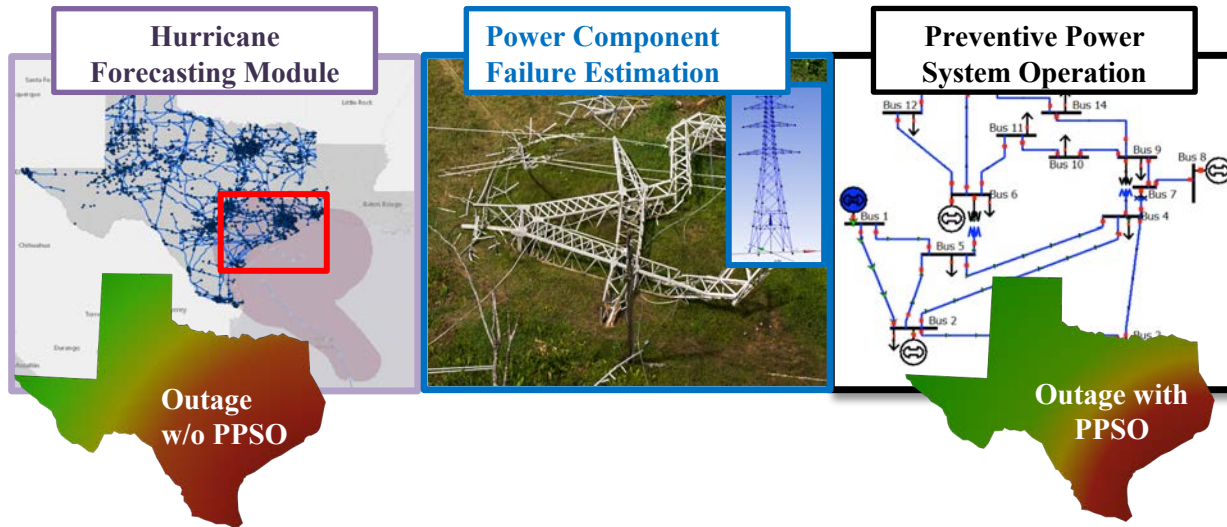


Preventive Power System Operation During Hurricanes



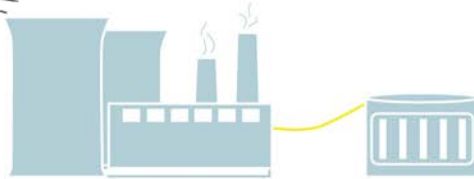
Mostafa Ardakani and Ge Ou
mostafa.ardakani@utah.edu



Hurricane Impacts on Power Systems

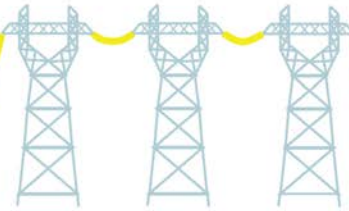


GENERATION



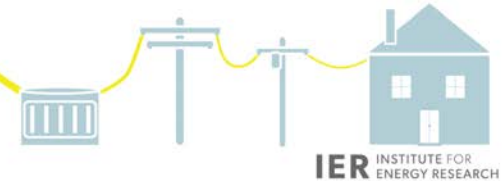
- **Damage level:**
Low
- **Main cause:**
Flooding
- **Wind:**
Rarely an issue

TRANSMISSION



- **Damage level:**
High
- **Main cause:**
Wind force
- **Flooding:**
May aggravate the situation

DISTRIBUTION



- **Damage level:**
High
- **Main cause:**
Wind force
- **Flooding:**
May aggravate the situation



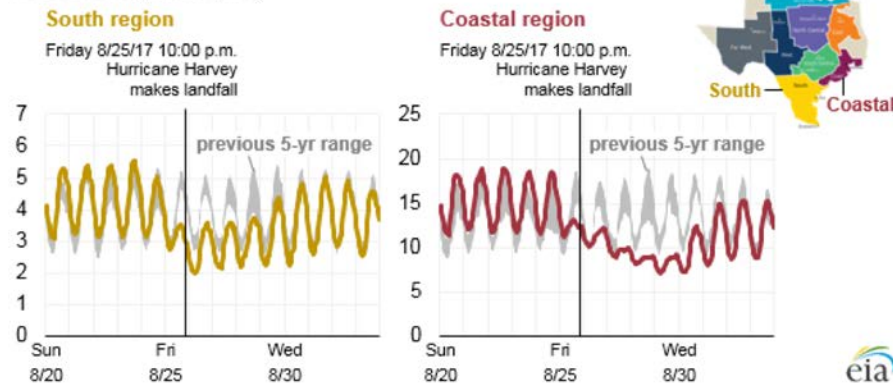
Power Outage Statistics

- Hurricane season of 2017:

