

# Resilience of Intermodal Transportation Infrastructure under Multiple Hazards

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## Structural reliability and natural hazard risk mitigation

- Key emphases in the areas of
  - Portfolio structural vulnerability analysis and protection
  - Multi-hazard assessment (earthquake, surge, wave, aging...)
  - Sustainability and resilience quantification
- Application to transportation and energy/industrial infrastructure



Bridges



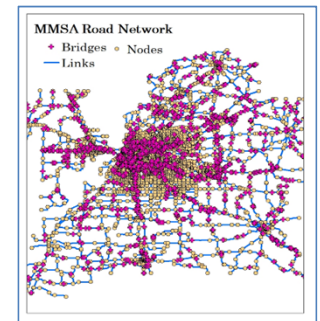
Above Ground Storage  
Tanks



Port Structures and  
Facilities



Industrial Complexes  
and Communities



Regional Scale  
Transportation

# Our Group at Rice University





Figure 1. Dominant hazard per region based on infrastructure exposure (Koks et al. 2019)

Koks et al. (2019) "A global multihazard risk analysis of road and railway infrastructure assets" *Nature Communications*. <https://doi.org/10.1038/s41467-019-10442-3>



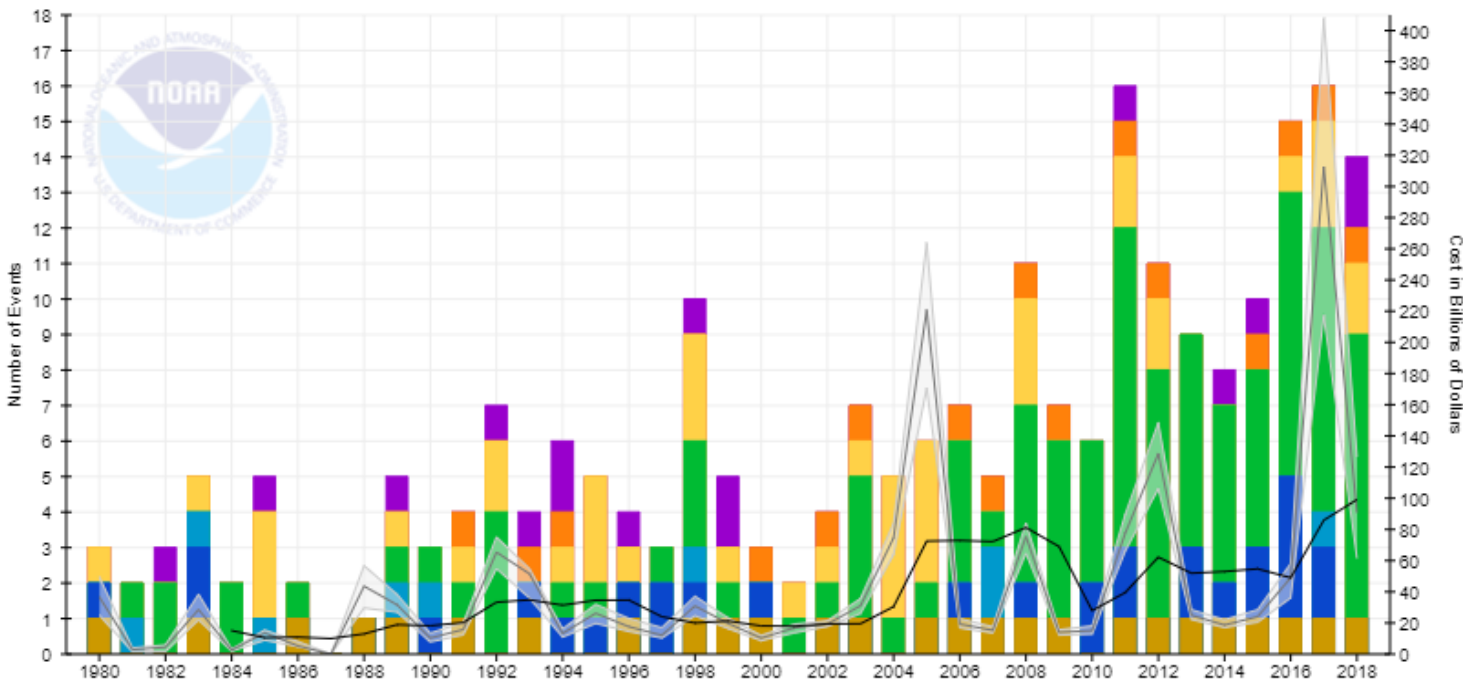


RICE

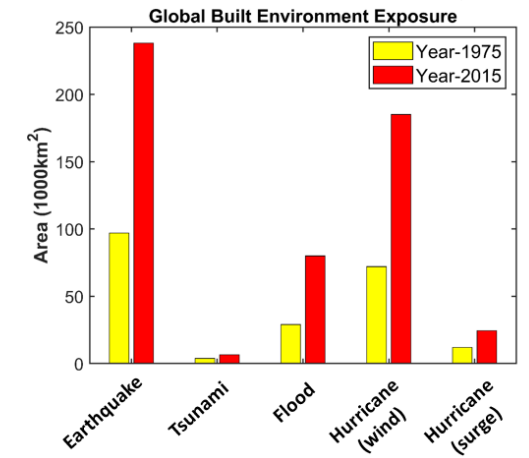
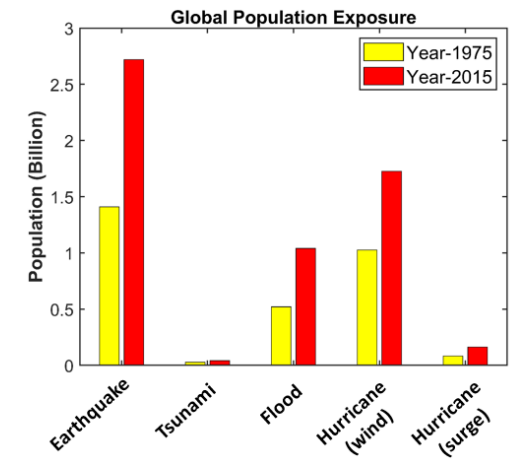
# Increasing Hazard Exposures and Losses

### Billion-Dollar Disaster Event Types by Year (CPI-Adjusted)

- Winter Storm
- Wildfire
- Trop Cycl
- Severe Storm
- Freeze
- Flooding
- Drought
- Cost w/ 95% CI
- 5-Year Mean



(Source: NOAA 2019)



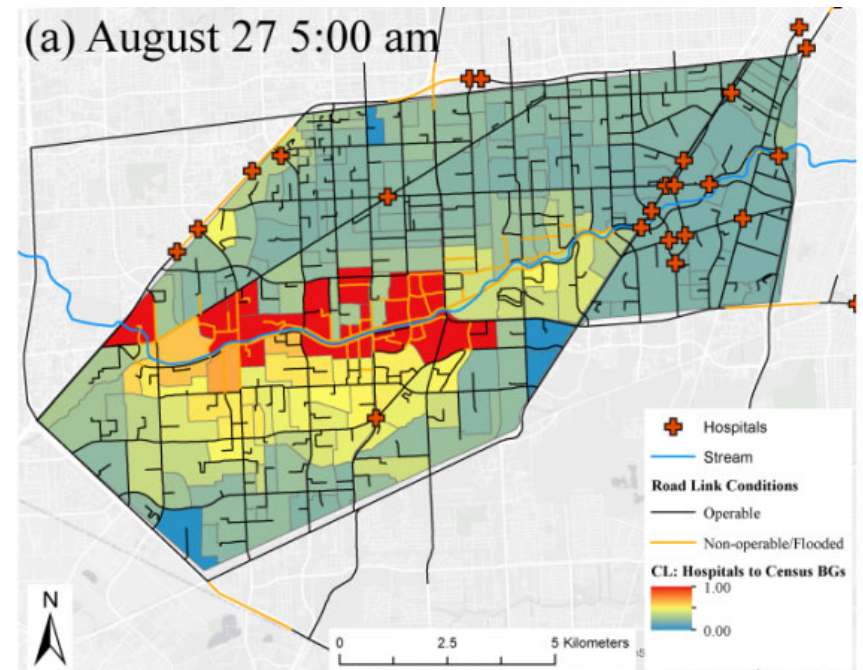
(Adapted from Pesaresi et al. 2017)

# Transportation Infrastructure Resilience

- Transportation infrastructure resilience is crucial during/after hazard events.



Hurricane Harvey (2017)

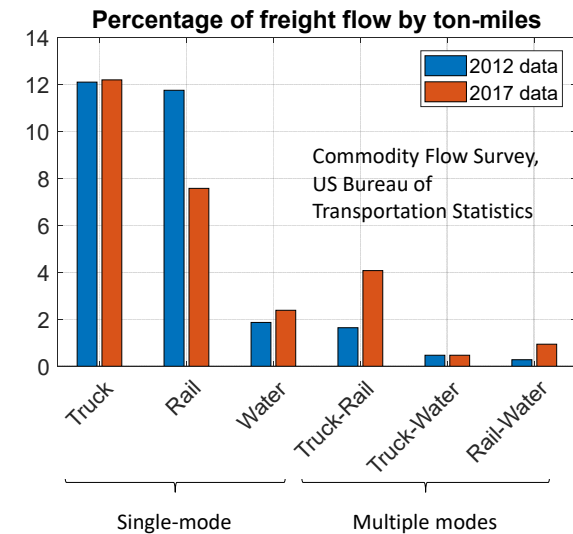


Gori et al. (2019)

# Importance of Intermodal Transport



- Impact of multiple hazards on transportation infrastructure
- Importance and complexity of modeling intermodal transportation systems
- Potential to disrupt goods flow



1. Transportation infrastructure resilience framework
2. Key input models
3. Application examples
4. Conclusions, challenges and opportunities



# Definitions of Resilience

The ability to prepare and plan for, absorb, recover from, and more successfully adapt to adverse events. (National Academy of Science, 2012)

