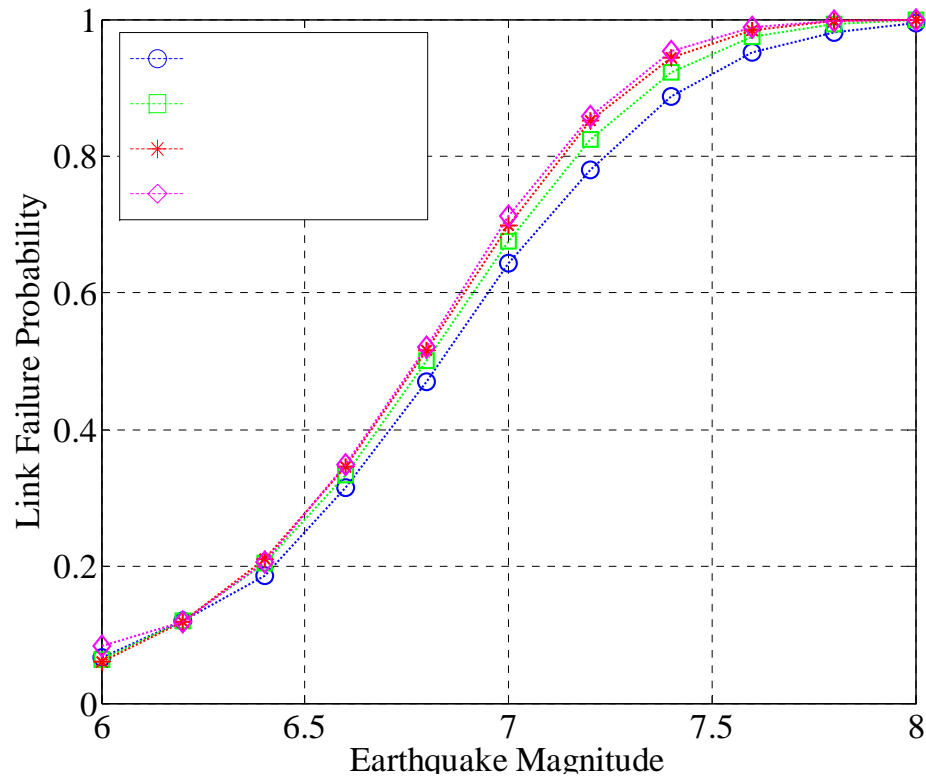


- Gas pipeline network of Memphis Light, Gas, and Water (MLGW), Shelby County, TN
- A simplified network in Chang et al. (1996) was modified based on comments from R. Bowker (MLGW)
- 37-node and 40-arc network: nodes representing pipelines and stations
- Earthquake hazard scenarios: Epicenter at N35.54°-W90.43° at Blytheville, AR
- Fragilities of pipelines and stations – *HAZUS-MH*
- PGV and PGA maps from *MAEviz*

