

# Resilience Evaluation for Water Distribution Networks

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# Motivation and Background

- Water distribution systems face multiple challenges.
- Water utilities need to be able to predict how their system will perform during disruptive events and understand how to best absorb, recover from, and more successfully adapt.
- Simulation and analysis tools can help water utilities explore how their network will respond to expected, and unexpected, events.

Potential Hazards	Potential Impacts
Natural Disasters	Pipe Break
<ul style="list-style-type: none"><li>• Drought</li><li>• Earthquakes</li><li>• Floods</li><li>• Hurricanes</li><li>• Tornados</li><li>• Tsunamis</li><li>• Wildfires</li><li>• Winter Storms</li></ul>	Other Infrastructure Damage/Failure
Terrorist Attacks	Power Outage
Cyber Attacks	Service Disruption (source water, treatment, distribution, or storage)
Hazardous Materials Release	Loss of Access to Facilities/Supplies
Climate Change	Loss of Pressure/Leaks
	Change in Water Quality
	Environmental impacts
	Financial impacts (e.g., loss of revenue, repair costs)
	Social Impacts (e.g., loss of public confidence, reduced workforce)



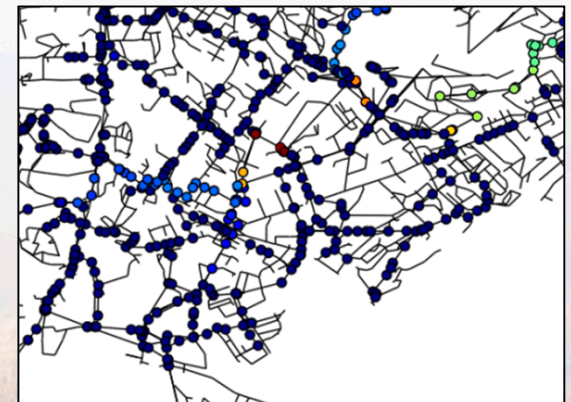
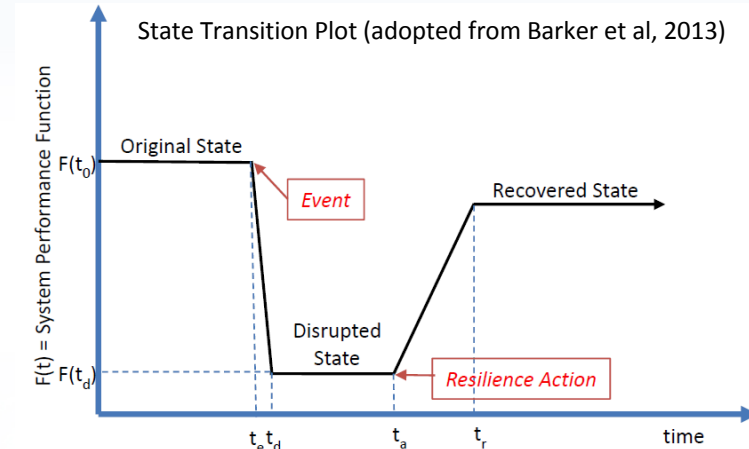
# Motivation and Background

- What is resilience?
- PDD 21 “Critical Infrastructure Security and Resilience” establishes national policy to proactively increase infrastructure resilience.
- Resilience of the water sector is tightly linked to the resilience of other critical infrastructure such as energy, food and agriculture, health care and public health.
- Resilience of drinking water systems refers to the design, maintenance, and operations of water infrastructure that limits the effects of disasters and enables rapid return to normal delivery of safe water to customers.



# Quantifying Resilience

- Numerous metrics have been suggested to quantify reliability, robustness, redundancy, and security for water distribution networks
  - Topographic metrics
  - Hydraulic metrics
  - Water quality metrics
- State transition plots graphically represents the meaning of resilience
  - System performance function, event, and resilience action must be clearly defined
  - Metrics used to define reliability, robustness, redundancy, and security might prove useful to measure resilience
  - Resilience is typically defined as a system measure, but could be measured for individual components of the network.