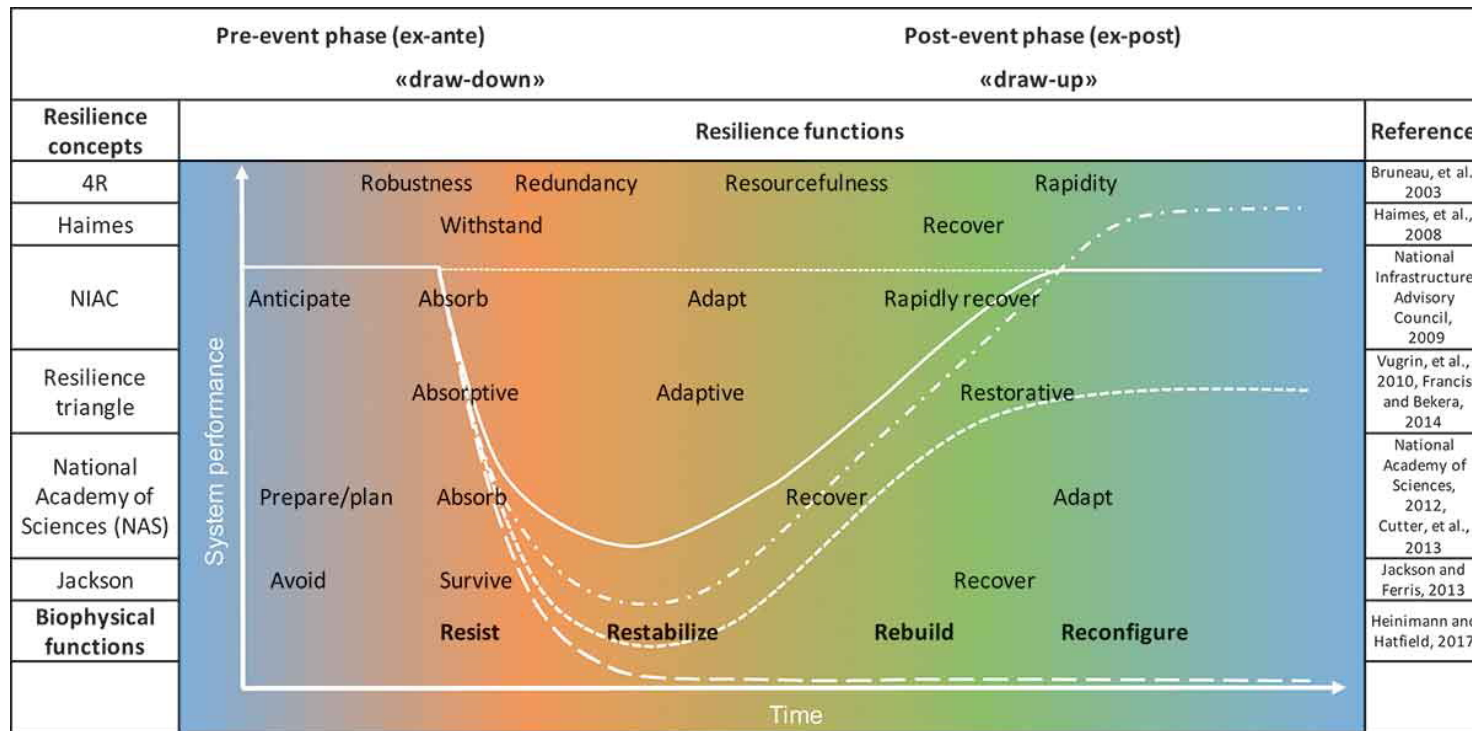
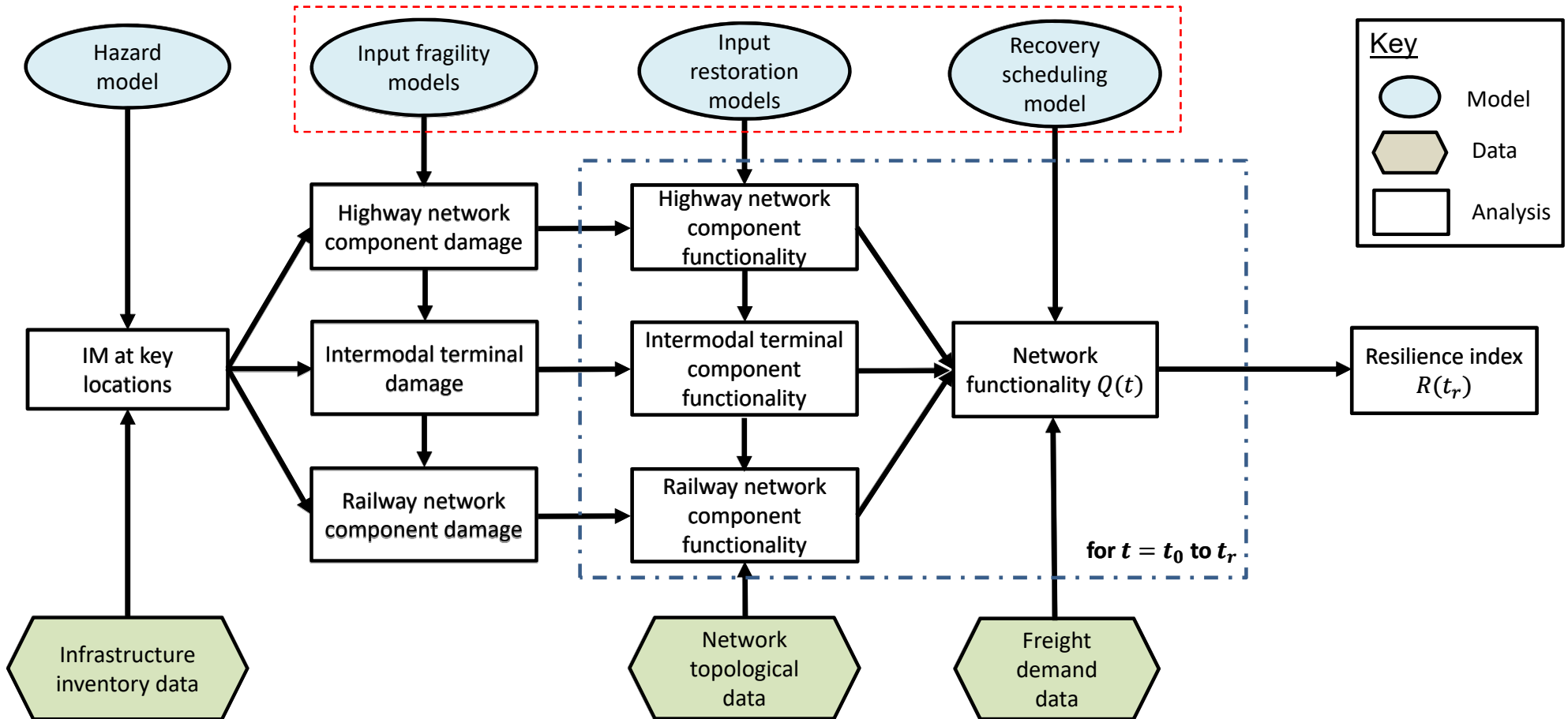


Figure 1. Example resilience characterization drawing from state-of-art review (Gasser et al. 2019)

Gasser et al. (2019) "A review on resilience assessment of energy systems" *Sustainable and Resilient Infrastructure*. doi.org/10.1080/23789689.2019.1610600

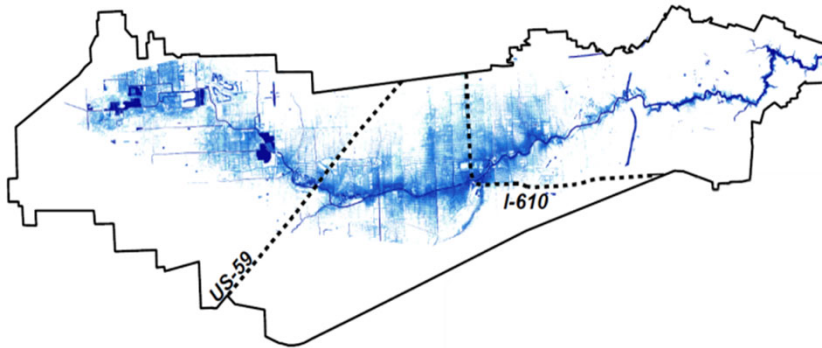


# RICE Intermodal Resilience Analysis Framework

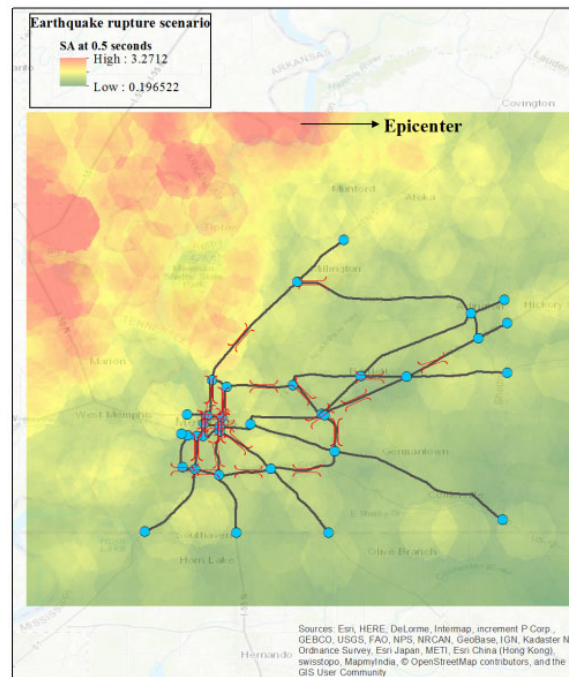


# Key input: *Hazard models*

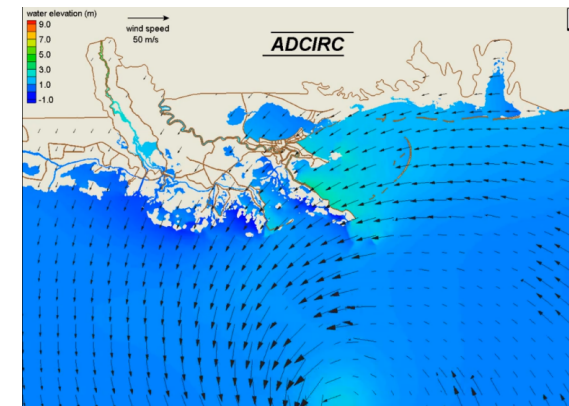
- Hazard occurrence characterization and spatial hazard intensity modeling
  - Physics based
  - Surrogate model
  - Single / multi-hazard



Flood (Gori et al. 2019)



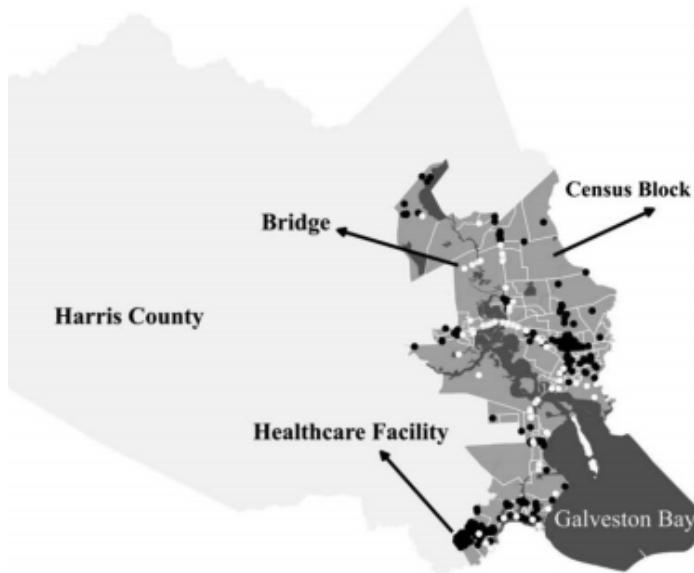
Earthquake (Vishnu et al. 2018)



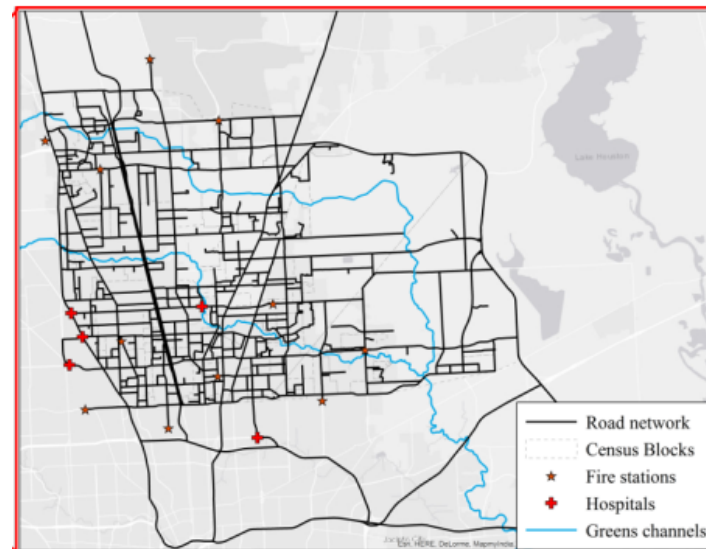
Hurricane  
(Dawson et al, 2017)

# Key input: *Exposure models*

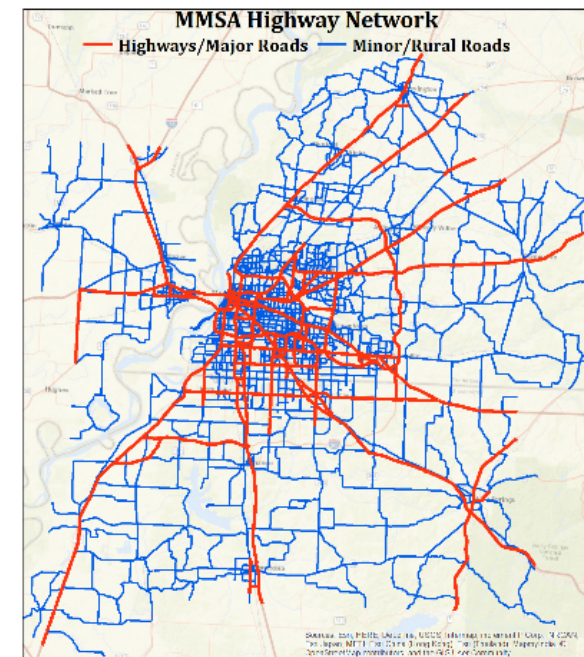
- Provide information (e.g., location, structural type, and replacement cost) of the spatially distributed structures



Balomenos et al. (2019)



Gori et al. (2019)

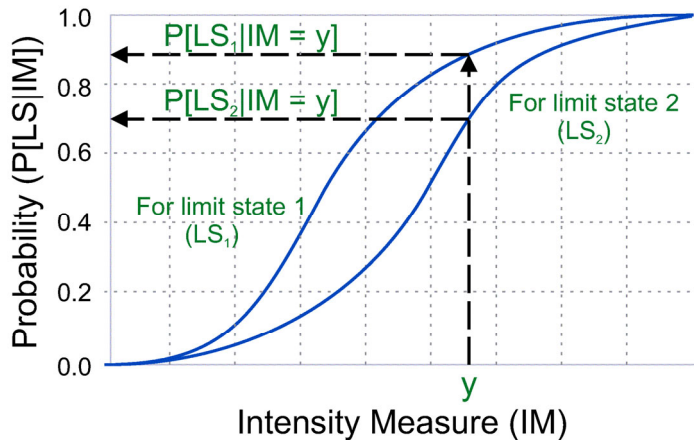


Vishnu (2018)