Risk at multiple scales

Lower-scale: pipelines

Failure probability of **Link 25**

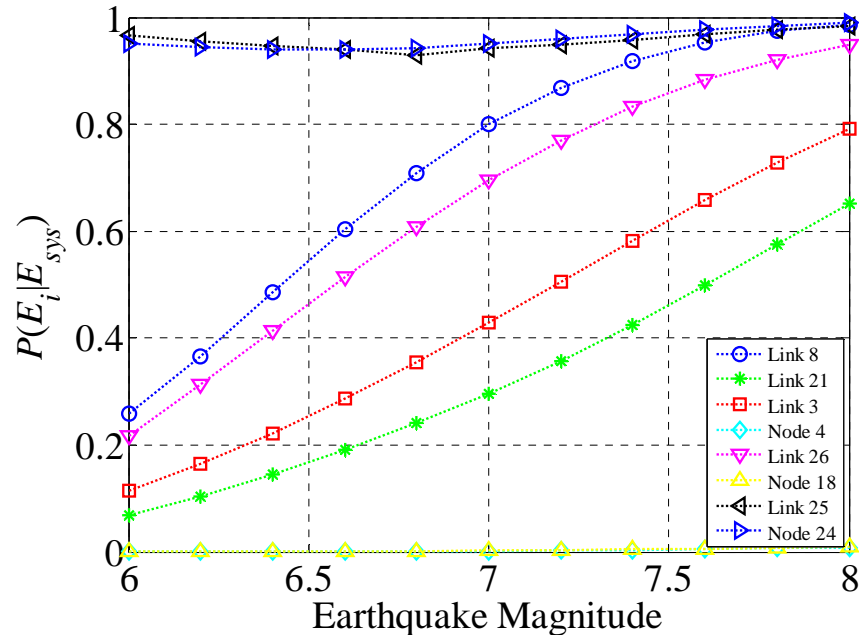


Higher-scale: service nodes

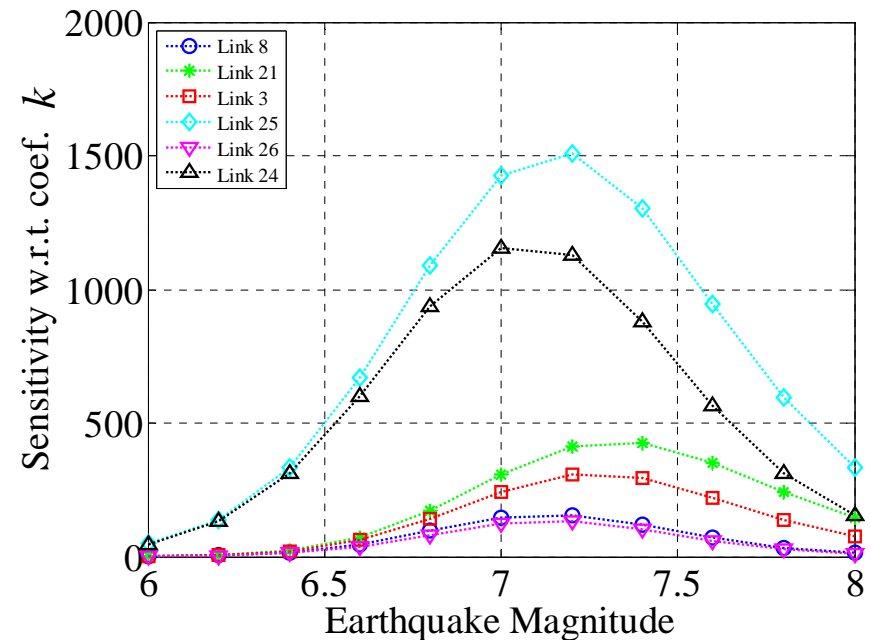
Prob. of Disconnection at **Node 2**

Probabilistic inference and sensitivity

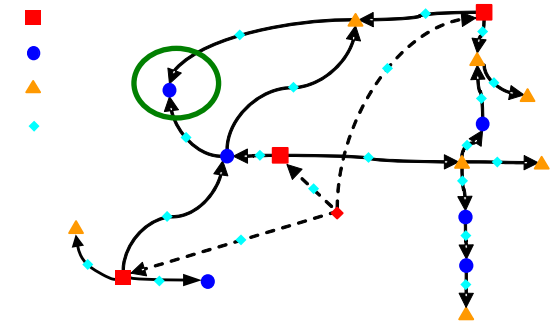
Conditional Probabilities



