



Outage Prediction

Using weather data to predict outage risk and assist resident preparation



Joanna Huang, Calvin Kao, Justin Plumley
December 19, 2017

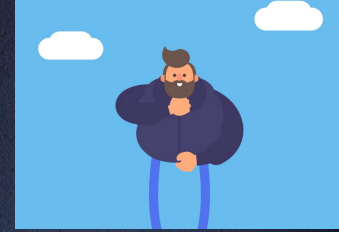
Background:

- Power outages due to downed power lines are a frequent occurrence in the US
- Extreme weather events result in economic, social, and physical disruptions
 - Aging power grid
 - Potential cost of \$150 billion per year



Project objective:

- Use weather and power datasets to predict outages and their potential magnitude to inform residents and assist in preparation



Acquisition and Organization of Information for Analytics

NOAA

(National Oceanic and Atmospheric Administration)

Provides weather history over a long period of time and / or locations

Data fields: sky conditions, visibility, temperature, humidity, wind speed, pressure

Data size: up to 13,000+ observations/year (per city)

<https://www.ncdc.noaa.gov/>

| STATION | STATION_NAME | ELEVATION | LATITUDE | LONGITUDE | DATE | REPORT TPYE | HOURLY SKY CONDITI ONS | HOURLY VISIBILITY | HOURLY DRY BULB TEMPF | HOURLY WET BULB TEMPF | HOURLY Dew Point TempF | HOURLY Relative Humidity | HOURLY Wind Speed | HOURLY Wind Direction | HOURLY Wind Gust Speed | HOURLY Station Pressure |
|------------|------------------|-----------|----------|-----------|----------------|----------------|---------------------------------|----------------------|-----------------------------|--------------------------------|---------------------------------|--------------------------------|-------------------------|-----------------------------|---------------------------------|-------------------------------|
| WBAN:14765 | PROVIDENCE RI US | 16.8 | 41.7225 | -71.4325 | 12/1/2015 0:51 | FM-15 | CLR:00 | 10 | 31 | 30 | 27 | 85 | 3 | 300 | | 30.32 |
| WBAN:14765 | PROVIDENCE RI US | 16.8 | 41.7225 | -71.4325 | 12/1/2015 1:00 | FM-12 | | | 31 | 30 | 27 | 85 | 3 | 300 | | 30.32 |
| WBAN:14765 | PROVIDENCE RI US | 16.8 | 41.7225 | -71.4325 | 12/1/2015 1:51 | FM-15 | CLR:00 | 10 | 30 | 29 | 26 | 85 | 0 | 0 | | 30.32 |
| WBAN:14765 | PROVIDENCE RI US | 16.8 | 41.7225 | -71.4325 | 12/1/2015 2:51 | FM-15 | CLR:00 | 10 | 30 | 29 | 27 | 88 | 5 | 360 | | 30.32 |
| WBAN:14765 | PROVIDENCE RI US | 16.8 | 41.7225 | -71.4325 | 12/1/2015 3:51 | FM-15 | CLR:00 | 10 | 30 | 29 | 27 | 88 | 8 | 360 | | 30.3 |
| WBAN:14765 | PROVIDENCE RI US | 16.8 | 41.7225 | -71.4325 | 12/1/2015 4:51 | FM-15 | CLR:00 | 10 | 29 | 28 | 26 | 89 | 0 | 0 | | 30.3 |
| WBAN:14765 | PROVIDENCE RI US | 16.8 | 41.7225 | -71.4325 | 12/1/2015 5:51 | FM-15 | CLR:00 | 10 | 29 | 28 | 26 | 89 | 3 | 360 | | 30.3 |
| WBAN:14765 | PROVIDENCE RI US | 16.8 | 41.7225 | -71.4325 | 12/1/2015 6:51 | FM-15 | SCT:04 50 BKN:07 250 | 10 | 30 | 29 | 27 | 88 | 6 | 360 | | 30.29 |

Acquisition and Organization of Information for Analytics

EIA Electric Power

Major Disturbances and Unusual Occurrences, Year-to-Date 2016

Data fields: Year, Month, Event Date and Time, Restoration Date and Time, Duration, Utility/Power Pool, NERC Region, Area Affected, Type of Disturbance, Loss (megawatts), Number of Customers Affected