# Key input: *Fragility models*

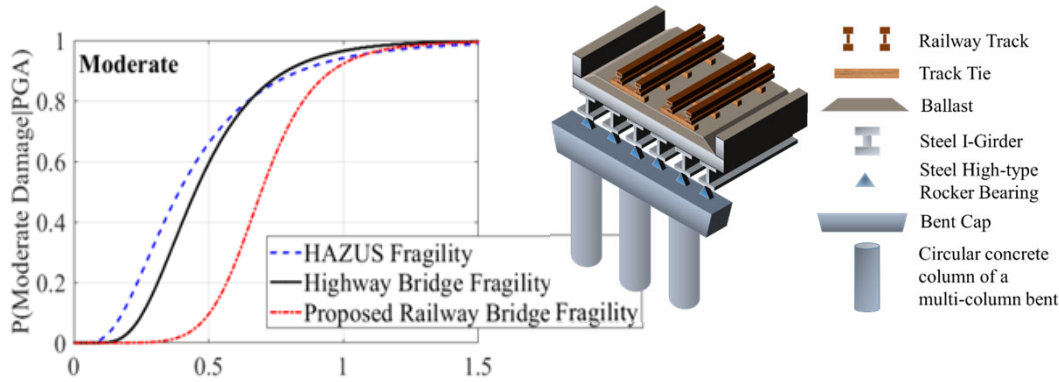
- Provide probability of exceeding a certain limit state for the structures given the hazard intensities
  - Surrogate demand modeling
  - Limit state capacity characterization
  - Fragility curves/Parameterized fragility function



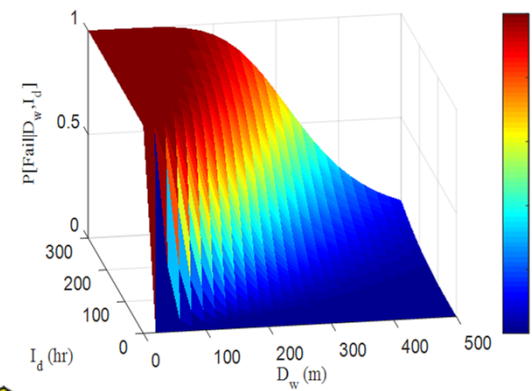
$$P(\text{Failure}|\mathbf{IM}, \mathbf{X}) = \frac{1}{1 + \exp(-l(\mathbf{IM}, \mathbf{X}))}$$

↓  
Logit function

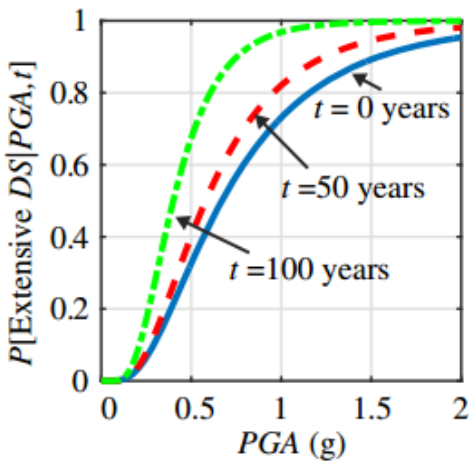
# Key input: *Fragility models*



Railway bridge fragility (Misra & Padgett, 2019)

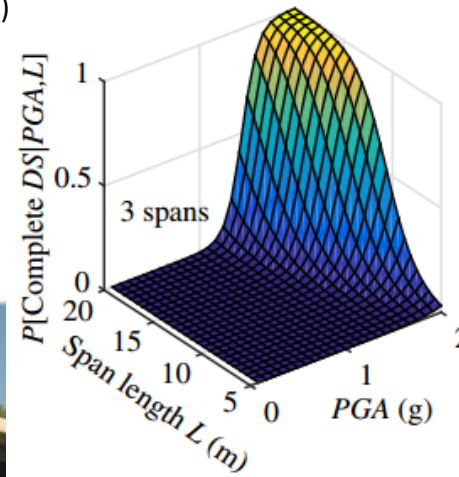


Roadway fragility (Gidaris et al., 2019)

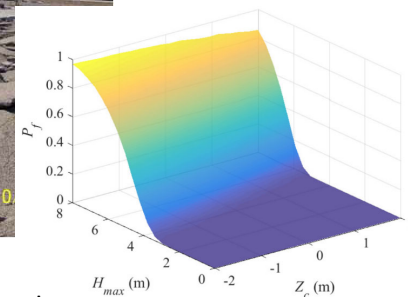


Time-dependent bridge fragility (Ghosh & Padgett, 2010)

Parameterized highway bridge fragility (Kameshwar & Padgett, 2014)

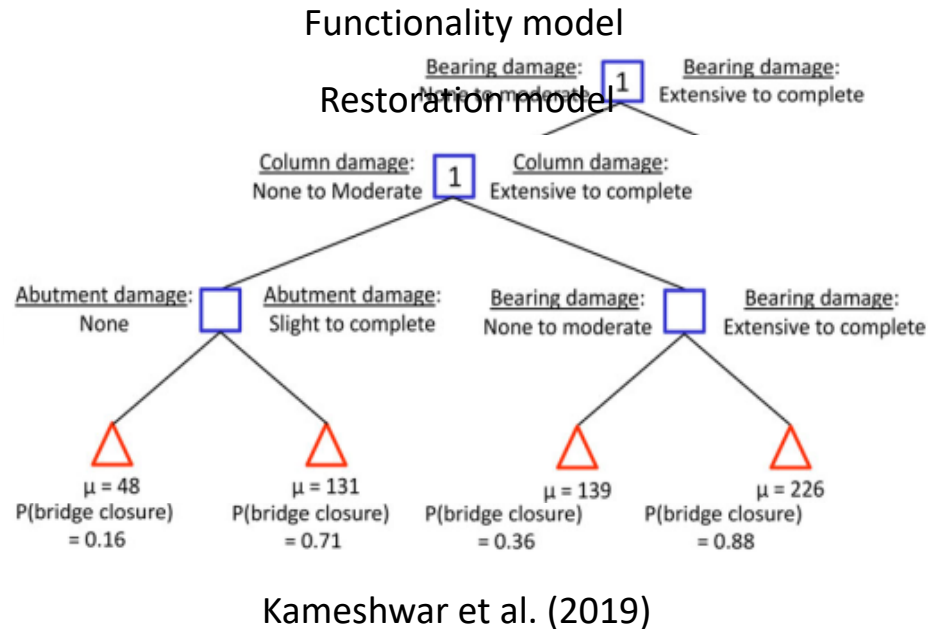
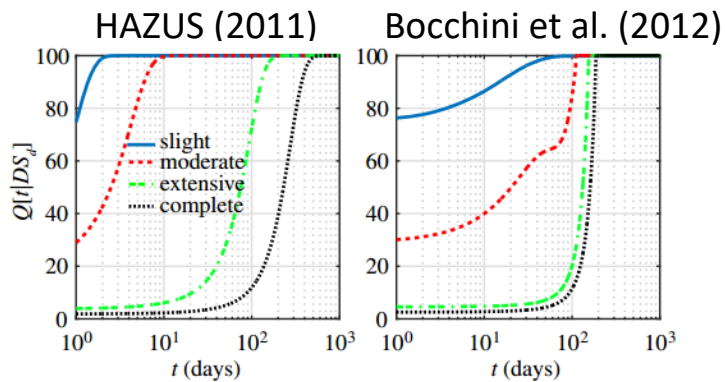
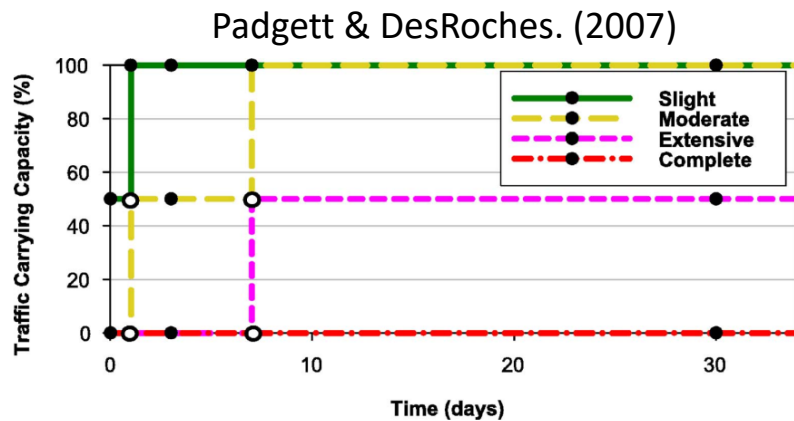


Piers and wharfs at ports (Balomenos and Padgett, 2019)

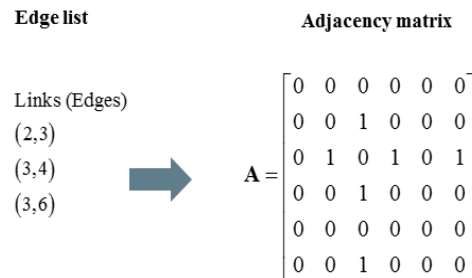
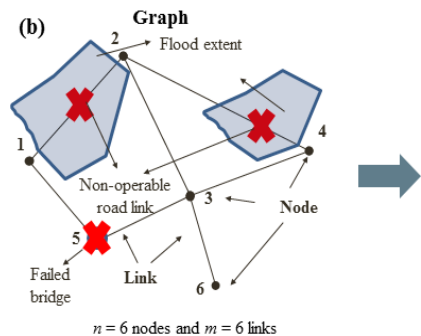
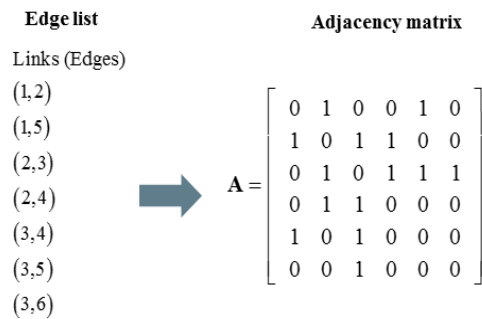
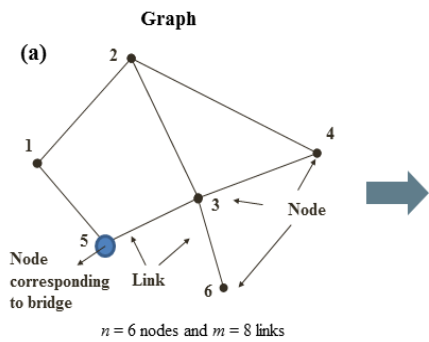


# Key input: *Restoration models*

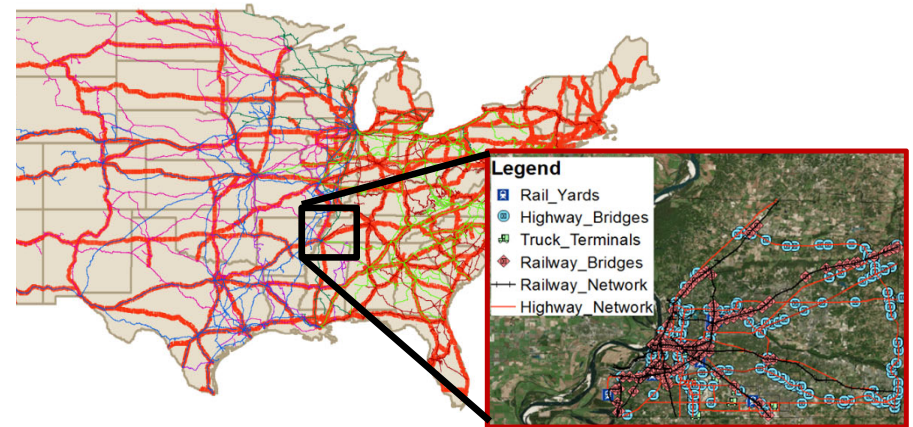
- Quantify temporal functionality and restoration of structures & systems



# Key input: *Recovery Scheduling and Network Analysis*



Bernier et al. (2019)



Misra and Padgett (2019)

- Heuristic or optimal deployment of repair crews and recovery resources across networks