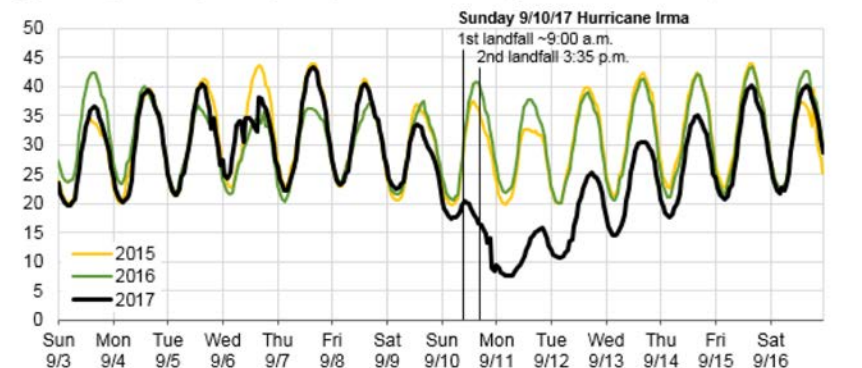
Harvey	Irma	Maria
August	September	September
<ul style="list-style-type: none"> 300,000 customer outages in Texas 	<ul style="list-style-type: none"> 6 Million customer outages in FL (59%) ~1 Million customer outages in GA (22%) 	<ul style="list-style-type: none"> 100% customer outage in PR



Hourly electricity load in ERCOT southern and coastal regions thousand megawatts (MW)



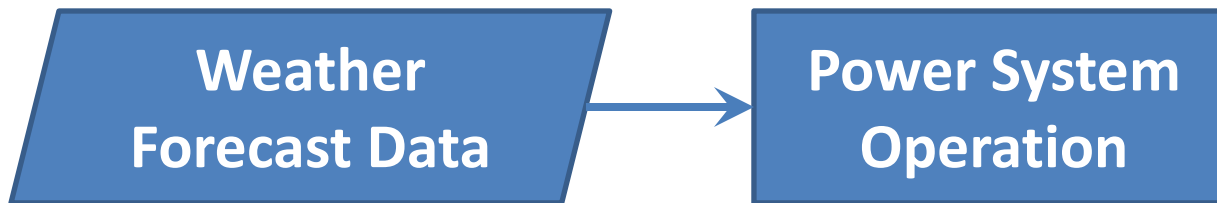
Florida hourly electricity demand, September 2015-2017 gigawatts (previous years aligned by week number & day of week to 2017 data)





Employment of Weather Data

- Would integration of weather data in power system operation reduce the size and duration of power outages?