Data size: 50-100 observations a year

https://www.eia.gov/electricity/monthly/epm_table_grapher.php?t=epmt_b_1

| Year | Month | Event Date and Time | Restoration Date and Time | Duration | Utility/Power Pool | NERC Region | Area Affected | Type of Disturbance | Loss (megawatts) | Number of Customers Affected |
|------|-------|---------------------|---------------------------|----------------------|---|-------------|--|--|------------------|------------------------------|
| 2016 | 1 | 01/10/2016 8:46 PM | 01/11/2016 5:25 AM | 8 Hours, 39 Minutes | ISO New England | NPCC | Maine: Connecticut: Massachusetts: Vermont: New Hampshire: Rhode Island: | Loss of electric service to more than 50,000 customers for 1 hour or more-Weather | Unknown | 59859 |
| 2016 | 1 | 01/22/2016 3:52 PM | 01/24/2016 12:30 PM | 44 Hours, 38 Minutes | Duke Energy Progress | SERC | North Carolina: South Carolina: | Loss of electric service to more than 50,000 customers for 1 hour or more-Weather | Unknown | 150000 |
| 2016 | 1 | 01/23/2016 7:49 AM | 01/23/2016 9:05 AM | 1 Hours, 16 Minutes | FirstEnergy Corp. Jersey Central Power & Light | RFC | New Jersey: | Loss of electric service to more than 50,000 customers for 1 hour or more-Weather | Unknown | 50900 |
| 2016 | 2 | 02/05/2016 11:21 AM | 02/06/2016 3:48 PM | 28 Hours, 27 Minutes | ISO New England | NPCC | Connecticut: Massachusetts: Rhode Island: | Loss of electric service to more than 50,000 customers for 1 hour or more-Weather | Unknown | 115057 |
| 2016 | 2 | 02/13/2016 12:44 PM | 02/13/2016 4:27 PM | 3 Hours, 43 Minutes | Pacific Gas & Electric Co | SERC | California | Electrical System Separation (Islanding) where part or parts of a power grid remain(s) operational in an otherwise blacked out area or within the partial failure of an integrated electrical system-islanding | 7 | 4300 |
| 2016 | 2 | 02/16/2016 8:35 AM | 02/16/2016 5:28 PM | 8 Hours, 53 Minutes | American Electric Power - (RFC Reliability Region) (8400 Smiths Mill Road, New Albany Ohio 43054) | RFC | Virginia: Roanoke County, Montgomery County; West Virginia: Kanawha County, Cabell County; Tennessee: Sullivan County: | Loss of electric service to more than 50,000 customers for 1 hour or more-Weather | Unknown | 52640 |
| 2016 | 2 | 02/19/2016 10:00 PM | 02/20/2016 11:13 PM | 25 Hours, 13 Minutes | Detroit Edison Co | RFC | Michigan | Loss of electric service to more than 50,000 customers for 1 hour or more-Weather | Unknown | 145314 |
| 2016 | 2 | 02/24/2016 2:45 PM | 02/25/2016 5:00 AM | 14 Hours, 15 Minutes | Duke Energy Carolinas | SERC | North Carolina: South Carolina | Loss of electric service to more than 50,000 customers for 1 hour or more-Weather | 400 | 284610 |

Acquisition and Organization of Information for Analytics

Weather Underground

Hourly forecasts accessed via API

Data fields: location, observation time, weather description, temperature, humidity, wind (dir, mph, pressure), precipitation)