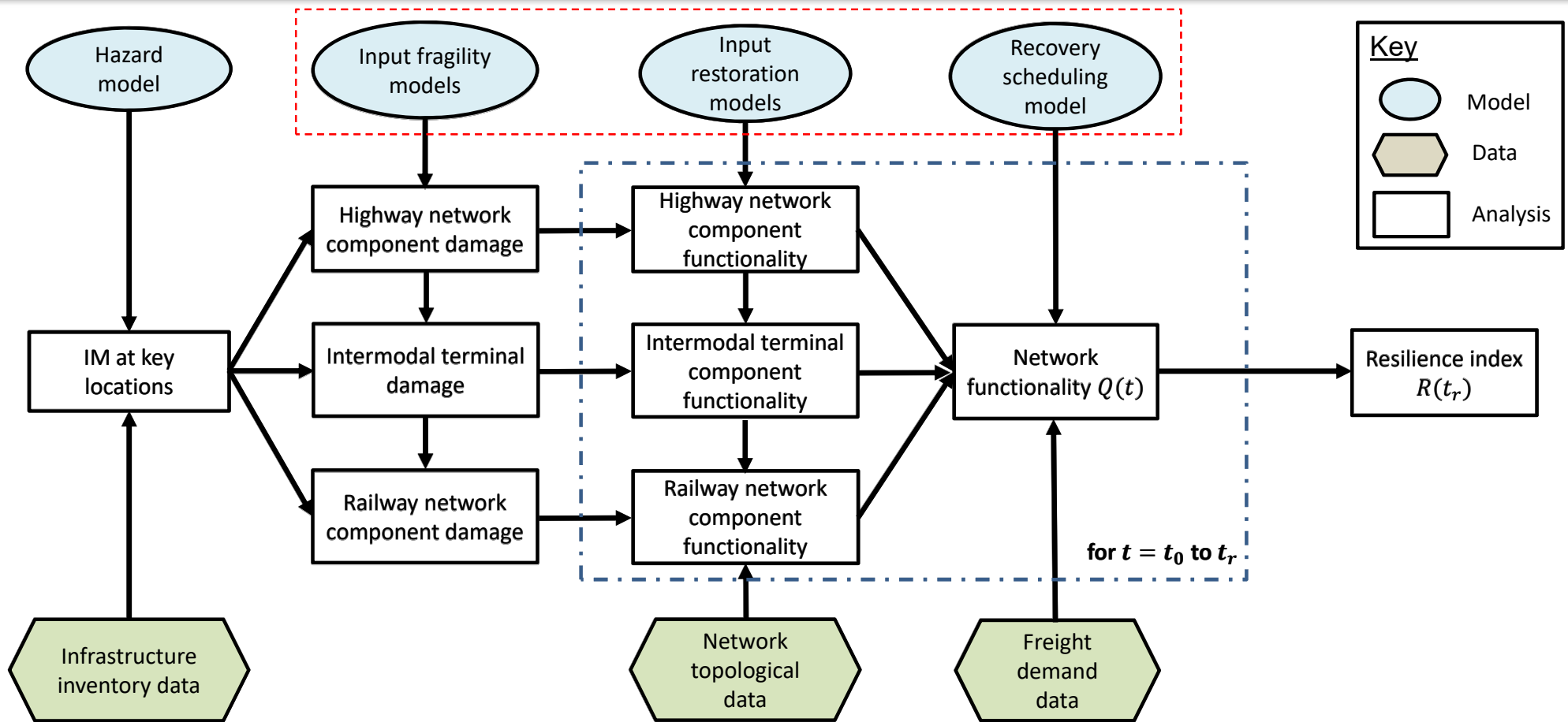
# Seismic Resilience Modeling of Rail-Truck Intermodal Transportation Networks

- *Memphis Metropolitan Statistical Area (MMSA)*

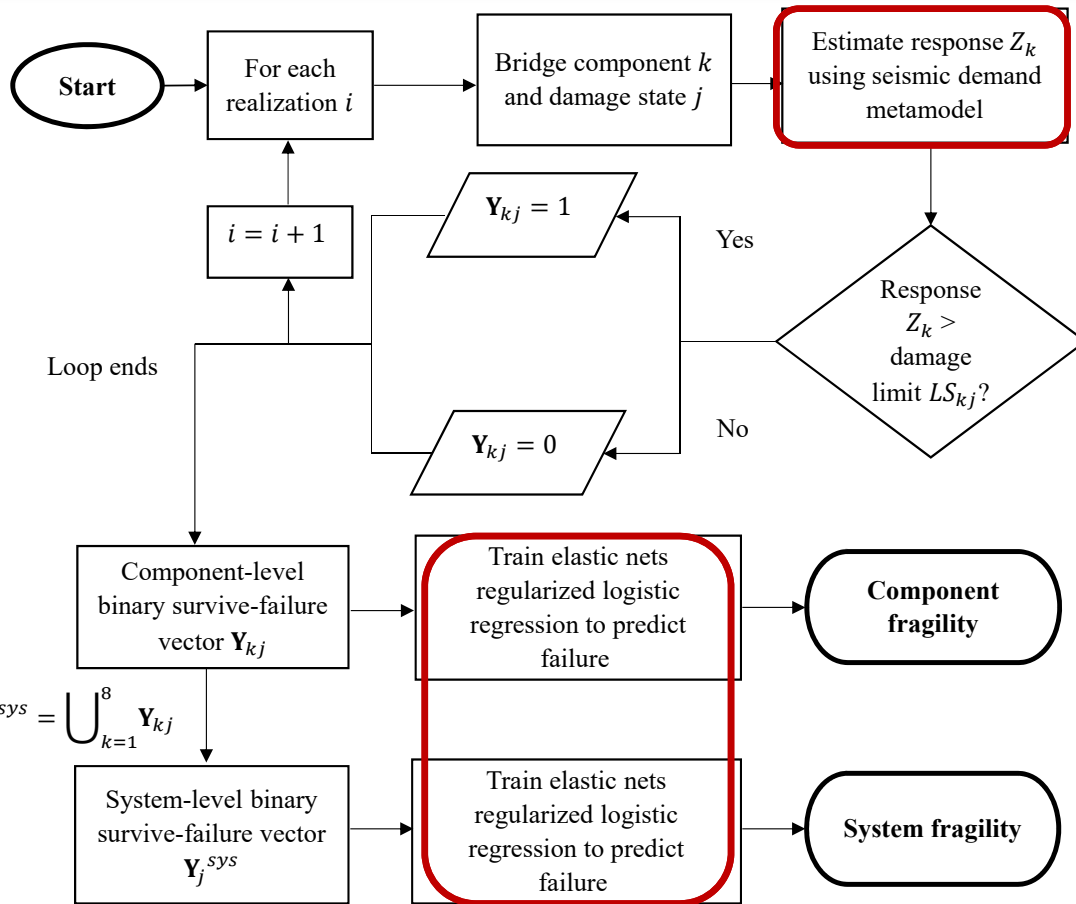


Misra, S., Padgett, J. E. (2019). Seismic Resilience of a Rail-Truck Intermodal Freight Network. *13th International Conference on Applications of Statistics and Probability in Civil Engineering(ICASP13)*, Seoul, South Korea

# Intermodal Resilience Analysis Framework

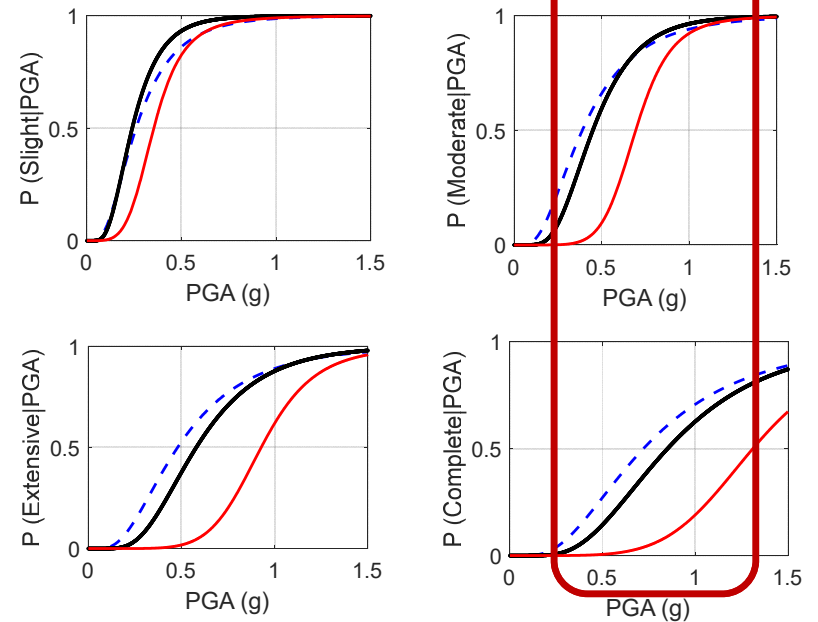


# Parameterized Fragility Method



$$P(Y_j > Y_{lim} | IM, Y_j) = \frac{e^{\theta_0 + \sum_{j=1}^p \theta_j X_j^{(i)}}}{1 + e^{\theta_0 + \sum_{j=1}^p \theta_j X_j^{(i)}}}$$

Ground Motion Suite | Bridge Model Simulation | Response of Bridge Components



- - - HAZUS Fragility  
 — Nielson (2005) Fragility  
 — Proposed Fragility

- Online surveys of experts in post-hazard repair and restoration of roads, railway tracks and bridges were carried out.
- Goals of the surveys – harvest data relating various damage levels (component damage for bridges) to **closure decisions** (both complete and partial) and their **durations**.

\* What type of traffic closure decision would be recommended immediately after inspection?

\* Which of the following forms of partial closure would you specifically recommend? (You may select more than option)

Lane restriction

Speed restriction

Load restriction

\* The closed bridge may be opened fully to traffic directly or in stages. After how many days can the following decisions be taken? (fill in relevant fields only)

Bridge is opened, but with traffic restrictions in place

Lane restriction is lifted

Speed restriction is lifted

Load restriction is lifted

Bridge is fully opened to traffic

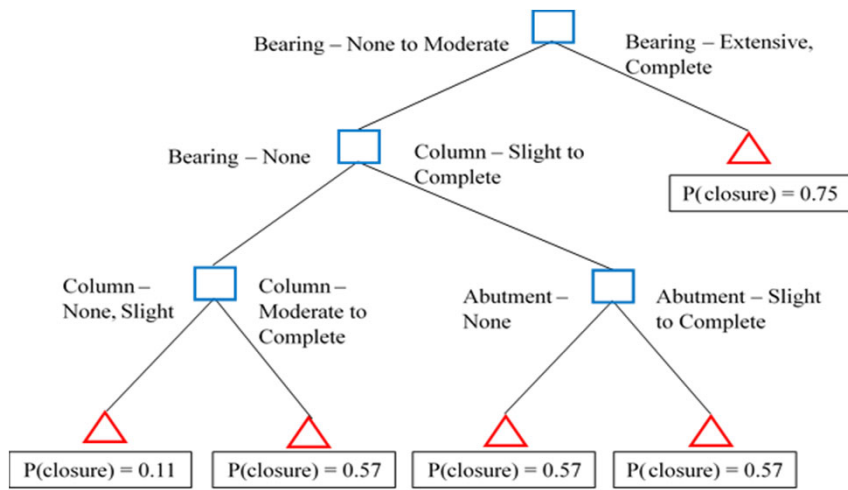
<sup>1</sup> Misra, S., Padgett, J. E., Barbosa, A. R., & Webb, B. M. (2020). An expert opinion survey on post-hazard restoration of roadways and bridges: Data and key insights. *Earthquake Spectra*, 36(2), 983-1004.

<sup>2</sup> Misra, S. and J. E. Padgett. (2017). Post-Earthquake Restoration Modelling of a Railway Bridge Network. *ASCE Congress on Technical Advancement*, Duluth, Minnesota, 10-13 September 2017.



