

SF BikeShare



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Dataset

SF Bikeshare Data : 2GB



Station



Status

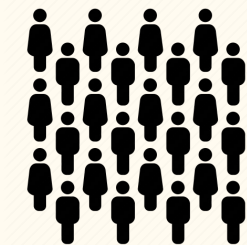


Trips



Weather

SF Population Data



Population by
Neighborhood in San
Francisco

Analytic goals



Analytic goals



Predict number of bikes available at a given station with

- station information
- weather condition
- type of day
- hour
- population

Related Works

- Predicting number of daily trips
 - Predictors:
 - weather condition
 - number of bikes available
 - type of day (business day vs holiday vs weekend)
- Most important features:
 - business_day, temperature, month

Preprocessing Algorithms

Total runtime: **31 mins**

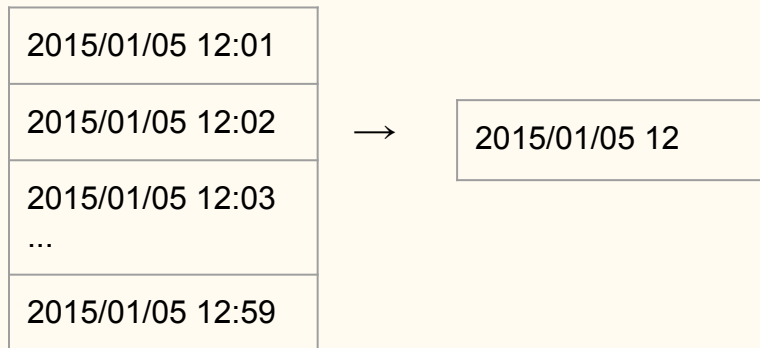
Weather:

- *precipitation* \rightarrow *is_rain*
- *mean_temperature*

Status:

- Convert *date* to DateTime() \rightarrow *is_weekend*

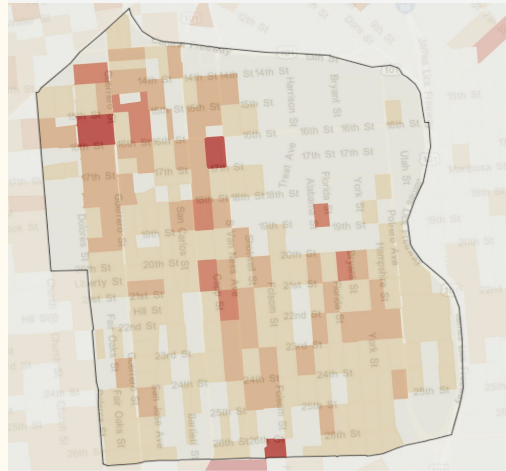
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Preprocessing Algorithms cont...

Adding **population** field:

- Station table has lat, long columns
- Used Socrata SF Data API & GeoNames API to map the lat, long columns to SF neighbourhoods
- Joined neighbourhood population data to the the station table