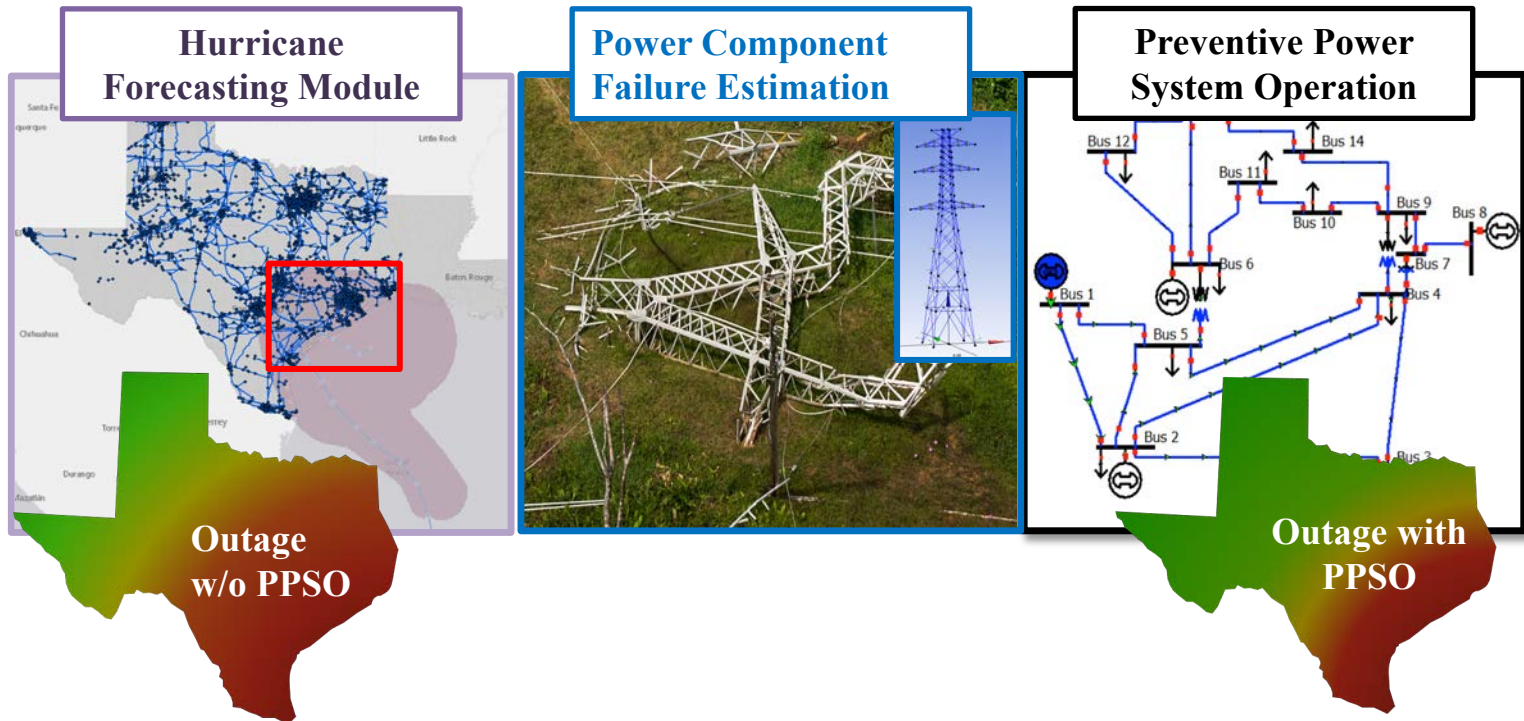


- **Availability of weather data:**
 - System operators have access to weather forecast services
 - In some cases, they also have access to meteorologists onsite
- **Existing technologies:**
 - Pre-storm outage forecast
 - Post-storm restoration planning
 - Long-term grid hardening
 - Emergency operation based on engineering judgement



Proposed Integrated Platform

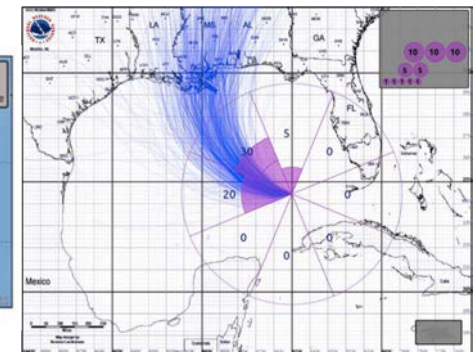
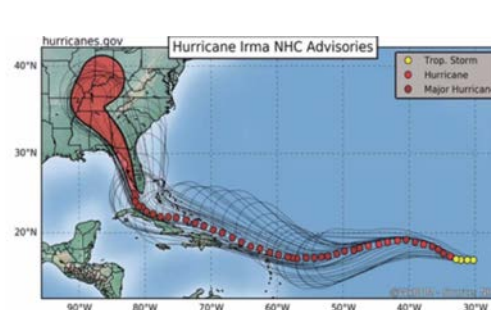
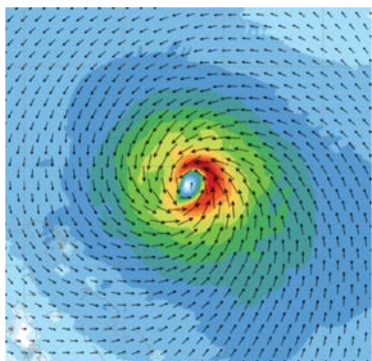
- Systematic integration of weather forecast data in power system operation
 - Translation of weather data into useful information for operation:
component damage probability





Weather Forecasting (Atmospheric Sciences)

- High-resolution wind field modeling
 - 1 Km horizontal
- Hurricane track and movement speed estimation
- Ensemble forecasting
 - Multiple tracks with different probabilities
- Forecast at different time scales
 - 5-day ahead, 48-hr ahead, day-ahead, hour-ahead