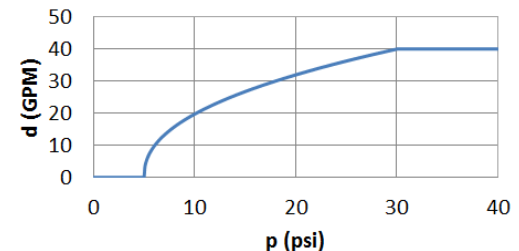
# Hydraulic Simulation of Disruptive Events

- Demand-driven simulation (such as with EPANET) might not be adequate to simulate hydraulic capacity during some disruptive events
- Pressure Dependent Demand Model (Wagner, 1988; Rossman whitepaper, 2015)
  - Demand at a node,  $d$ , depends on the pressure head  $p$  available at the node
  - Input parameters = minimum pressure ( $P_o$ ) and nominal pressure ( $P_f$ )

$$\begin{aligned}d &= D_f && \text{for } p \geq P_f \\d &= D_f \left( \frac{p - P_o}{P_f - P_o} \right)^{1/e} && \text{for } P_o < p < P_f \\d &= 0 && \text{for } p \leq P_o\end{aligned}$$

## Example

$P_o = 5$  psi  
 $P_f = 30$  psi  
 $D_f = 40$  GPM



- PYOMO interface is used to solve non linear equations
  - Python interface
  - IPOPT nonlinear solver
  - Smoothing functions near the boundaries



# Power Outage Case Study

- Simulate network wide and isolated power outages by changing operations of the pumps
  - $P_o = 0$ ,  $P_f = 80\%$  minimum pressure at each node when run under normal operations.
  - When power is off at a pump station, the pumps act as a by-pass. Water is not allowed to flow back into the reservoir. Pumps are closed if flow  $< 1e-9$  m<sup>3</sup>/s
  - Tanks can fill by gravity when power is off. Tanks are allowed to empty completely.
- Track the ability to deliver expected volume of water as a function of time and location.
- Disruptive Events
  - 3000 node network contains 61 pumps at 22 pump stations
  - Power outage starts at 8 AM
  - Assume back-up generation has failed, or run its course
- Resilience Action
  - Restore power 5,10,15 hours after outage



# Simulation and Analysis

- System performance function,  $F(t)$ 
  - Fraction of delivered volume (FDV) (Ostfeld, 2002)
  - For a particular scenario, the fraction is computed over all nodes.

$$F(t) = FDV(i, t) = \frac{\sum_n^N RV_{i,n,t}}{\sum_n^N EV_{i,n,t}} \quad \forall i \text{ in } I, t \text{ in } T$$

where:

RV = requested volume

EV = expected volume

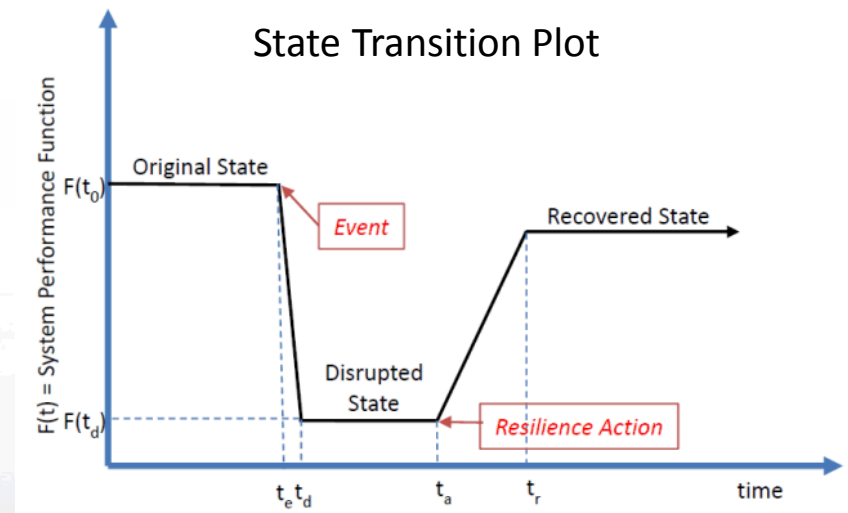
I = set of scenarios

N = set of consumer nodes

T = set of simulation time steps

- Characteristics

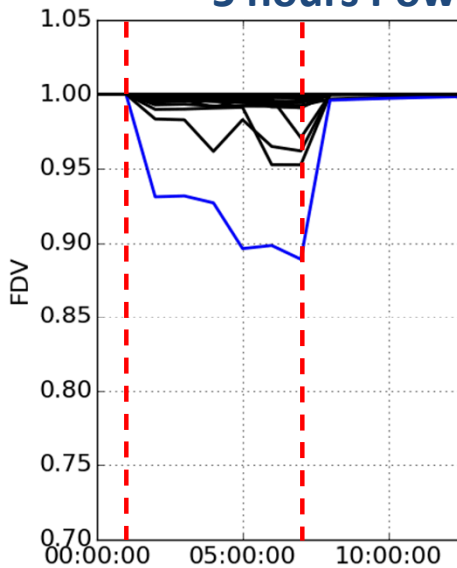
- Minimum state =  $F(t_d) = \min(F(t))$
- Time to disruption =  $t_d - t_e$
- Time to recover =  $t_r - t_d$
- Number and location of disrupted nodes