Data size: Depends on number of cities

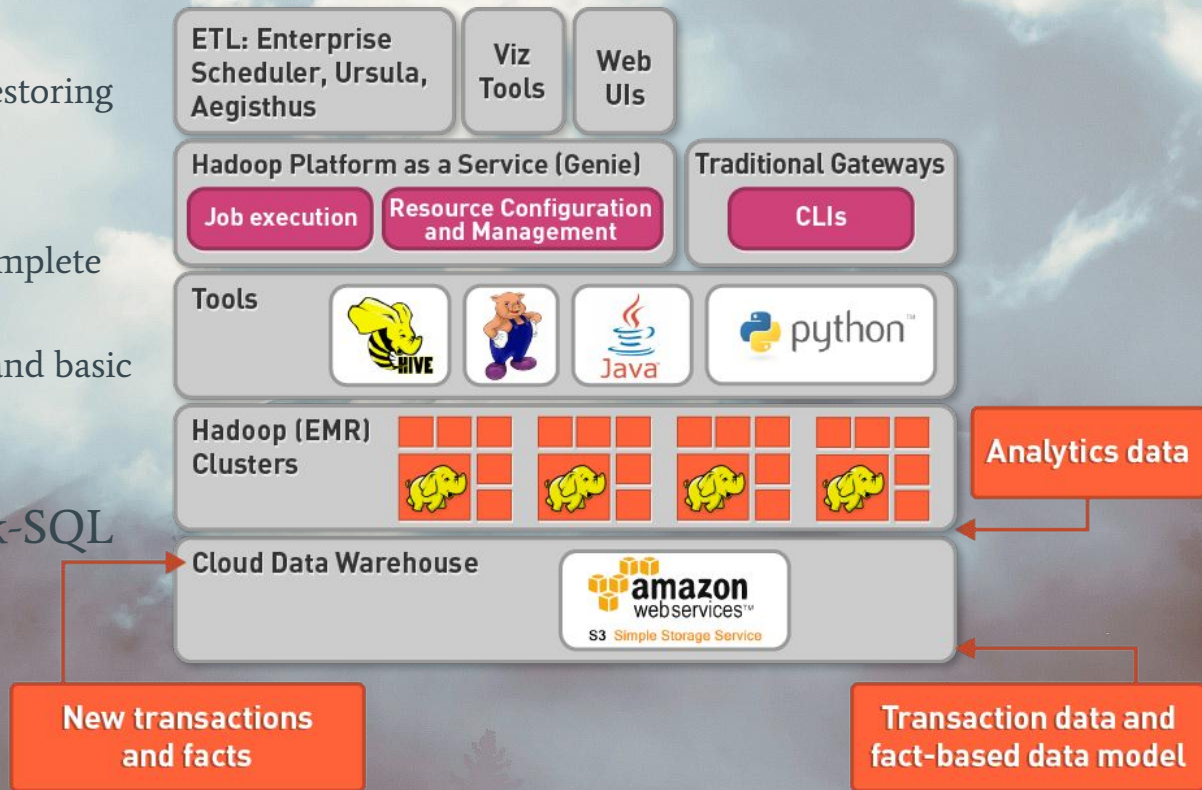
<http://www.wunderground.com/weather/api/>

```
{
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      "pop": "10"
    }
  ]
}
```


Future Architecture

- Architecture: Netflix
 - One data warehouse for restoring predictive analyses
- Storage: HDFS
 - Small, structured, semi-complete data
 - Low latency, low quality, and basic query requirements
 - Scales out
- Processing: Apache Spark-SQL
 - MLlib, RDDs, pySpark