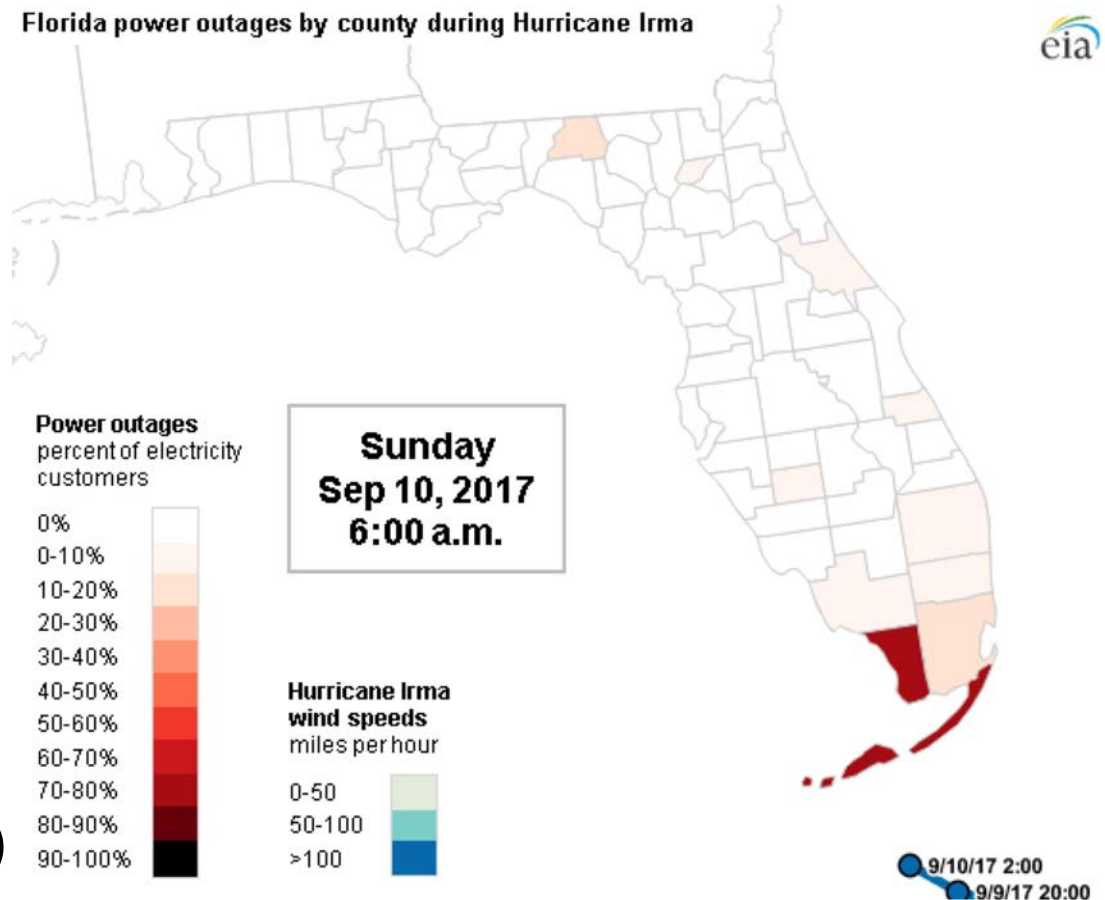
Component Failure Estimation

- Vulnerable components:
 - Transmission lines
 - Distribution lines
- We ignore distribution, because:
 - Distribution network is often radial, which makes preventive operation ineffective
 - Distribution-level damage causes local power outage
 - Transmission-level damage can cause power outage in areas, not directly affected by the hurricane
- Transmission failure is estimated based on:
 - The dynamic loading of the wind
 - Likely important factors that are neglected in our existing model:
 - Debris modeling
 - Flooding and precipitation



Why Focus on Transmission?

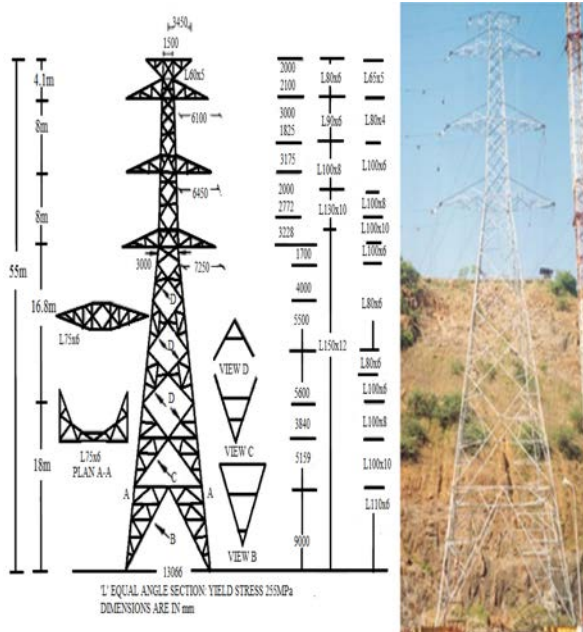
- Power outage in the areas, not in the hurricane track, is due to transmission-level damage.
- Such outages may be manageable, through weather-aware preventive operation.
- Transmission line outages in the past:
 - Harvey: 97 lines (>139 kV)
 - Sandy: 218 lines (>115 kV)