Parameter Sensitivity



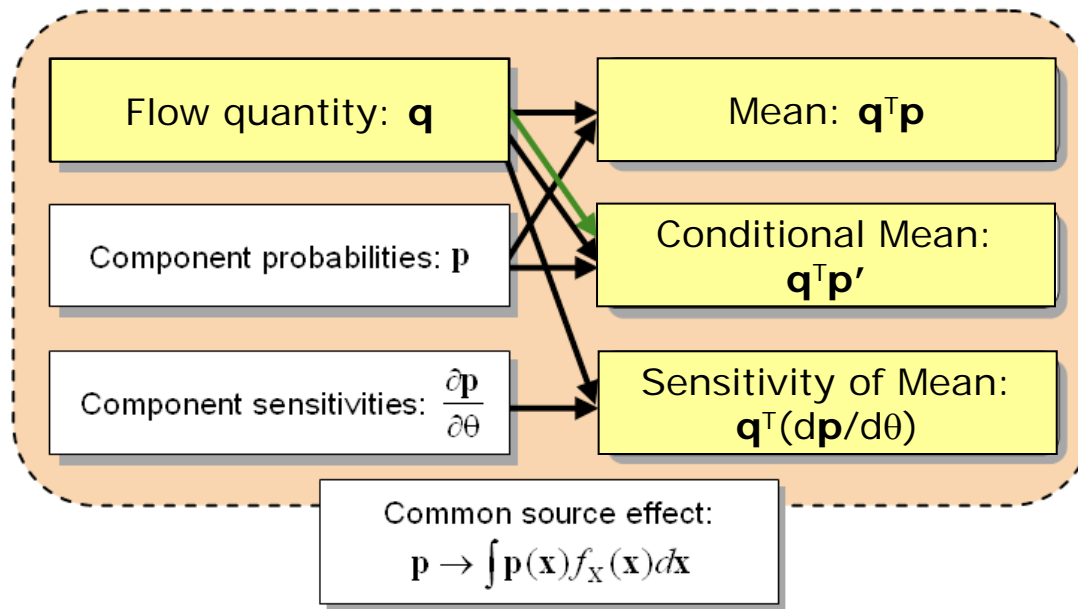
- Conditional probability of link failure probability given observed system event (e.g. disconnection)
- Sensitivity of system failure probability with respect to parameters in PGV-based model for failure occurrence rate: $v_i = k \cdot (PGV_i)^\gamma$