Netflix Architecture



Pilot Architecture



Storage Layer



Processing Layer



Streaming Layer

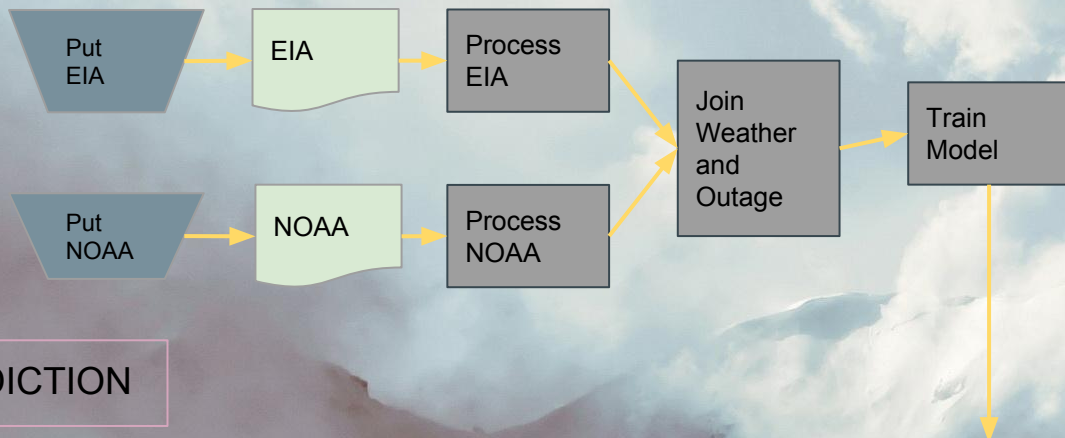


Serving Layer



Process: Pilot

LEARNING

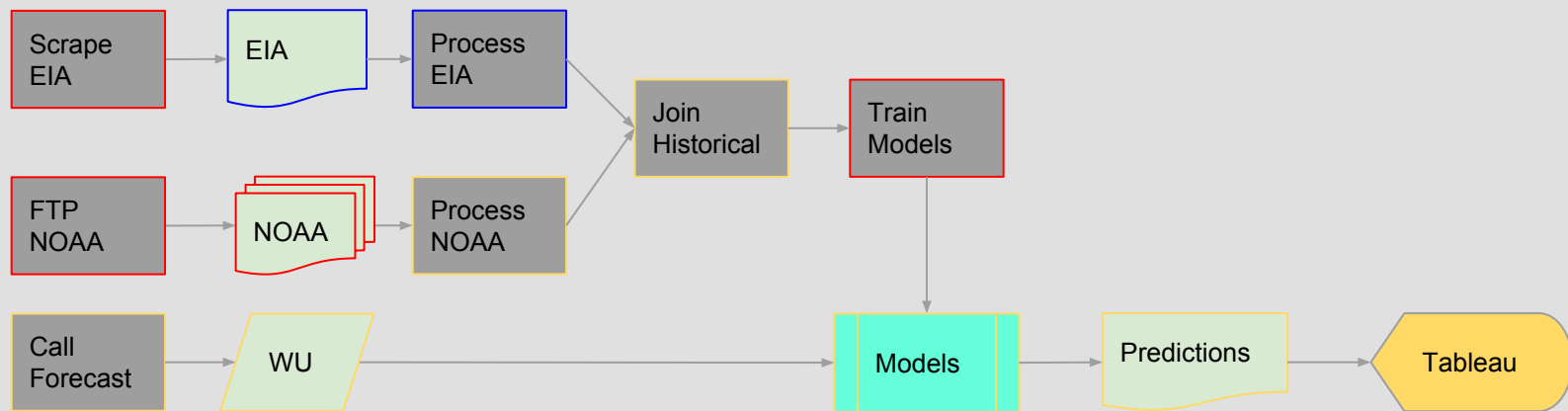


PREDICTION