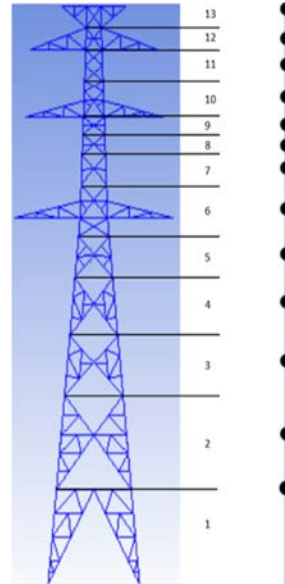


Transmission Failure Estimation I

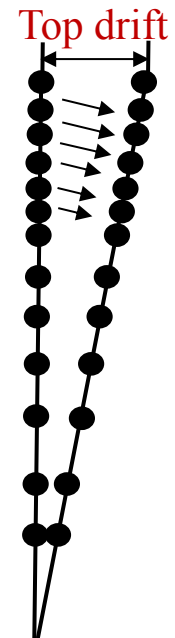
Structural Drawings



Finite Element Modeling



Stability under Dynamic Wind Loading

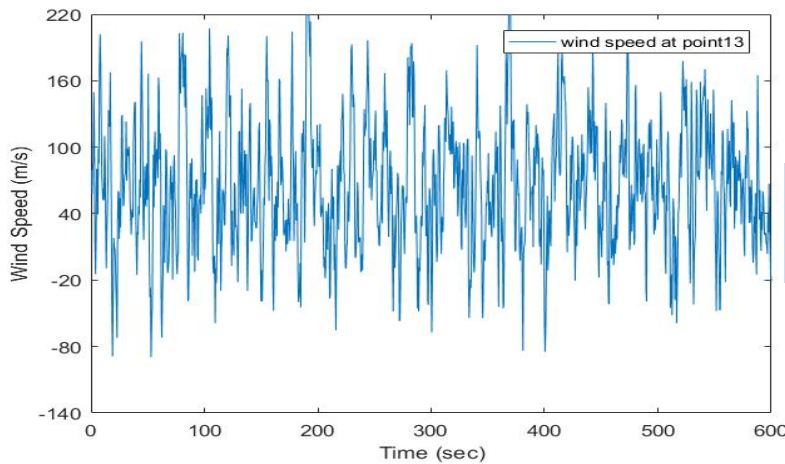


Tower drift
1.5%, 2%, 2.5%, 3%

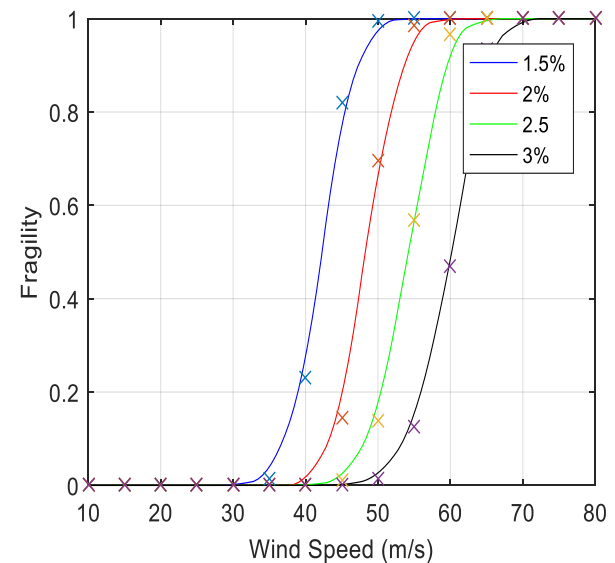
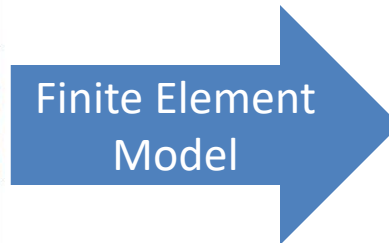


Transmission Failure Estimation II

- Finite element (FEM) models can be computationally demanding
- FEM will be used to develop fragility curves
 - Probabilistic description of failure likelihood, based on the wind speed