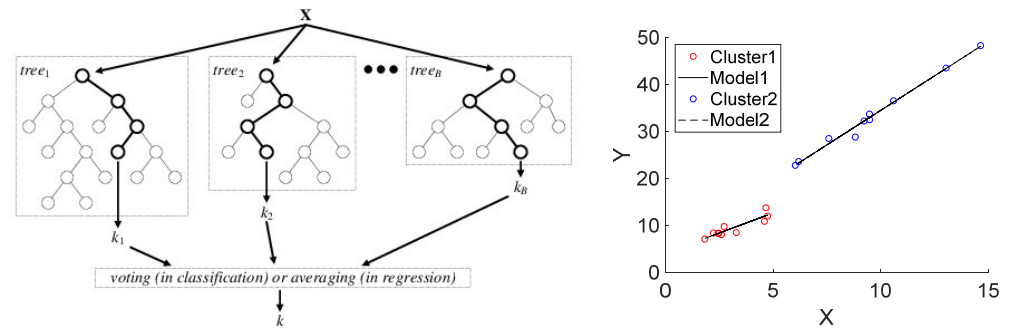


## Models for closure decision



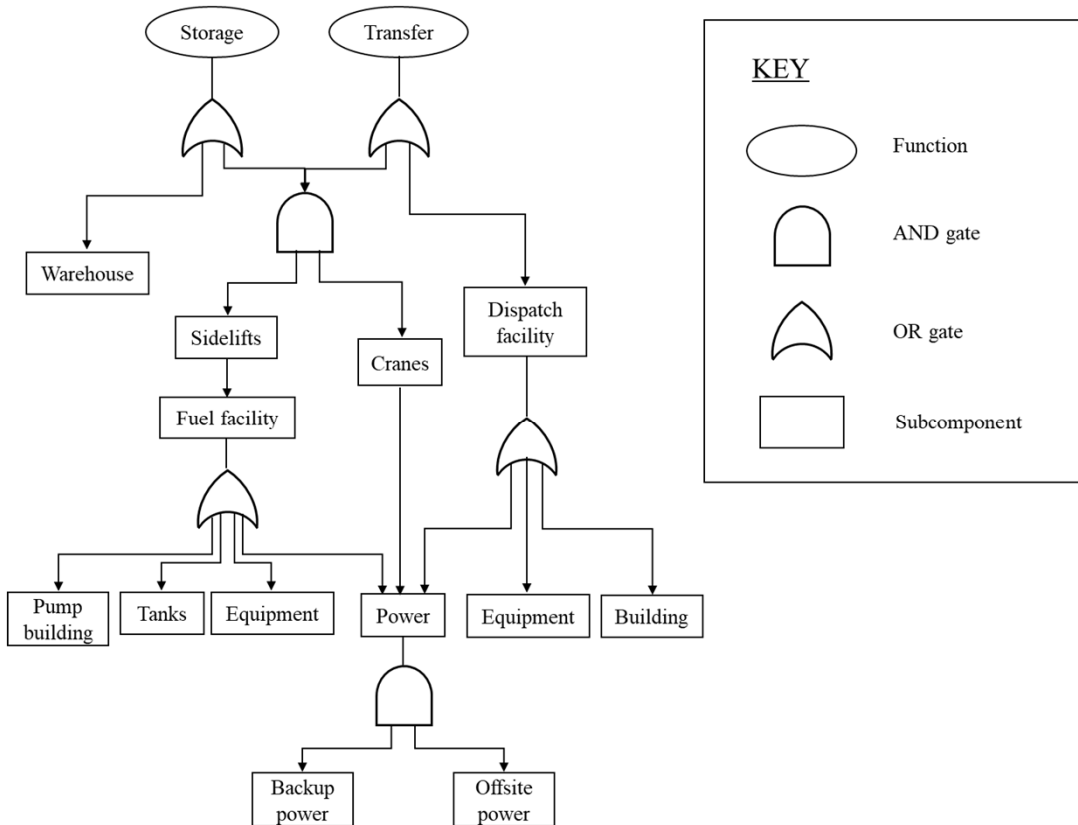
- **Decision trees** are leveraged
- Simple, interpretable, reasonable predictive capability

## Models for closure duration



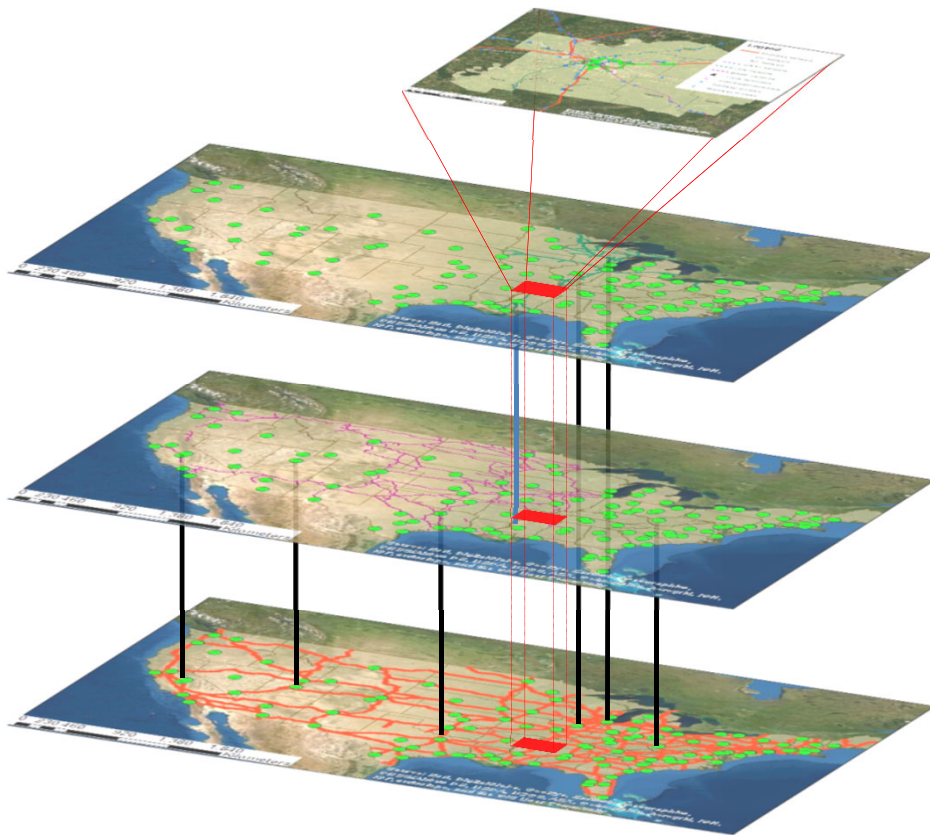
- **Clustered random forests** are leveraged
- Random forests – ensemble of decision trees
- Data divided into clusters – separate model fit to each cluster
- Reduces variance of predicted durations

# Restoration Models for Intermodal Terminals

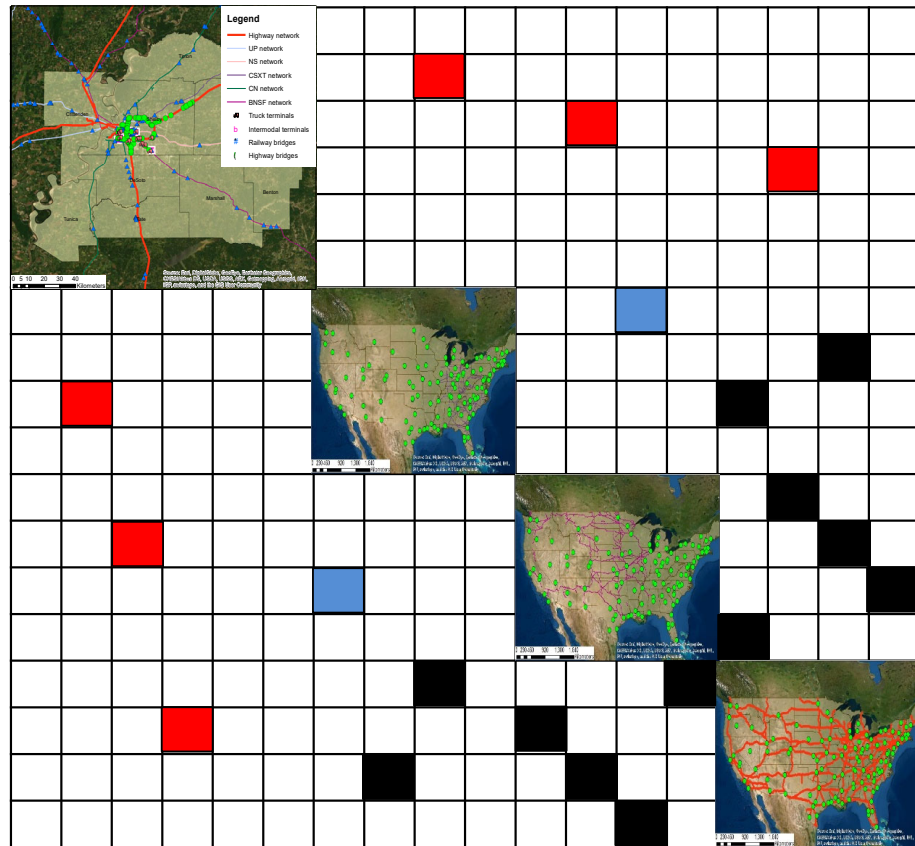


- A schematic ***fault tree model*** is proposed to illustrate the functional dependencies between the components and subcomponents of the intermodal terminal
- An AND connector indicates that all the subcomponents listed under the gate (children) must fail for the parent node to fail
- OR connector indicates that the parent fails if any one of the children fail
- Restoration models of each subcomponent obtained from HAZUS-MH (FEMA 2015)

## Layers of intermodal network



## Adjacency matrix



**Minimize**  $\alpha \times O1 + (1 - \alpha) \times (O2 + O3 + O4)$

- $O1$  = Cost of repair
  - $O2$  = Cost of freight transport
  - $O3$  = Cost of unmet demands
  - $O4$  = Cost of excess supply
- } Owner's cost  
 } Shipper's cost

### Subject to

- Flow conservation constraints
- Constraints relating functionality with repair actions
- Constraints relating component closure status to functionality
- Link capacity constraints
- Resource constraints

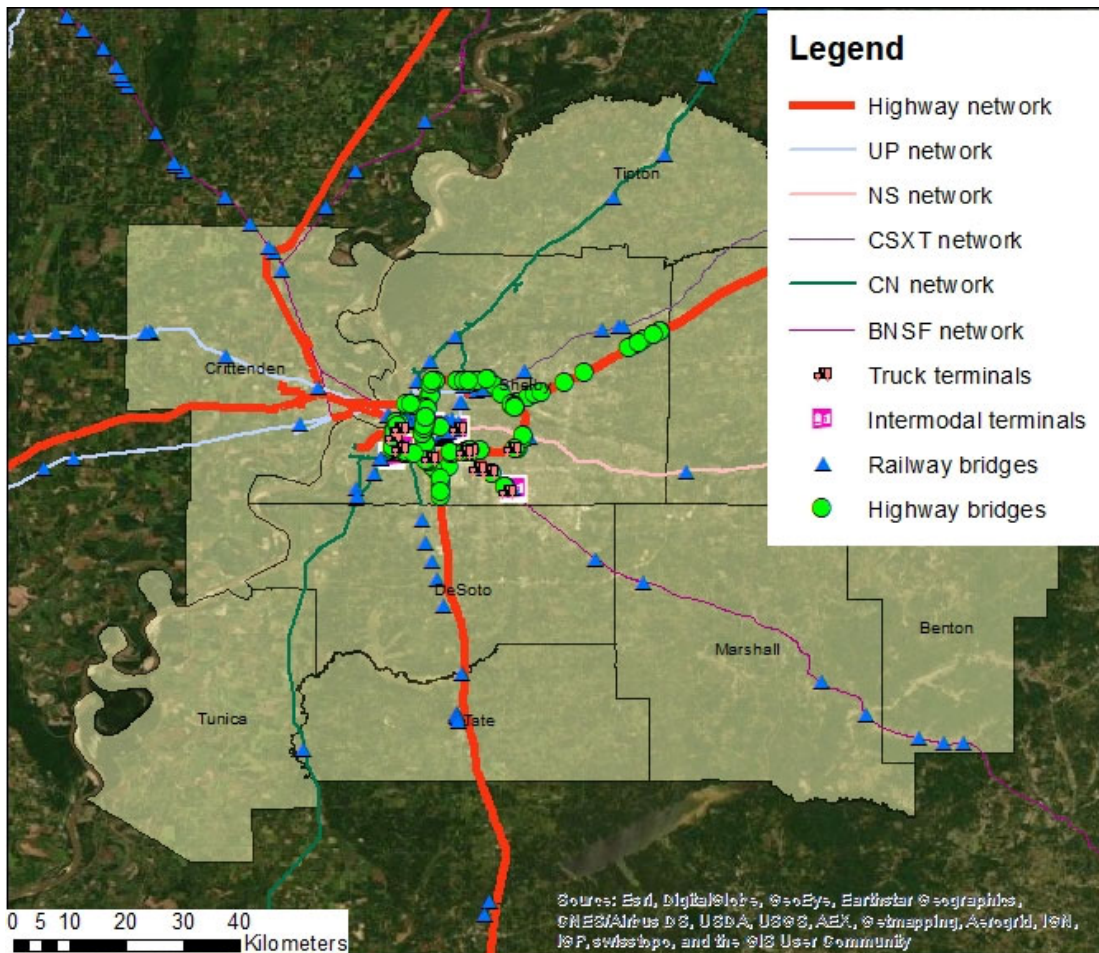
Key Decision Variables	Description
$\Delta y_{bt}$	Binary variable stating if crews are assigned to component $b$ at time $t$
$y_{bt}$	Functionality of component $b$ at time $t$
$x_{ijtk}$	Freight containers on link $ij$ at time $t$ carrying shipment $k$

Gomez, C., González, A. D., Baroud, H., & Bedoya-Motta, C. D. (2019). *Risk Analysis*, 39(9), 1913-1929.