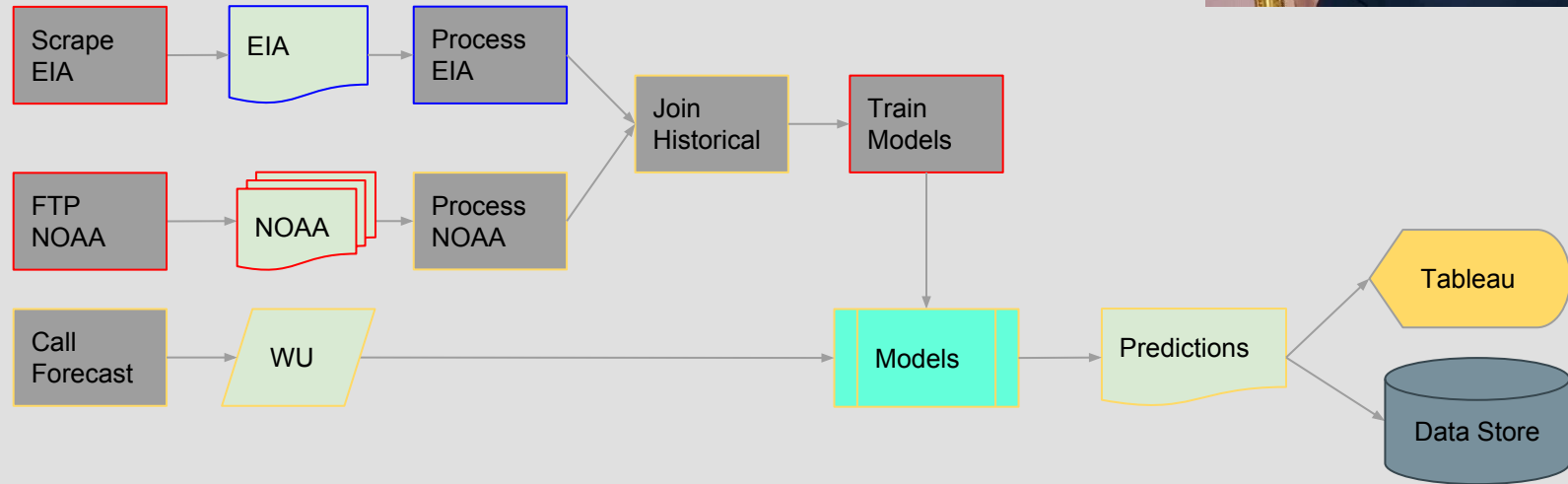


Demo

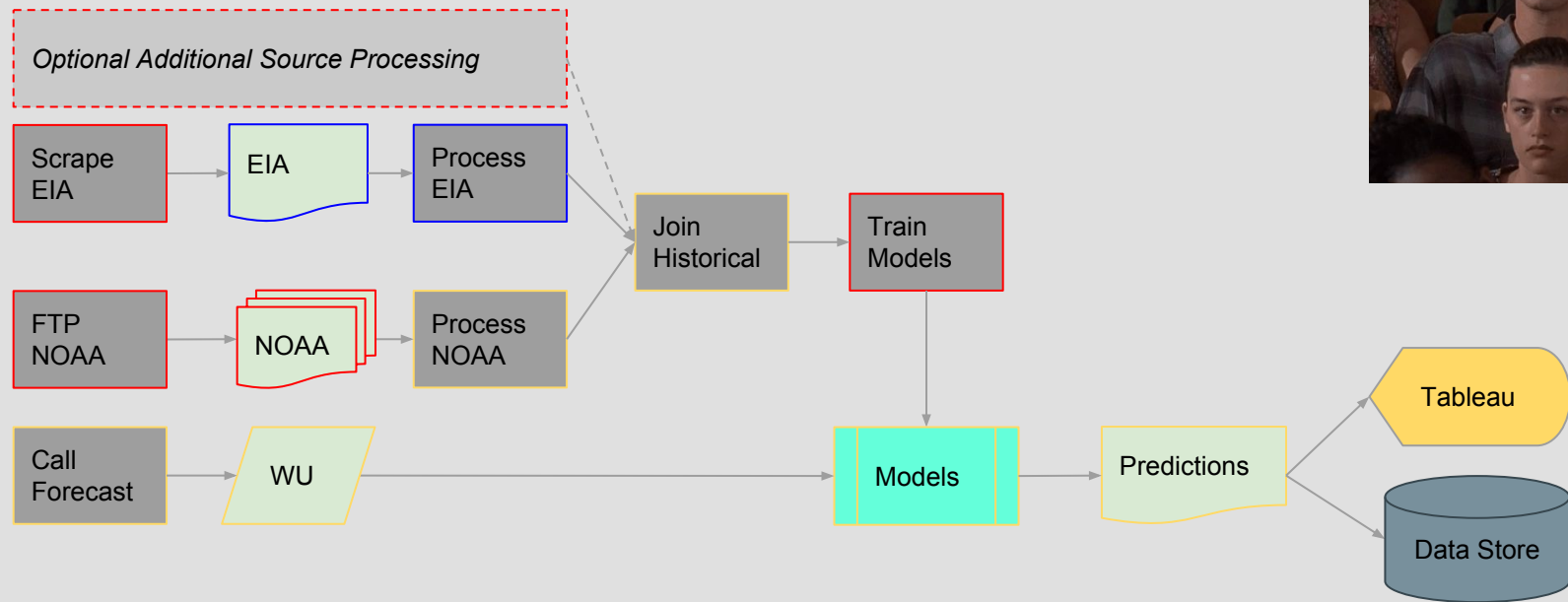
Process: Next Iteration



Process: Next Iteration



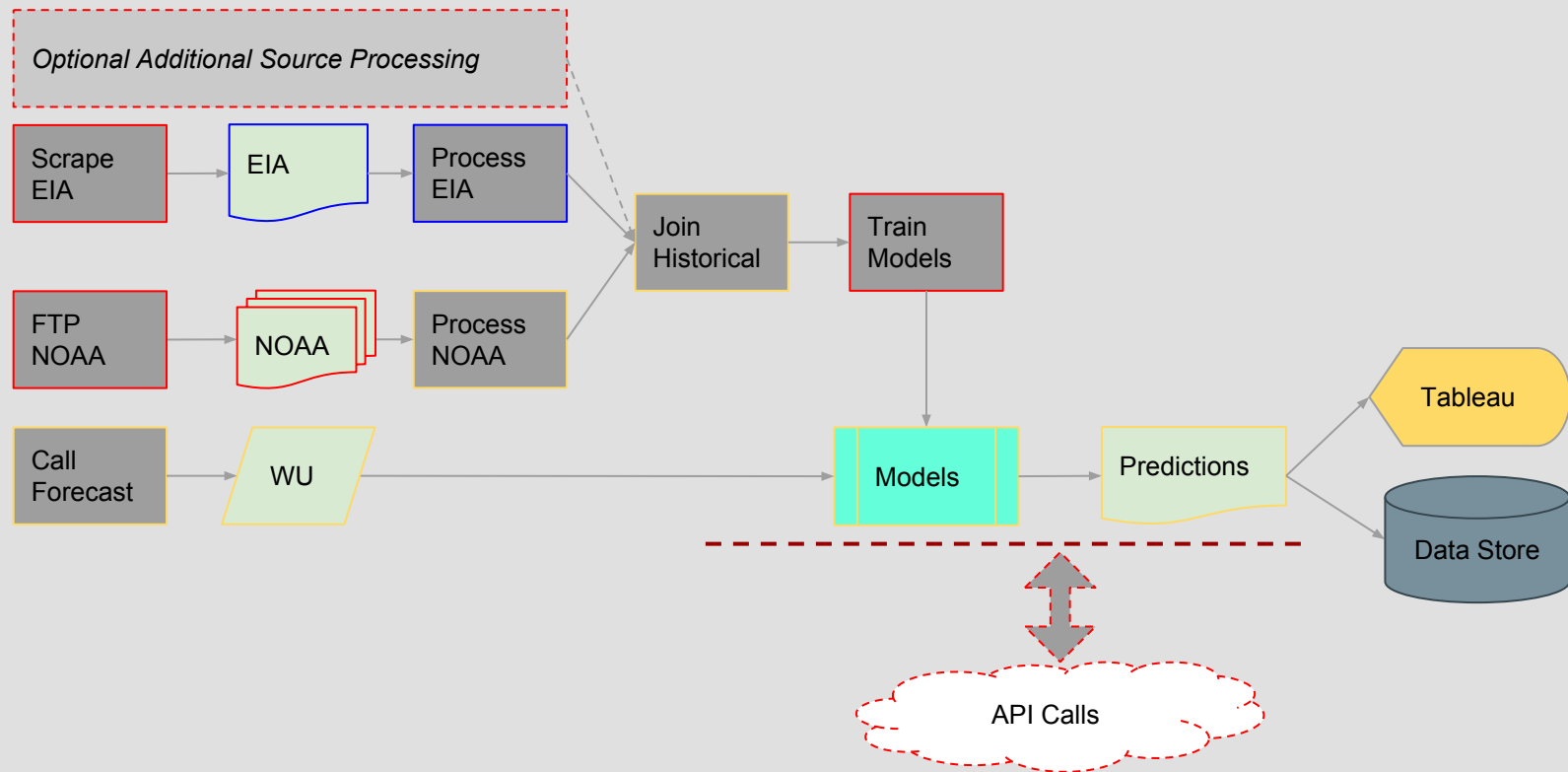
Process: Next Iteration



Live Class Reaction



Process: Next Iteration



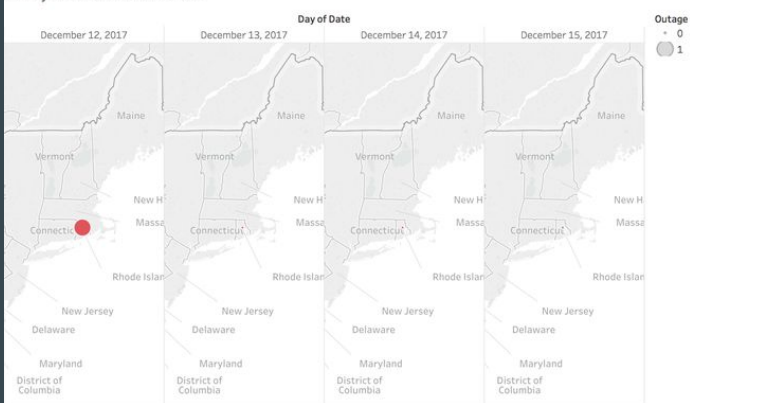