App2: Post-hazard flow capacity of bridge transportation network

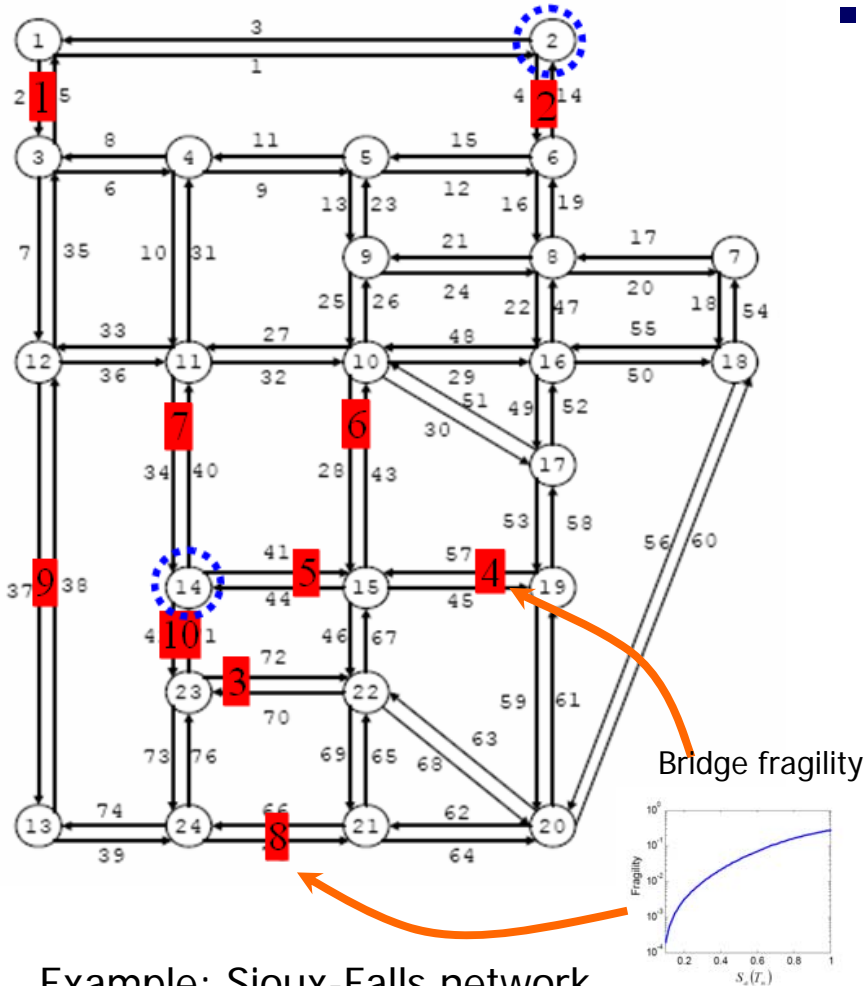
Extension for network “flow” analysis

* Lee, Y.-J., J. Song, and P. Gardoni (2009). Post-hazard flow capacity of bridge transportation network considering structural deterioration, *ICOSSAR2009* (will be presented), Osaka, Japan.



- **Multi-state failures:** can handle more than two failure/damage states
- **Separation between network flow analysis and vulnerability analysis:**
 - 1) No need to repeat network flow analysis (\mathbf{q}) for time-varying fragility (deterioration, etc.)
 - 2) No need to repeat probability calculations (\mathbf{p}) for changes in network (new routes, etc.)
 - 3) Flow analysis for damage scenarios with high likelihood only (for approximation)

Post-hazard flow capacity of a bridge network



Example: Sioux-Falls network

Red: bridges; Circles: Starting & Ending points

- Traffic flow **capacity** between two points in a network \rightarrow determined by combinations of bridge damages

\mathbf{q} : a vector of network flow capacity for bridge failure combinations (obtained by maximum flow capacity analysis)