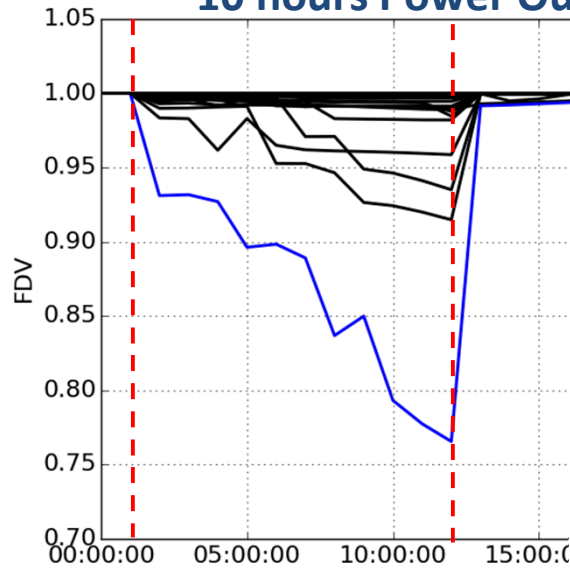
# Results

- State Transition Plot using Fraction of Delivered Volume per scenario

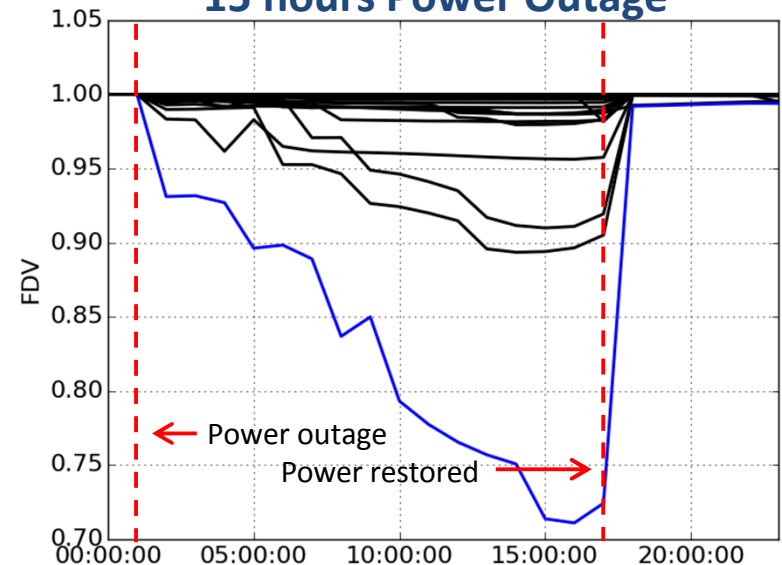
### 5 hours Power Outage



### 10 hours Power Outage



### 15 hours Power Outage



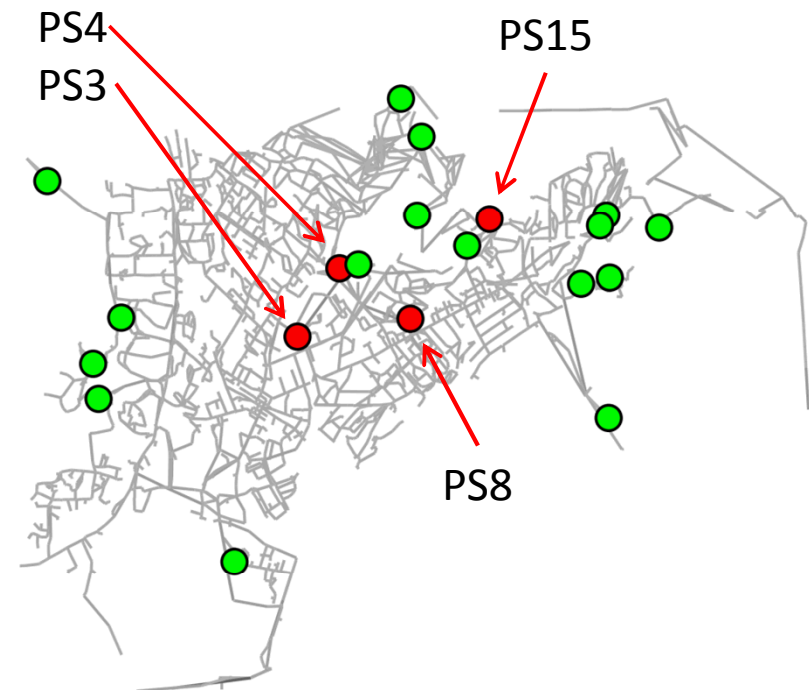
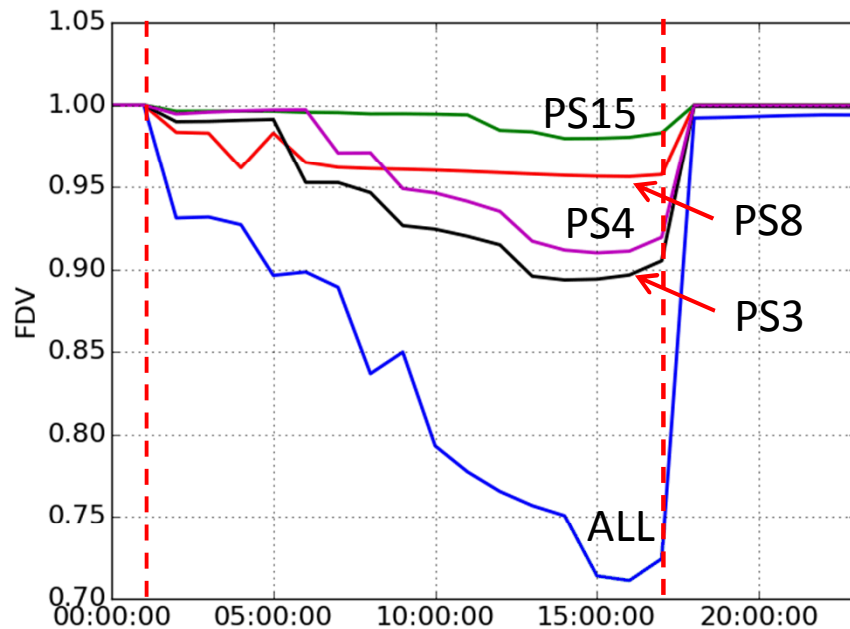
— Power outage at single pump station  
— Network wide power outage  
- - - Time of outage/restoration