Usage

Local Residents (Public Service)

Direct or through Media Channels

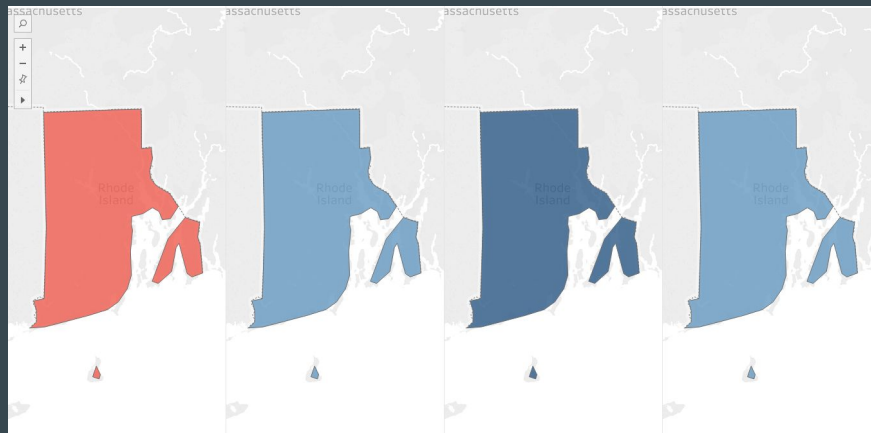
Warning / Preparation

3 Day Forecast State View



3 Day Forecast Table View

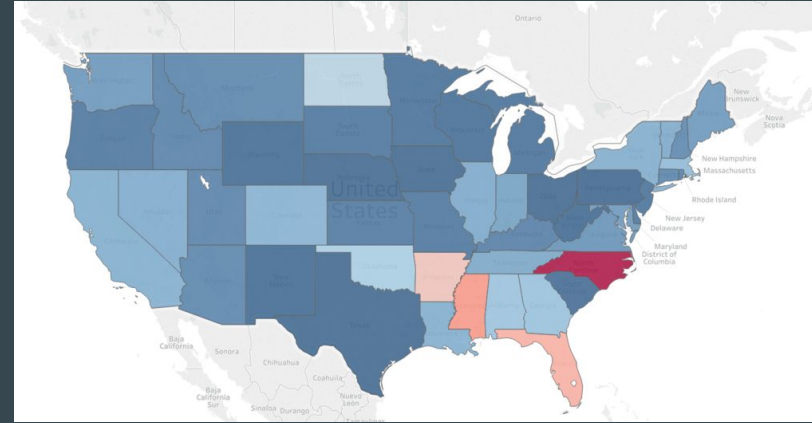
| State | Day of Date | | | | Outage |
|--------------|-------------------|-------------------|-------------------|-------------------|--------|
| | December 12, 2017 | December 13, 2017 | December 14, 2017 | December 15, 2017 | |
| Rhode Island | 1 | 0 | 0 | 0 | |