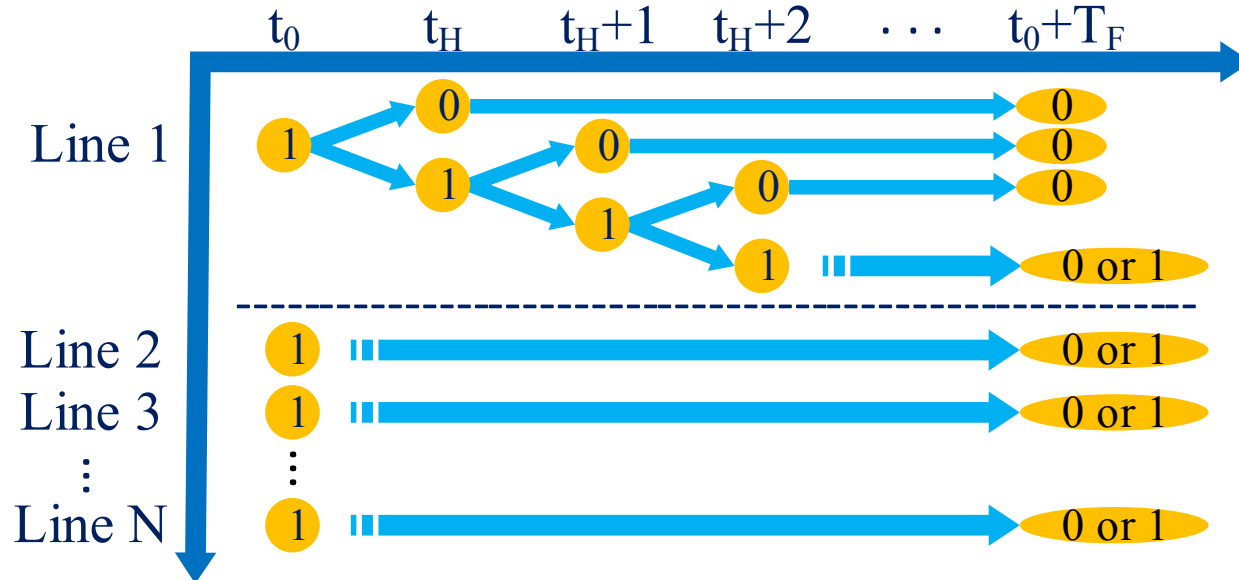


200 random wind speed samples





Line Outage Scenario Generation



The total number of scenarios can be calculated as

$$N_s = (T_F + 1)^{N_{br}}$$

the probability for each scenario can be calculated as

$$p_s = \prod_{k=1}^{N_{br}} \left(p_{k,t_k} \prod_{t=t_H}^{t_k-1} (1 - p_{k,t}) \right).$$

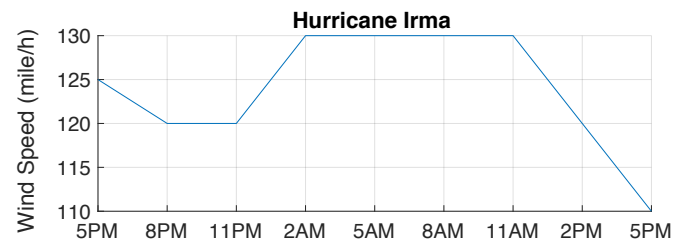
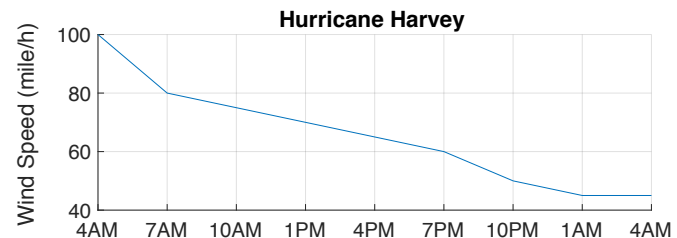
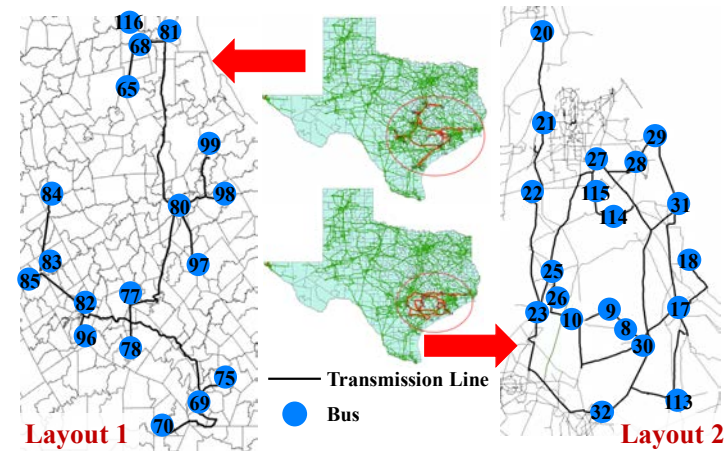


Preventive Power System Operation

- Stochastic optimization
 - Scenarios: unplanned line outages
 - Load shedding is penalized by a large penalty factor
- Computational needs:
 - Stochastic optimization can be computationally demanding
 - Scenario reduction can help reduce the computational burden:
 - Elimination of unlikely scenarios, below a threshold
 - Elimination of inconsequential scenarios
- The solution will change the dispatch to minimize load shedding