González, A. D., Chapman, A., Dueñas-Osorio, L., Mesbahi, M., & D'Souza, R. M. (2017). *Computer-Aided Civil and Infrastructure Engineering*, 32(12), 991-1006.

Collaboration with Andres Gonzalez (OSU)

# Testbed Regional Intermodal Network

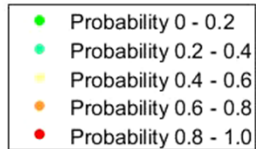
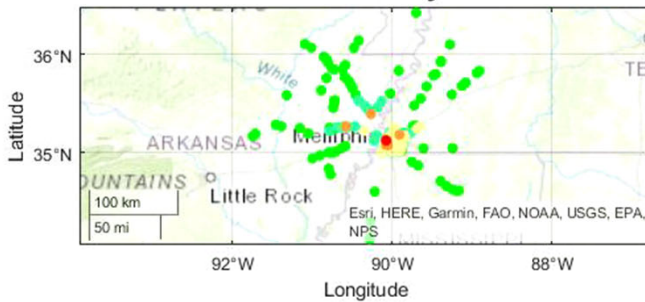


- Testbed regional intermodal network: Memphis, TN
- 5 Class I railroads operate in the region (BNSF, CN, CSXT, NS and UP)
- Hub of freight traffic
  - 153 highway bridges
  - 202 railway bridges
  - 6 intermodal terminals
- Scenario earthquake
  - Magnitude **7.7**
  - Point source at [**35.3°N, 90.3°W**]



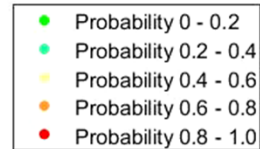
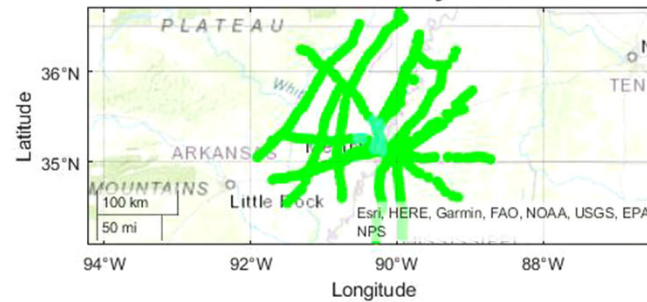
## Bridges – probability of closure

Time = 0 days



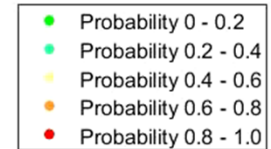
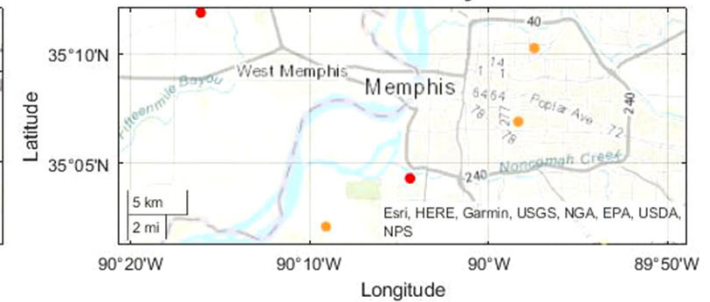
## Roadways and railway tracks – probability of closure

Time = 0 days



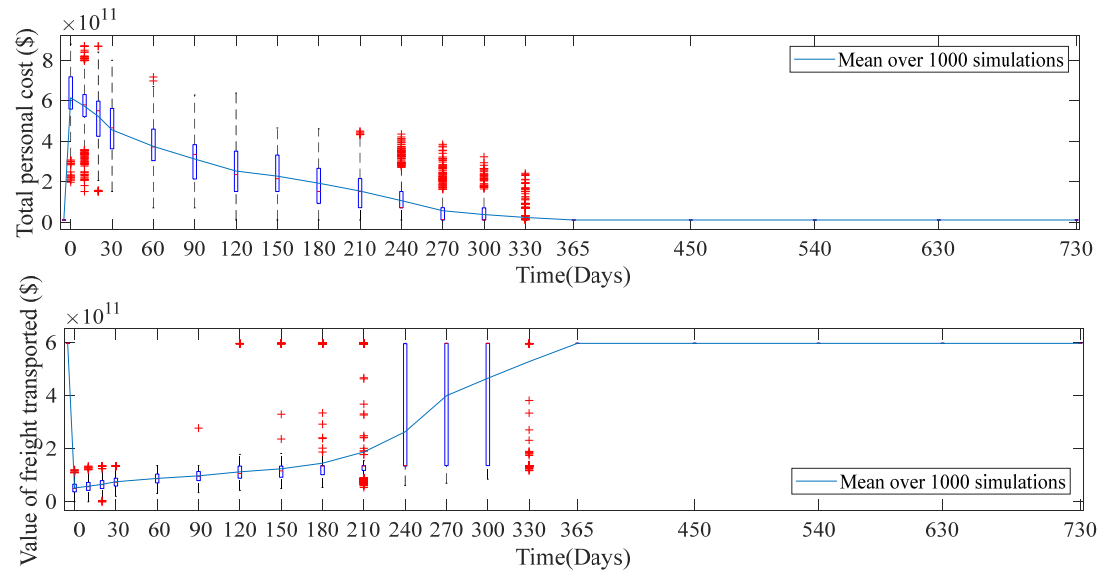
## Intermodal terminals – probability of failure

Time = 0 days



- Monte Carlo Simulation of intermodal network with randomly assigned restoration
- Initial loss of network functionality is dictated by intermodal terminals
- Long term loss of network functionality is dictated by bridges

- Network performance over time



\$ Cost to Shipper

\$ Freight Flow Disrupted

Figure 1. Example intermodal performance metrics

Functionality at any time  $t$  is defined as the **ratio** of some expected **post-event network throughput** to the **pre-event network throughput** (Miller-Hooks et al. 2012).

$$Q(t) = \frac{1}{\sum_{k \in K} d_k^-} E \left( \sum_{k \in K} d_k(t) \right)$$

where  $d_k(t) = V_k \frac{1}{I_k(t)}$

$I_k(t)$  is the travel impedance function for shipment  $k$  at time  $t$

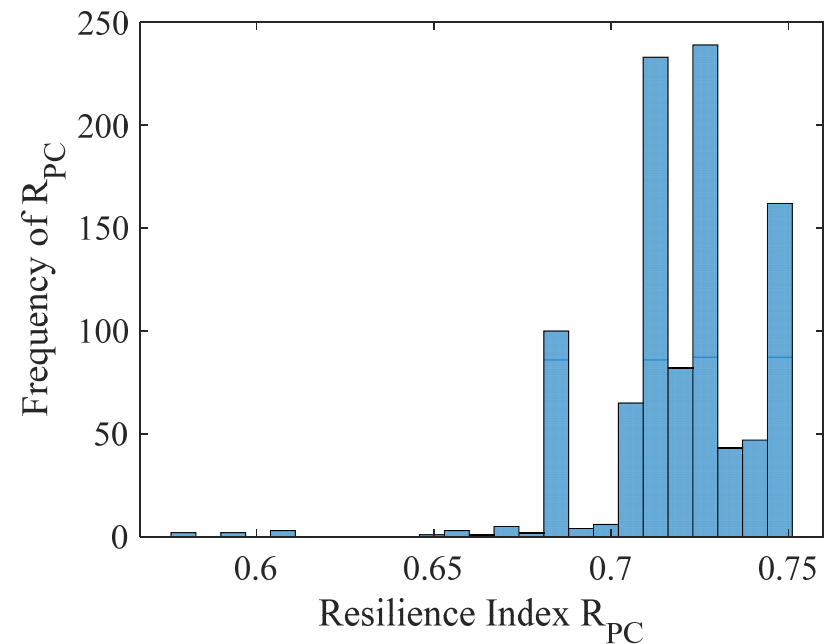
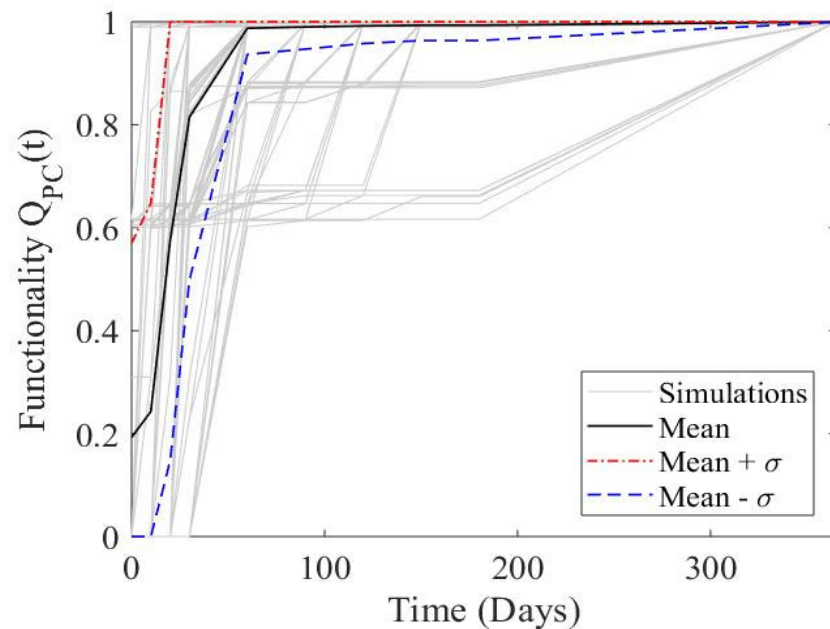
$d_k(t)$  is the **post-event network throughput** for shipment  $k$  at time  $t$

$d_k^-$  is the **pre-event network throughput** for shipment  $w$  at time  $t$

$V_k$  is the value of goods in shipment  $k$

$k \in K$  is the set of all shipments being transported.

Functionality at any time  $t$  is defined as the **ratio** of some expected **post-event network throughput** to the **pre-event network throughput** (Miller-Hooks et al. 2012).



# Optimal Restoration Scheduling

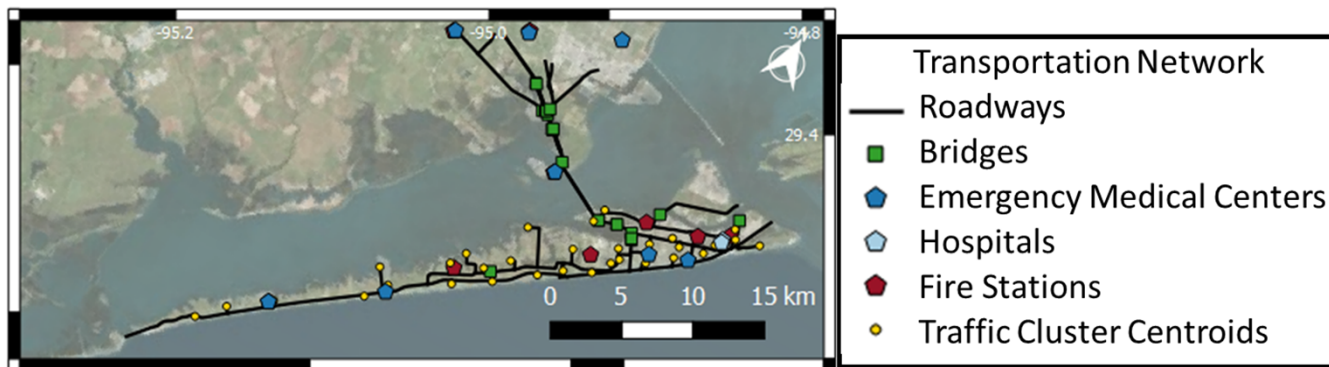
Crew assignment

Restoration of bridge functionality

- The optimal restoration scheduling algorithm restores all the bridges necessary for fulfilling the predefined network demands in the most efficient manner permissible under given resource constraints.