# Results

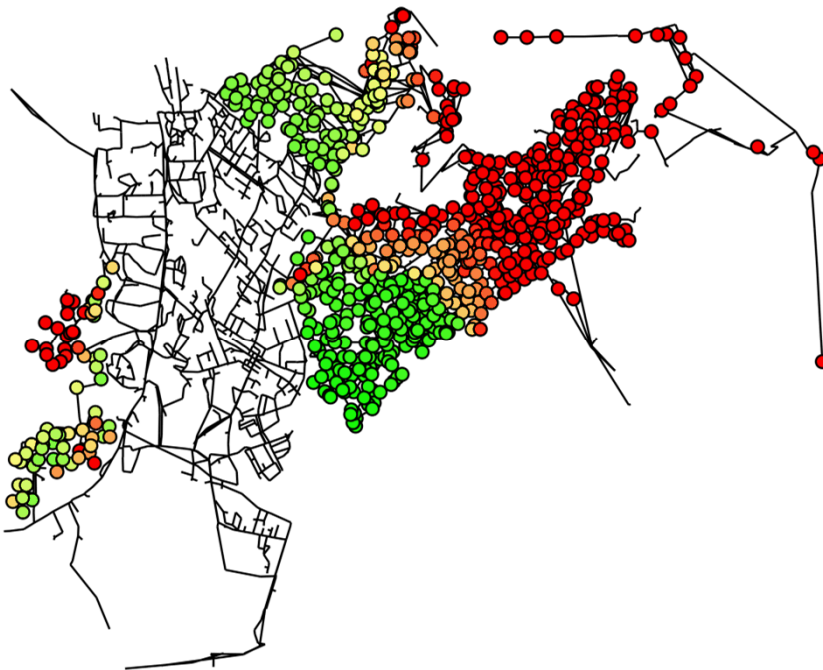
- 15 hour power outage
- State transition plots with minimum disruption < 98%



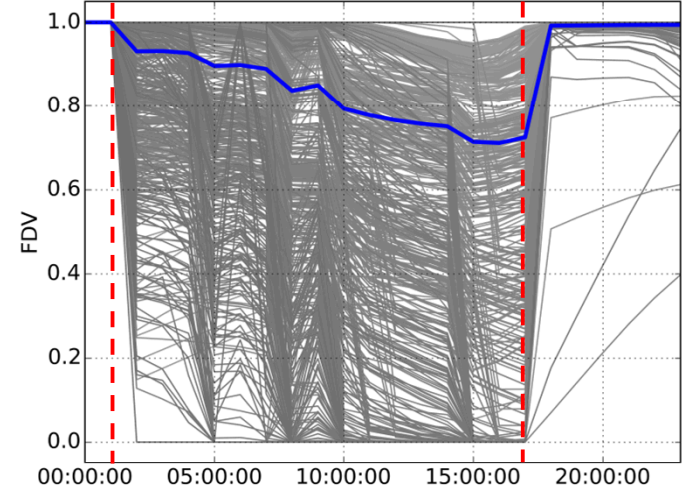
# Results

- 15 hour network wide power outage
  - 24 tanks fall below 2m
  - 53% of consumer nodes deliver less than 98% expected volume

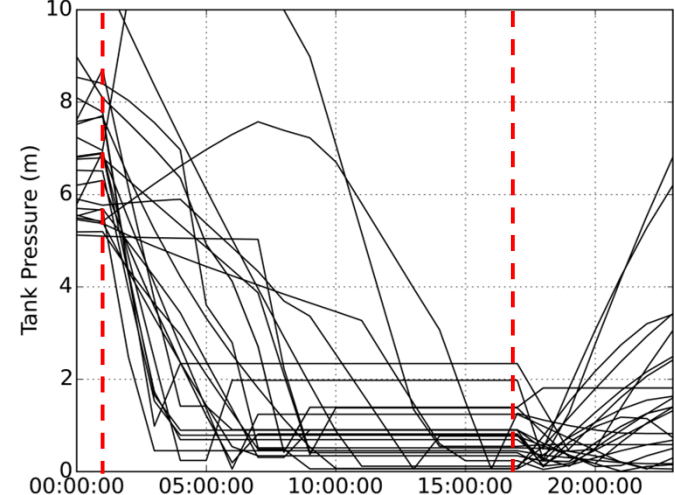
Minimum State (FDV per node)