Usage

Utilities

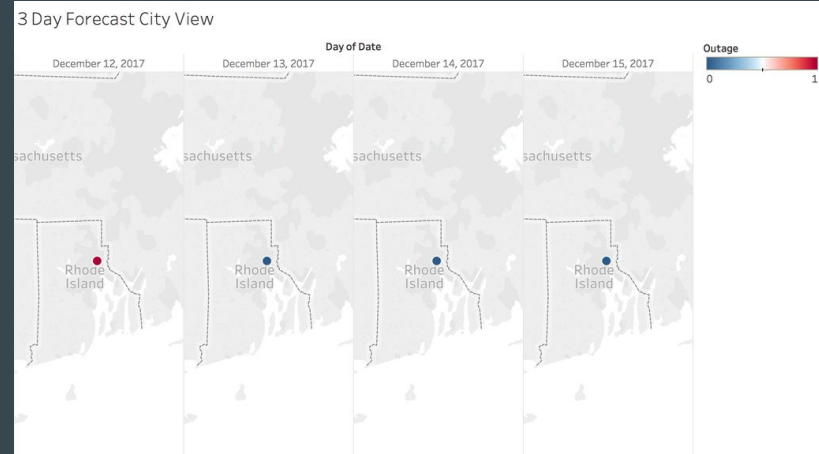
Complement Existing Process
(labor, capital costs)



Private Sector

Demand Forecasting

Supply Chain Risk



Recap

- Pilot designed for ease of scaling
 - Prepared for Volume and Velocity increases
 - Workstream supports various sources and can support Variety growth
- Suggested Developments in the Future
 - Increase breadth and depth of geographic areas served
 - Scale parts of solution separately, need-based
 - Add serving layers
 - Expand sources -> more predictors

Thank You!

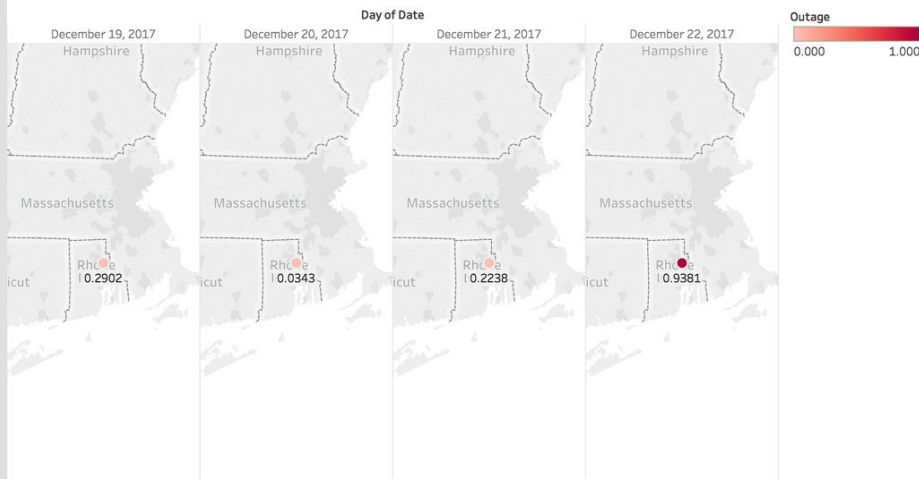
Appendix

Additional Visualizations

3 Day Forecast City View with Probability - Table



3 Day Forecast City View with Probability



Reference of Previous Work

High-speed winds during a thunderstorm may cause trees around an electric grid to crash into the distribution system feeders causing an outage in that area. This model predicts potential vulnerability to ensure the trees in the most critical areas with the highest risk to be trimmed first.

“Power outages in the distribution system are primarily caused by **wind blowing trees and limbs onto power lines**,” says Guikema. “Therefore, our model includes a number of different measures of **wind, the type of trees, and the soil moisture conditions**, which provides a measure of the stability of the soil and the likelihood of a tree being uprooted.”

A group of researchers led by Seth Guikema at the University of Michigan has created a model that predicts the number of power outages “based on **wind speed estimates, population density, soil moisture levels, drought indices, and information about trees in each census tract**.”

Solution Architecture Draft



Learning
Data



xlsx → csv → RDD



NOAA

csv → RDD

RDD

1) query for past outages due to severe weather

2) label historical weather data

3) teach power outage classifier

4) predict outages based on current forecast

5) surface predictions in viz layer



json → raw data → RDD



Prediction
Data