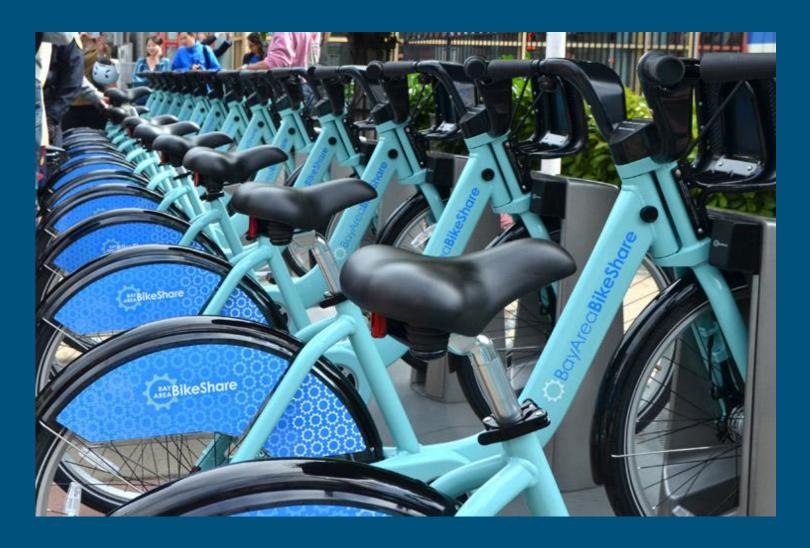
# BikeShare Demand Forecasting



Tom Yedwab & Adam Spitzig & Chris Murray

## Overview of the problem

- Bike sharing is a popular solution for commuters in cities
- Check out a bike from one station and return to another
- However, problems can occur if:
  - Bikes are not available at a station (station is empty)
  - Docks are not available for bike return (station is full)
- Bike demand potentially correlated with weather
- ⇒ Predict bike sharing demand based on weather forecast

## Acquisition and organization of data

Bike sharing data (CSV format):

http://www.bayareabikeshare.com/open-data

2 years  $\rightarrow$  2.6 GB (trip data only 77 MB)

Weather data (JSON format):

https://developer.forecast.io/docs/v2

2 years  $\rightarrow$  30 MB

# Example bike sharing data

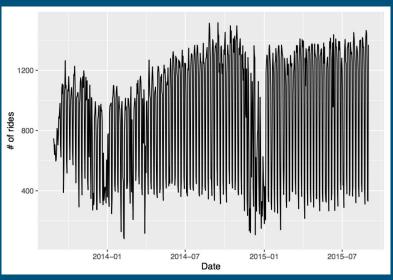
|         |          |               | Start    |               | End      |        | Subscription |          |
|---------|----------|---------------|----------|---------------|----------|--------|--------------|----------|
| Trip ID | Duration | Start Date    | Terminal | End Date      | Terminal | Bike # | Туре         | Zip Code |
|         |          | 8/29/2013 14: |          | 8/29/2013 14: |          |        |              |          |
| 4576    | 63       | 13:00         | 66       | 14:00         | 66       | 520    | Subscriber   | 94127    |
|         |          | 8/29/2013 14: |          | 8/29/2013 14: |          |        |              |          |
| 4607    | 70       | 42:00         | 10       | 43:00         | 10       | 661    | Subscriber   | 95138    |
|         |          | 8/29/2013 10: |          | 8/29/2013 10: |          |        |              |          |
| 4130    | 71       | 16:00         | 27       | 17:00         | 27       | 48     | Subscriber   | 97214    |
|         |          | 8/29/2013 11: |          | 8/29/2013 11: |          |        |              |          |
| 4251    | 77       | 29:00         | 10       | 30:00         | 10       | 26     | Subscriber   | 95060    |
|         |          | 8/29/2013 12: |          | 8/29/2013 12: |          |        |              |          |
| 4299    | 83       | 02:00         | 66       | 04:00         | 67       | 319    | Subscriber   | 94103    |

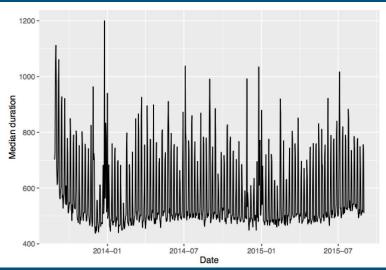
| Station |                                     |           |             | Dock  |               |              |
|---------|-------------------------------------|-----------|-------------|-------|---------------|--------------|
| ID      | Name                                | Lat       | Long        | count | Landmark      | Installation |
| 66      | South Van Ness at Market            | 37.774814 | -122.418954 | 19    | San Francisco | 8/23/2013    |
| 67      | Market at 10th                      | 37.776619 | -122.417385 | 27    | San Francisco | 8/23/2013    |
|         | Yerba Buena Center of the Arts (3rd |           |             |       |               |              |
| 68      | @ Howard)                           | 37.784878 | -122.401014 | 19    | San Francisco | 8/23/2013    |