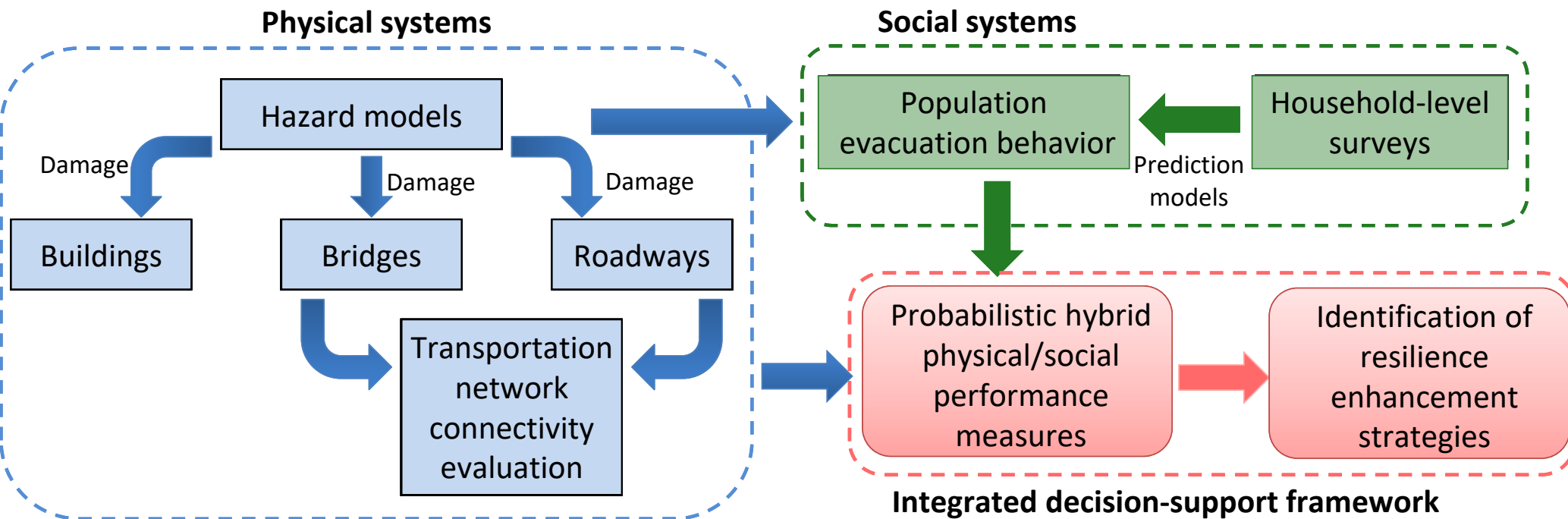


## Role of transportation infrastructure in modeling community resilience under hurricane hazards - Galveston, TX



E. Fereshtehnejad, I. Gidaris, N. Rosenheim, T. Tomiczek, J.E. Padgett, D.T. Cox, S. Van Zandt, W. G. Peacock (*Accepted*). Probabilistic risk assessment of coupled natural-physical-social systems: the cascading impact of hurricane-induced damages to civil infrastructure in Galveston, Texas. *Natural Hazards Review*

- **Integrated resilience framework incorporating physical and social systems**

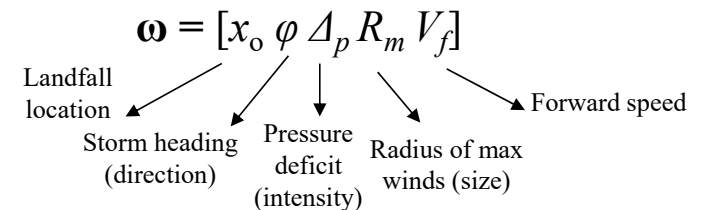
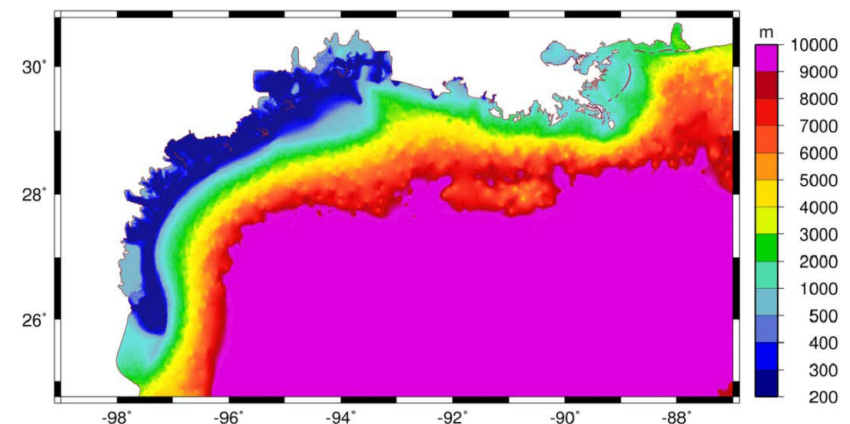


- Surrogate models of joint hazard potential

- Storm surge and wave modeling:
  - High computational cost
  - Prohibitive for probabilistic analysis
- Kriging modeling using a suite of 228 storms (USACE, 2015):

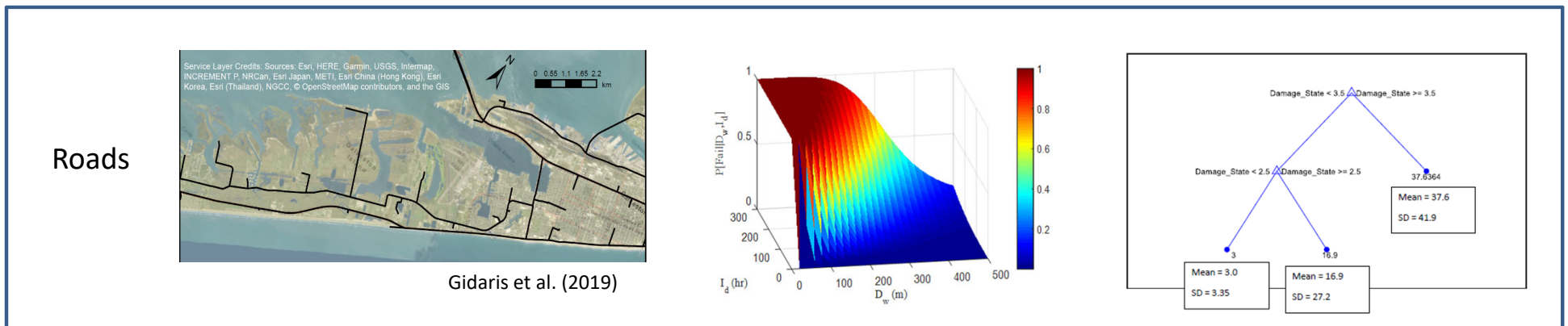
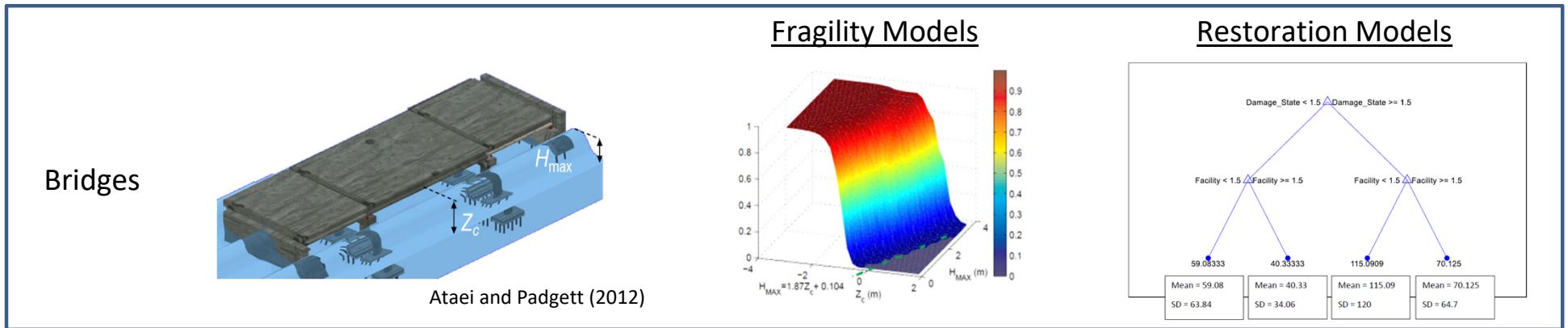
Model	Mean error (%)
Surge height ( $S(\omega)$ )	2.4
Significant wave height ( $H_s(\omega)$ )	7.6
Peak wave period ( $T_p(\omega)$ )	6.8

Texas coast ADCIRC mesh cell size (USACE, 2015)



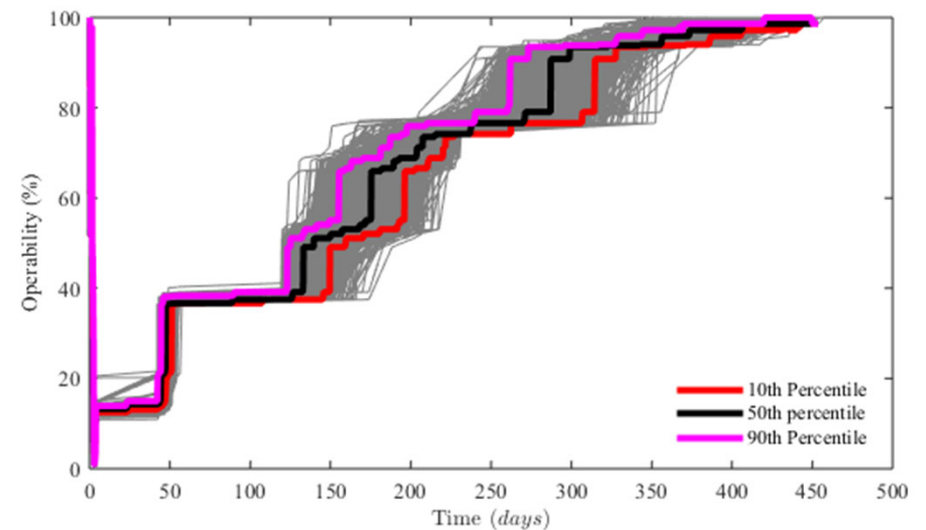
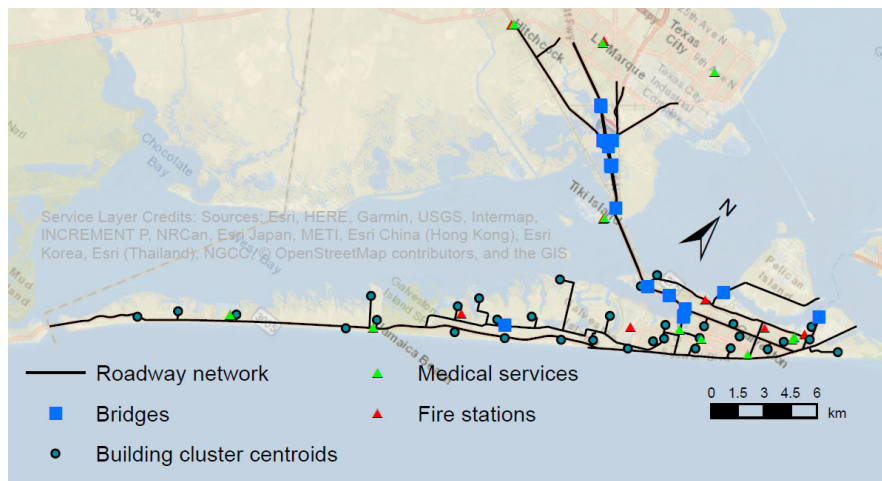
See for example Jia and Taflanidis (2013) or Bernier and Padgett (2019)

- Input models developed for key network components



# Temporal Evolution of Recovery

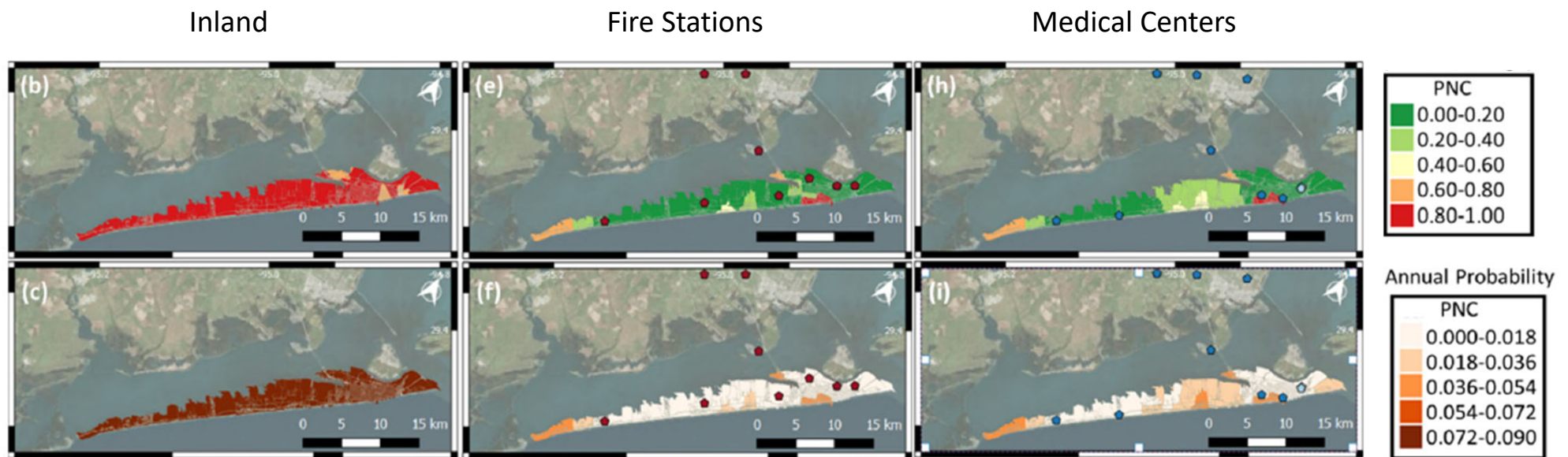
- Recovery is prioritized based on the importance of the component in the overall connectivity of the transportation network.
- The number of available crews for roadway and bridge recovery is variable.