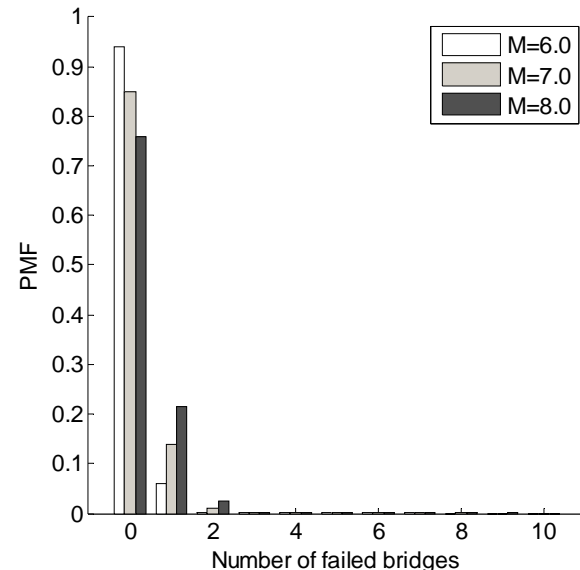
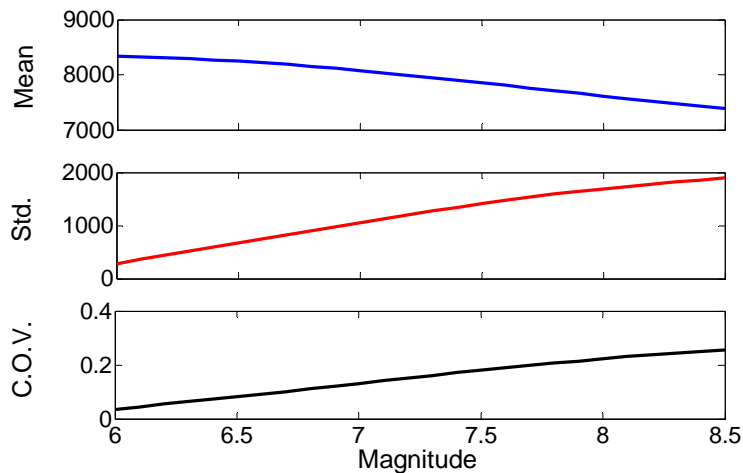
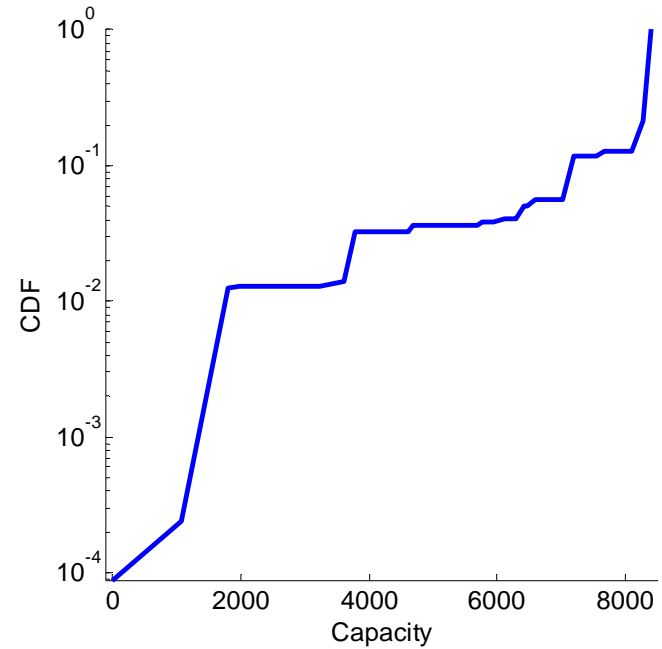
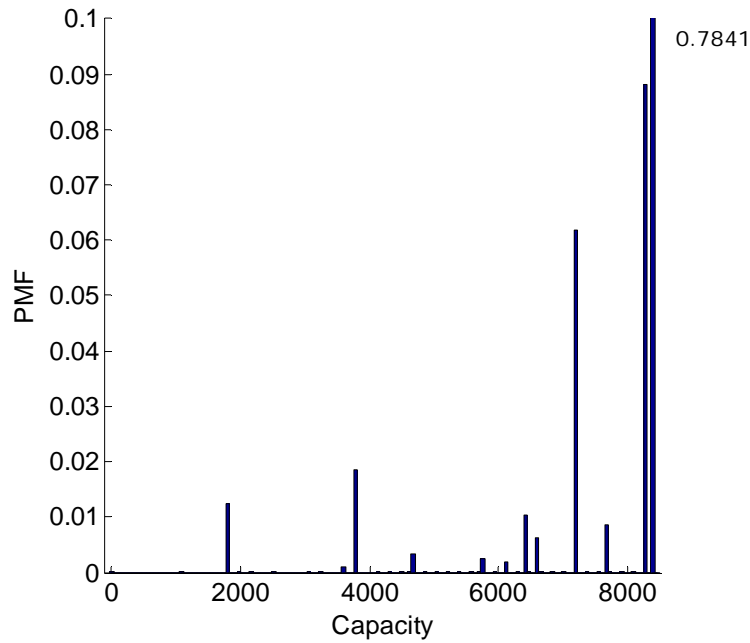
$\mu_Q = \mathbf{q}^T \mathbf{p}$: average post-hazard flow capacity

$\sigma_Q^2 = (\mathbf{q} * \mathbf{q})^T \mathbf{p} - (\mathbf{q}^T \mathbf{p})^2$
: variance of post-hazard flow capacity

