# **Predict Bart Station Capacity**

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### Data Sources - Historical

- 2016 historic daily rider exits per BART station
  - Origin Station
  - Destination Station
  - Number of Exits
  - Hour of the day
- 2016 historic daily weather for San Francisco (zip 94105)
  - Wind speed
  - Humudity
  - Minimum temp.
  - Max temp
  - Avg. temp
  - Sea pressure

### Data Sources - Live

- Live BART information pulled every ten minutes
  - Origin station
  - Destination station
  - Number of trains
  - Bike flag
  - Minutes until arrival
- Live weather information pulled every ten minutes
  - Wind
  - o Min. temp
  - o Max. temp
  - O Avg. temp
  - Humidity
  - Sea pressure

### Architecture

#### EC2 (airflow)

Bart arrival information
Weather information
minute update

#### **Static Data**

1) Bart 2016daily riders2) Daily

weather 2016

#### S3 'data lake'

- 1) Raw data
- 2) Normalized data
- 3) Predicted ridership

S3 webpage to host predictions

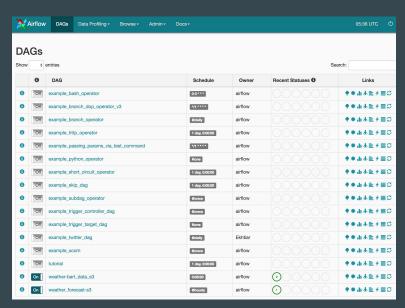
#### **EMR** cluster

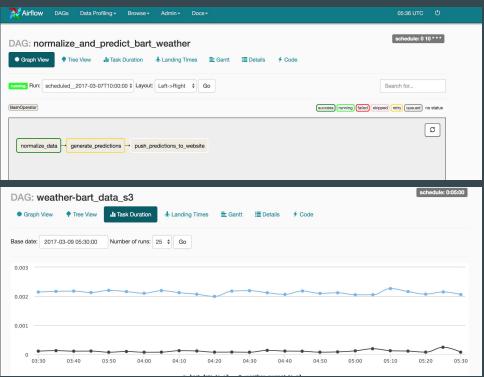
Data
normalization
GBoost
model

#### MongoDB on EC2

1) Normalized data

### **Airflow**





# Next Steps & Challenges

### • <u>Challenges</u>

- BART data in XML format. Needed to convert to ISON
- Sparse documentation for Airflow
- O Differences between spark interactive and spark-submit
- Unix time

### • Next Steps

- Change daily prediction to hourly prediction
- Incorporate direction of BART train (North or South)
- Incorporate localized weather data for each BART station
- Add visualizations
- Incorporate MUNI data
- Lambda architecture with 'speed' layer

## Results

http://bart-capacity-predictions.com.s3-website-us-east-l.amazonaws.com/