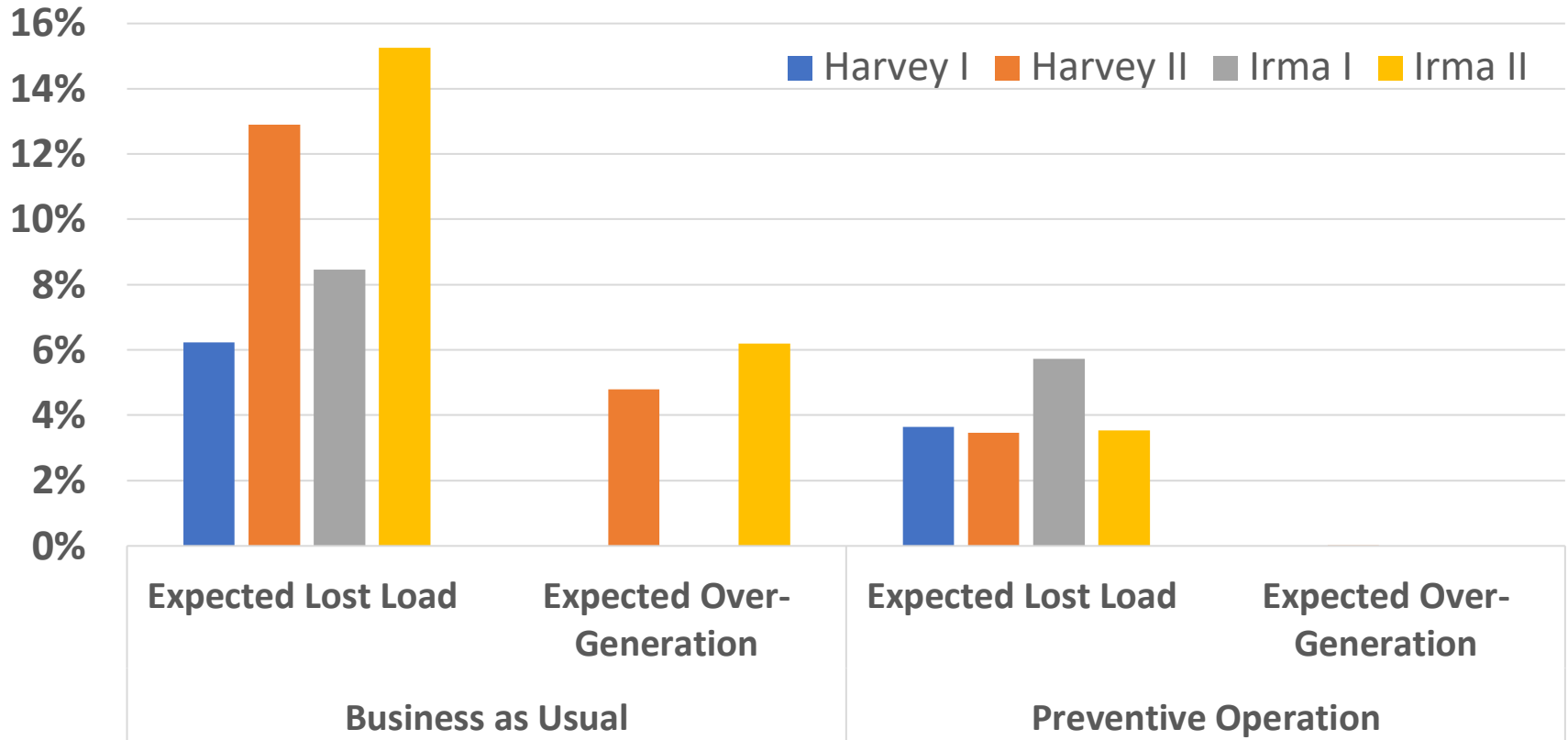
Case Study Setup

- Test system: IEEE 118-bus
- Two layouts:
 - Areas affected by the hurricane
- Two synthesized hurricanes:
 - Harvey
 - Irma