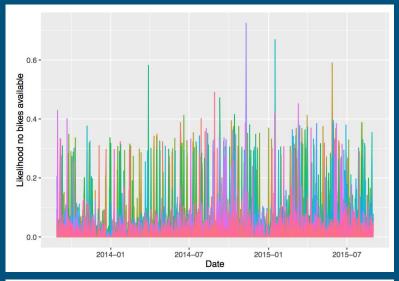
## Example weather data

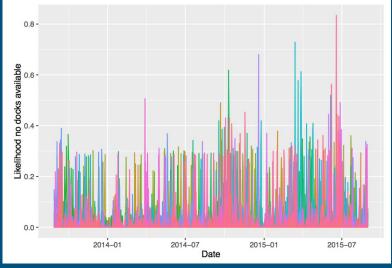
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"time": 1421049600.
"summary": "Clear throughout the day.",
"icon": "clear-day",
"sunriseTime": 1421076352,
"sunsetTime": 1421111525.
"moonPhase": 0.73,
"precipType": "rain",
"temperatureMin": 49.31,
"temperatureMinTime": 1421082000,
"temperatureMax": 58.94,
"temperatureMaxTime": 1421103600,
"apparentTemperatureMin": 48.66,
"apparentTemperatureMinTime": 1421071200,
"apparentTemperatureMax": 58.94,
"apparentTemperatureMaxTime": 1421103600,
"windSpeed": 1.12,
"windBearing": 126,
"pressure": 1<u>02</u>2.<u>4</u>7
```

## Initial bike share data analysis

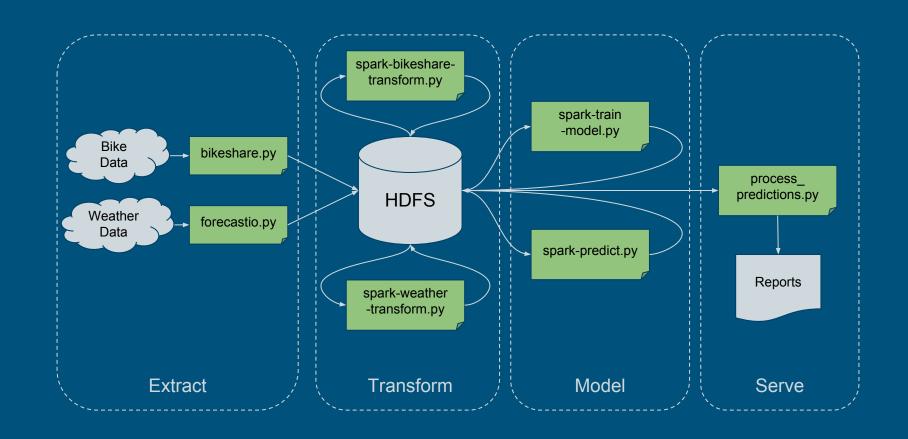




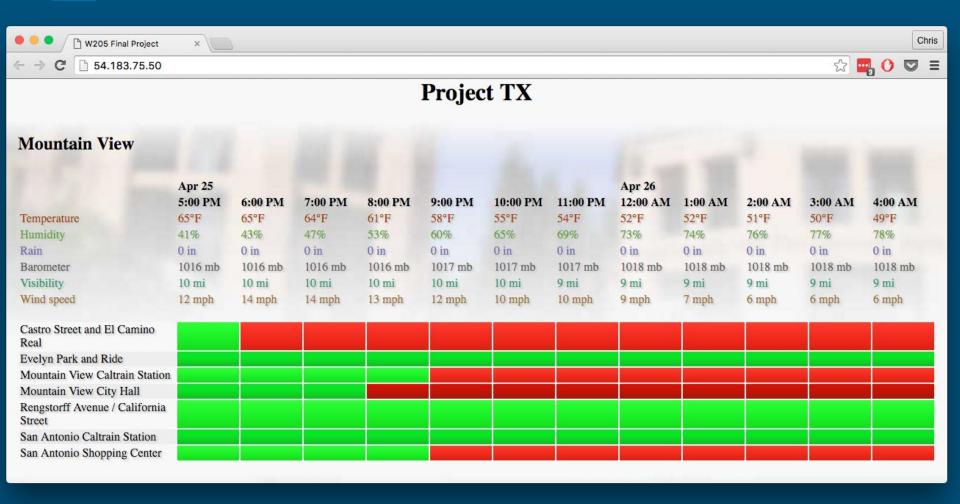




### Overall architecture of the solution



### Visualization of results



### Results

- Settled on data sources
  - Pivoted away from problematic public transportation data
- Data extraction code finished
- Data analysis complete
- Basic prediction modeling using Spark complete
  - Support Vector Machine regression model used
  - Predict when no bikes are available

## Roadmap for improving the solution

#### How to scale the solution

- Station status data is huge (2+ GB)
  - Pre-filtering can dramatically reduce size
- Aggregate trip data and station data over time to reduce size

#### How to evolve the project

- Use streaming weather data to provide real-time demand forecasting
- Include more cities

#### Additional data sources

- Stadium event data
  - Do people ride more bikes when there is a major public event?
- Road construction / Traffic data
- Data from other public transit systems
  - Is availability in different systems correlated?