FDV, network average and per node



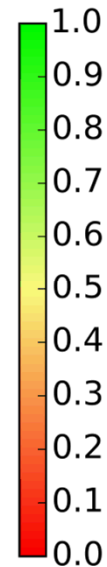
Tank levels that fall below 2m



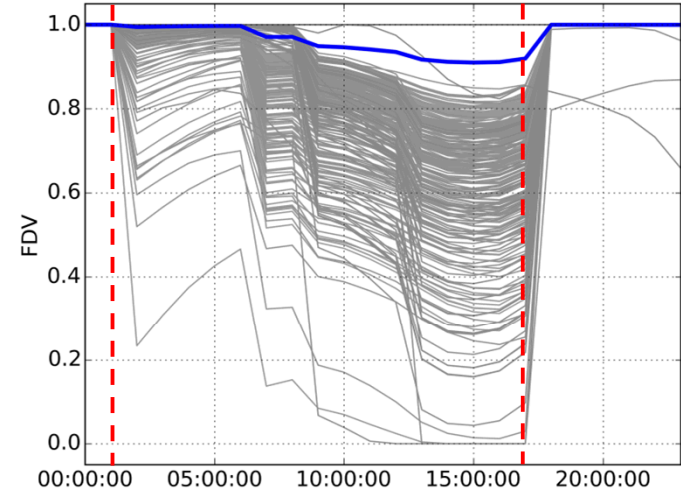
# Results

- 15 hour power outage at pump station 4
  - 4 tanks fall below 2m
  - 24% consumer nodes deliver less than 98% expected volume
  - Pump station 3 and 4 have similar response

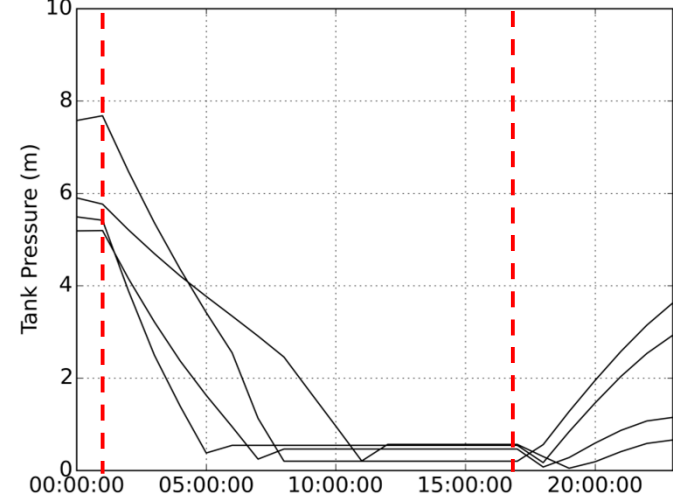
Minimum State (FDV per node)



FDV, network average and per node



Tank levels that fall below 2m



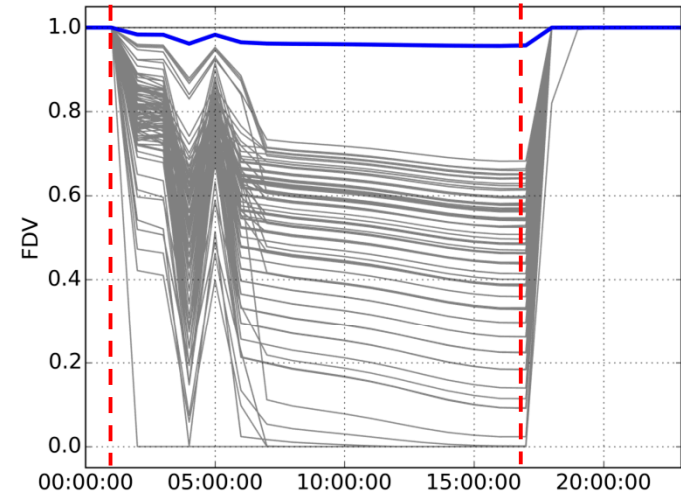
# Results

- 15 hour power outage at pump station 8
  - 2 tanks fall below 2m
  - 8% consumer nodes deliver less than 98% expected volume