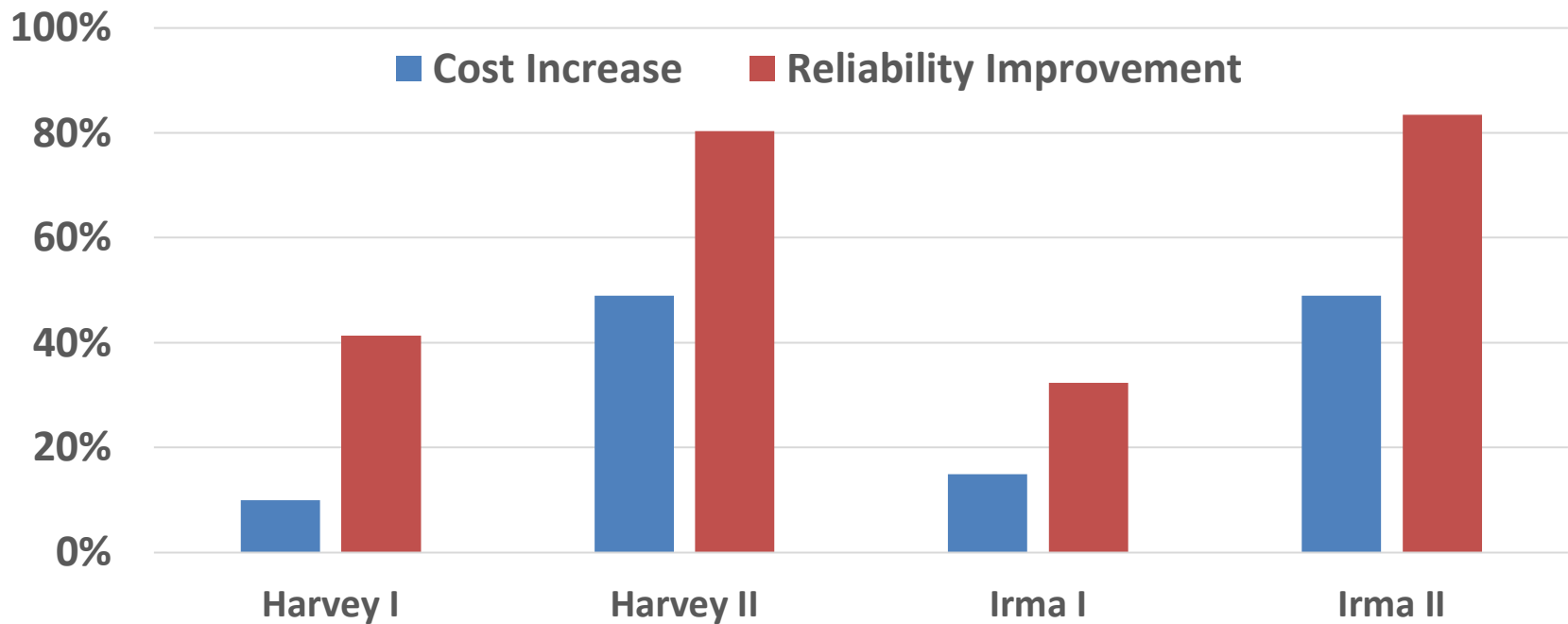


Results: Reliability Improvement





Results: Reliability versus Cost





Conclusions

- Predictable weather-related natural hazards are the cause of about half of the blackouts in the US.
- Weather forecast data can be used to estimate component damage likelihood.
- Component damage estimations can be used to guide preventive operation.
- The simulation results confirms the effectiveness of our integrated platform in substantially reducing power outages.
- Appropriate integration of weather forecast data within power system operation can enhance system reliability.



Discussion and Future Work

- Stochastic optimization was used in this work:
 - Computationally demanding
 - Power system operation software by in large use deterministic models
 - We are currently working to develop proxy deterministic rules that:
 - Capture the majority of stochastic optimization
 - Do not substantially add to the computational burden
- The framework is general and can be applied to other weather hazards such as ice storms.



Acknowledgement

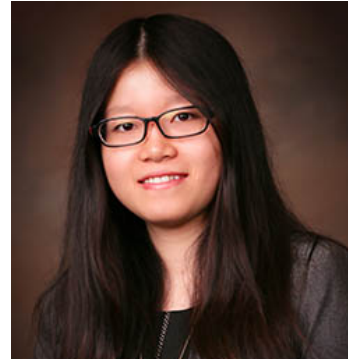
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Funding
Agency



U S T A R

UTAH'S TECHNOLOGY CATALYST



References and Further Reading

- Y. Sang, J. Xue, M. Sahraei-Ardakani, and G. Ou, “Effective Scenario Selection for Preventive Stochastic Unit Commitment during Hurricanes,” *2018 IEEE International Conference on Probabilistic Methods Applied to Power Systems (PMAPS)*, Boise, ID, USA.
- M. Sahraei-Ardakani and Ge Ou, “Day-Ahead Preventive Scheduling of Power Systems During Natural Hazards via Stochastic Optimization,” *IEEE PES General Meeting 2017*, Chicago, IL, USA.
- Y. Sang, J. Xue, M. Sahraei-Ardakani, and G. Ou, “Reducing Hurricane-induced Power Outages through Preventive Operation,” *working paper, available at: <https://ardakani.ece.utah.edu/wp-content/uploads/sites/75/2018/05/HurricanePaper.pdf>*

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Thank You!