$$P(Q < a) = \sum_{\forall i: q_i < a} p_i$$

: probability that flow capacity is lower than a

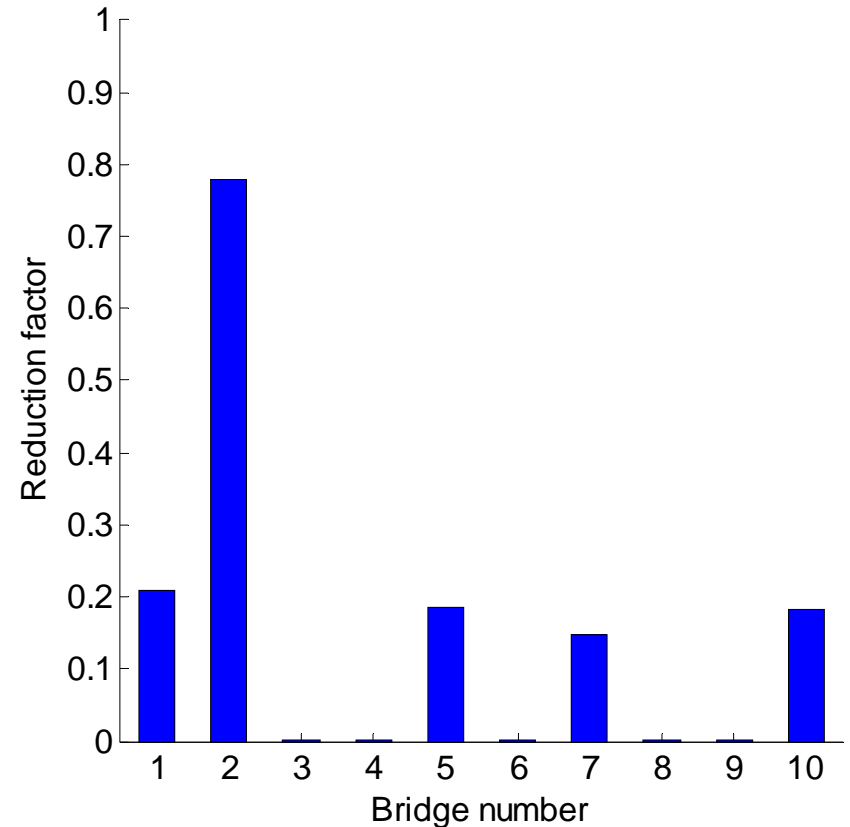
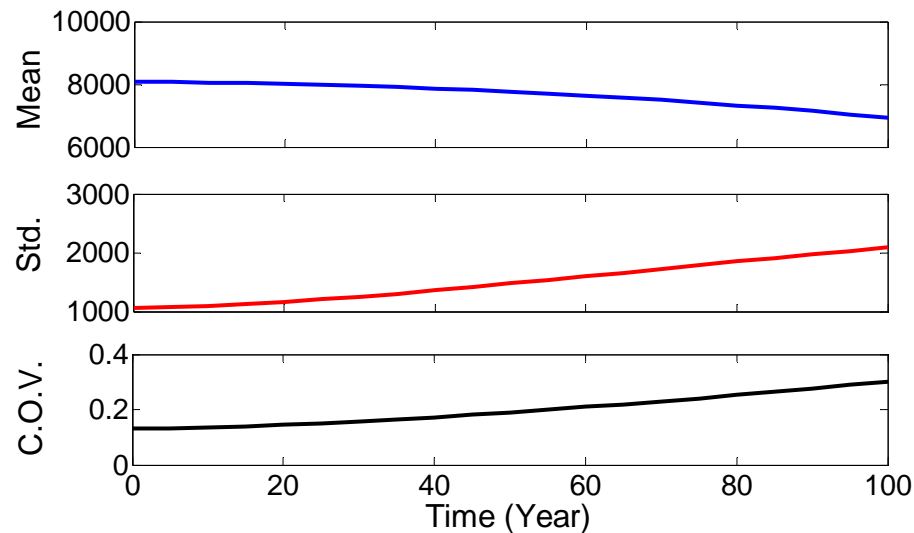
Uncertainty quantification of flow capacity



Time-variant analysis; Reduction factor

Time-variant mean, std., and c.o.v. of flow capacity

Reduction of mean flow capacity by the failure of a bridge



$$\mu_Q(t) = \mathbf{q}^T \mathbf{p}(t)$$

$$\sigma_Q(t) = \sqrt{(\mathbf{q} \cdot \mathbf{q})^T \mathbf{p}(t) - \mu_Q^2(t)}$$

$$RF = 1 - \frac{\mu_{Q|\text{bridge failure}}}{\mu_Q}$$

Summary

▪ Research needs and recent developments

▪ Multi-scale system reliability analysis:

- Spatial correlation of seismic intensity
- Large size network
- Control modeling efforts at multiple scales
- Component importance measures
- Parameter sensitivities

▪ Separation between network flow analysis and vulnerability analysis:

- Flexible uncertainty quantification of network flow metrics
- Efficient network flow analysis due to separation
- Importance measures w.r.t. network flow

▪ Ongoing/future research

- Online integration with information technology including monitoring system
- Near-real-time probabilistic inference through machine learning algorithms

Thank you!