# Connectivity to Emergency Services

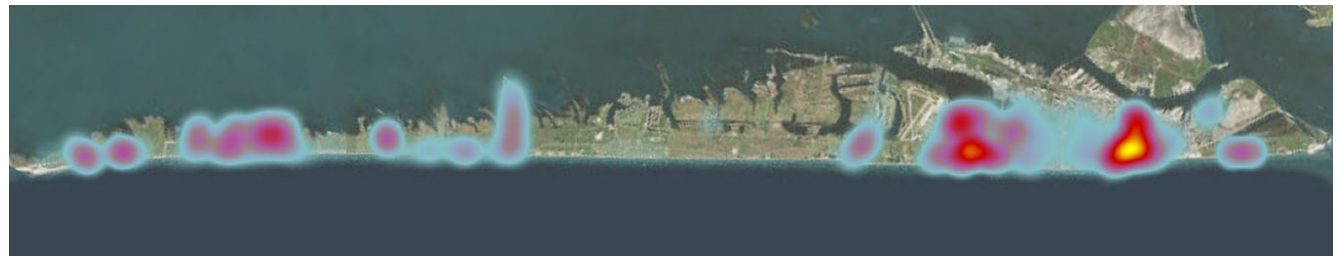
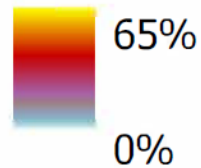
- Probability of disconnection of households from emergency services



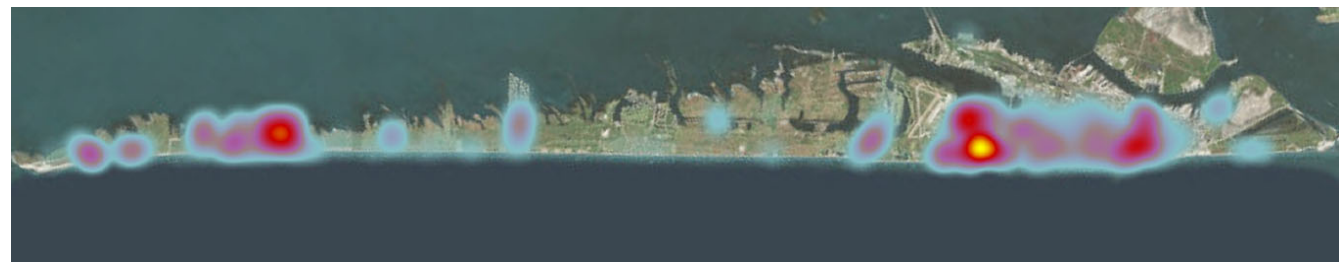
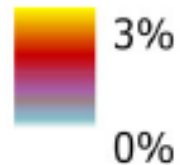
- Integrated physical/social performance metrics

Probability of households being hot: 1) Damaged 2) Non-evacuees inside, and 3) Disconnected from emergency/medical services

Scenario-based

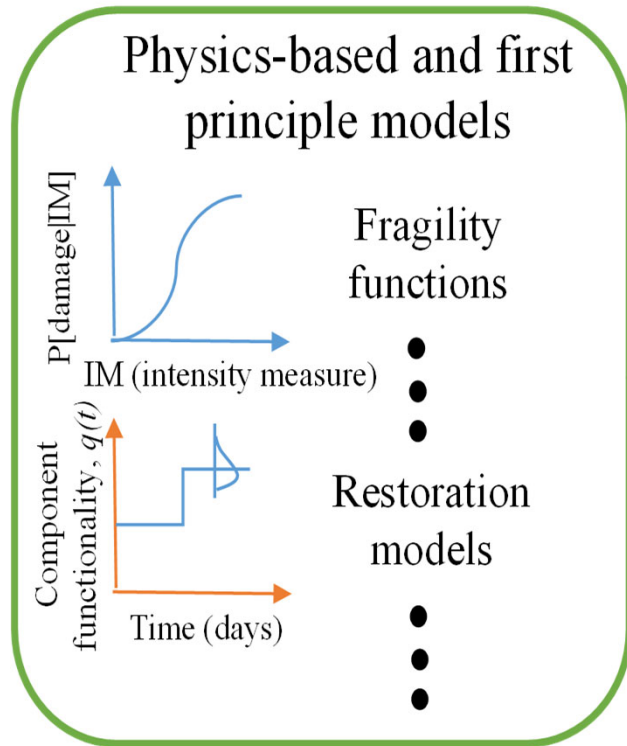


Annual probability

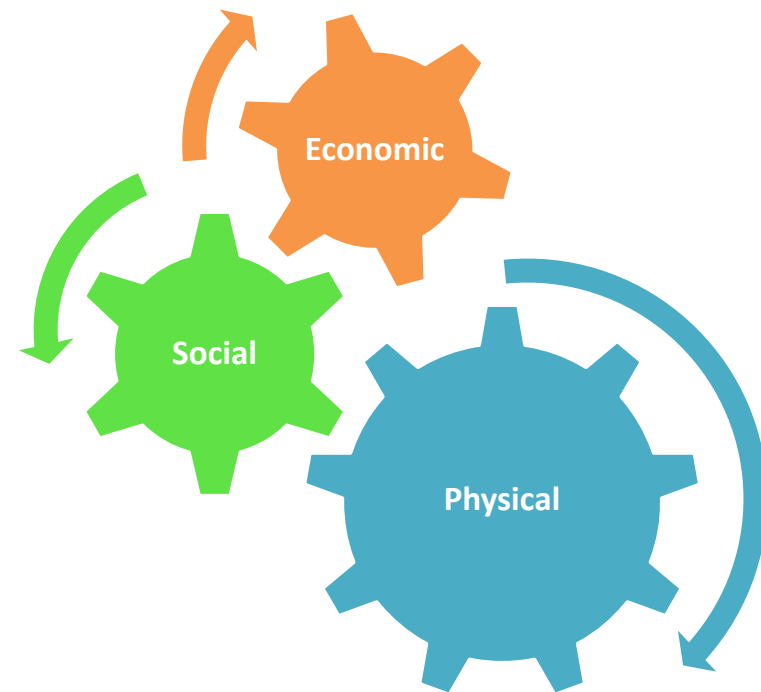




## Address Fragility and Restoration Knowledge Gaps



## Enhance Systems Analysis and Coupling





## Pursue “Smart Resilience”

- Leverage diverse and emerging data sources
  - Resilience and sustainability assessment
  - Situational awareness

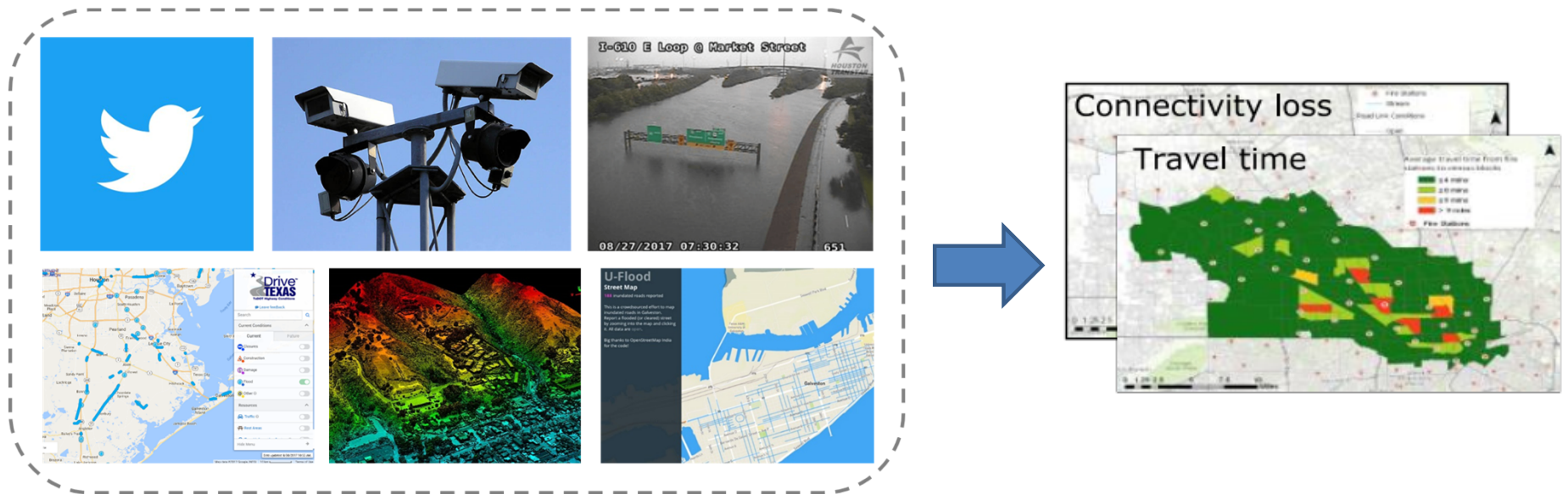


Figure 1. Data sources informing Smart Situational Awareness of Flood impacts on Transportation infrastructure (SSAFT).



## Promote a Culture of Data Sharing

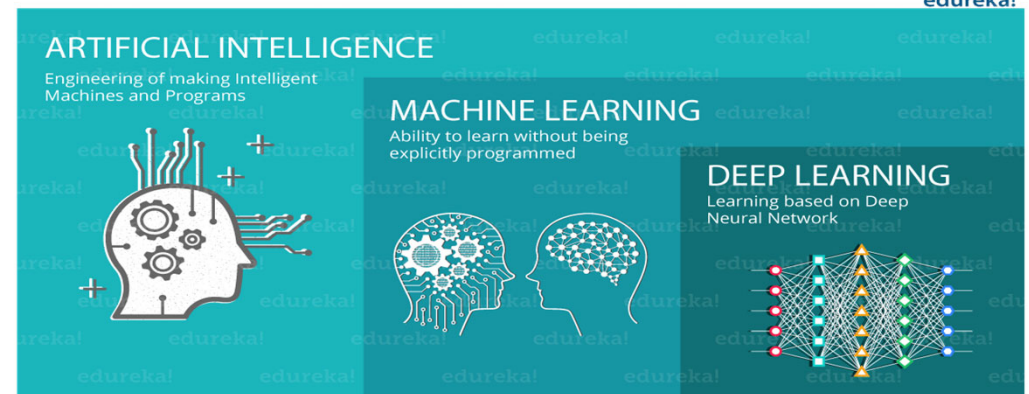
- Embrace coordinated data collection, curation and publication across urban scales and systems
- Leverage cyberinfrastructure platforms like DesignSafe
- Enable future data fusion, AI/ML/DL informing resilience quantification
- Support opensource code and software development



**DESIGNSAFE-CI**   
NHERI: NATURAL HAZARDS ENGINEERING RESEARCH INFRASTRUCTURE

Designsafe-ci.org

edureka!



Bakshi, A. (2019) What is deep learning? Edurica.co



<https://incore.ncsa.illinois.edu/>





# Thank you

## Acknowledgements



SSPEED Center

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