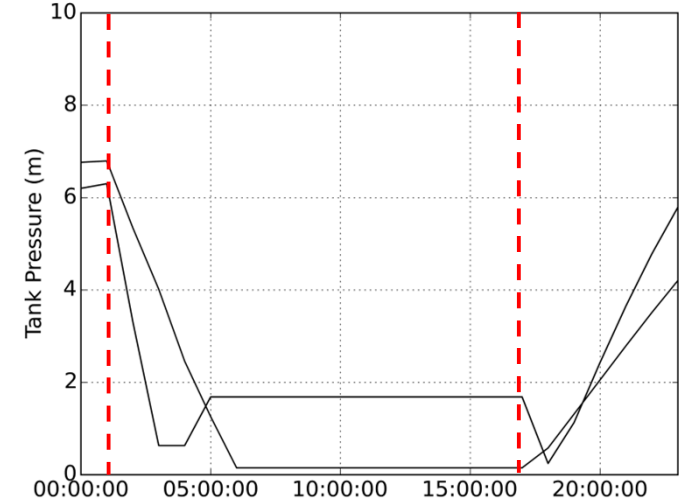
Minimum State (FDV per node)



FDV, network average and per node



Tank levels that fall below 2m

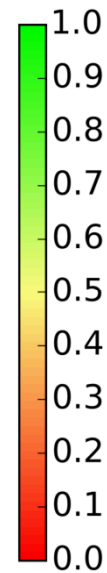




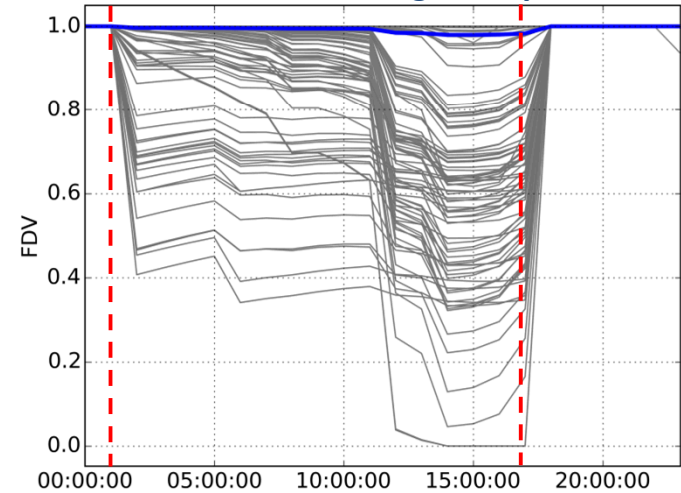
# Results

- Power outage at pump station 15
  - 4 tanks fall below 2m
  - 6% consumer nodes deliver less than 98% expected volume

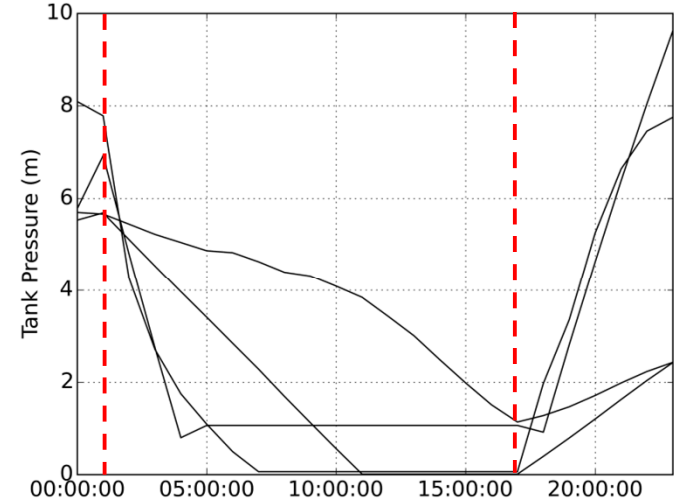
Minimum State (FDV per node)



FDV, network average and per node



Tank levels that fall below 2m



# Summary

- What does it mean to be resilient in face of a power outage?
  - Reduce the magnitude and duration of disrupted service
  - Identify critical pumps and other infrastructure for backup generation
  - Understand where and when customers will be impacted
  - Identify worst case scenarios
  - Simulate response and adaptation strategies
  - Modify network operations to be better prepared for future events
- Future steps
  - Test other resilience metrics as the system performance function
  - Model other types of disruptive events
  - Expand case studies to include a wide range of expected and unexpected events
  - Include information about critical infrastructure into the analysis



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