Machine Learning Outcomes

Spark
ML

Linear Regression: 3.4037

Decision Tree: 2.92

Gradient-Boosting: 2.93

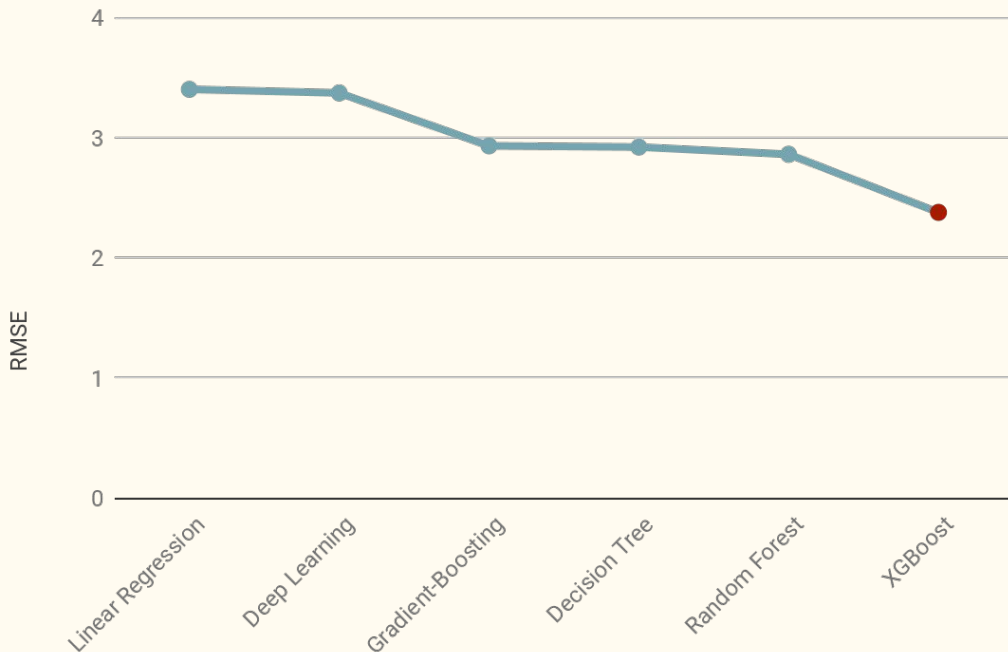
Random Forest: 2.86

H2O

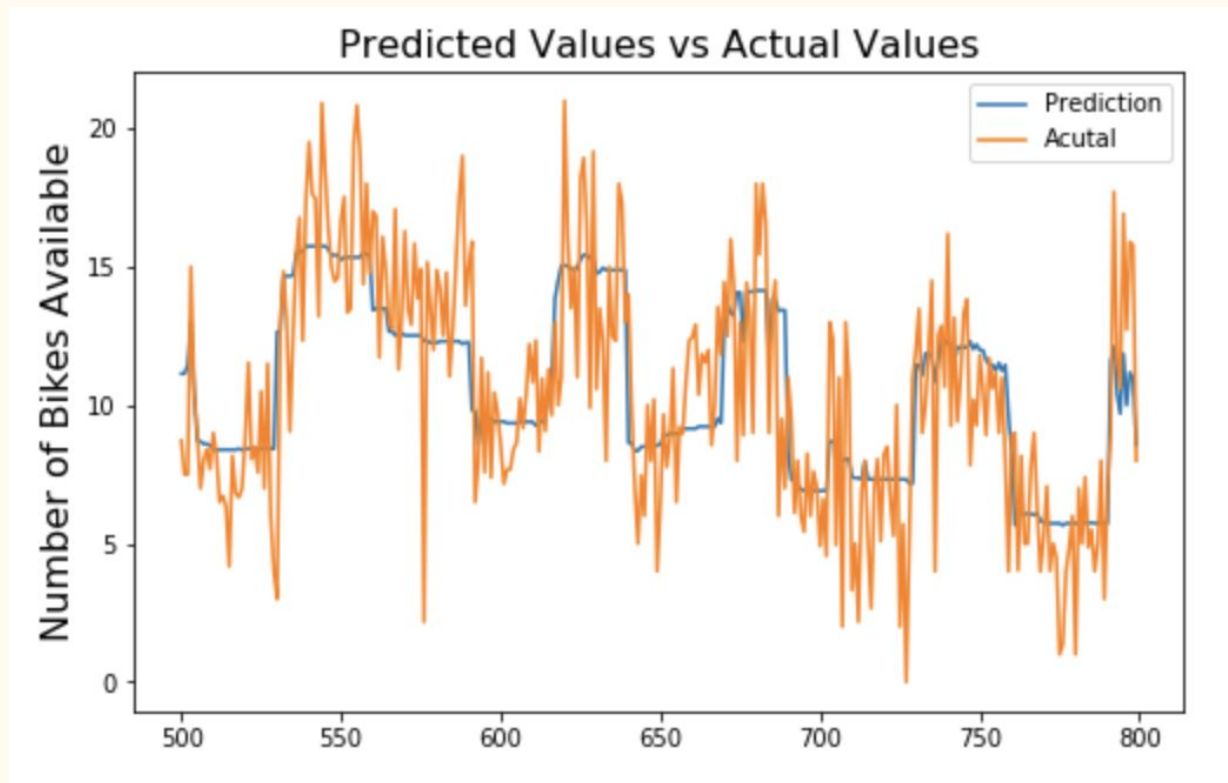
Deep Learning: 3.37

AutoML - XGBoost: 2.71

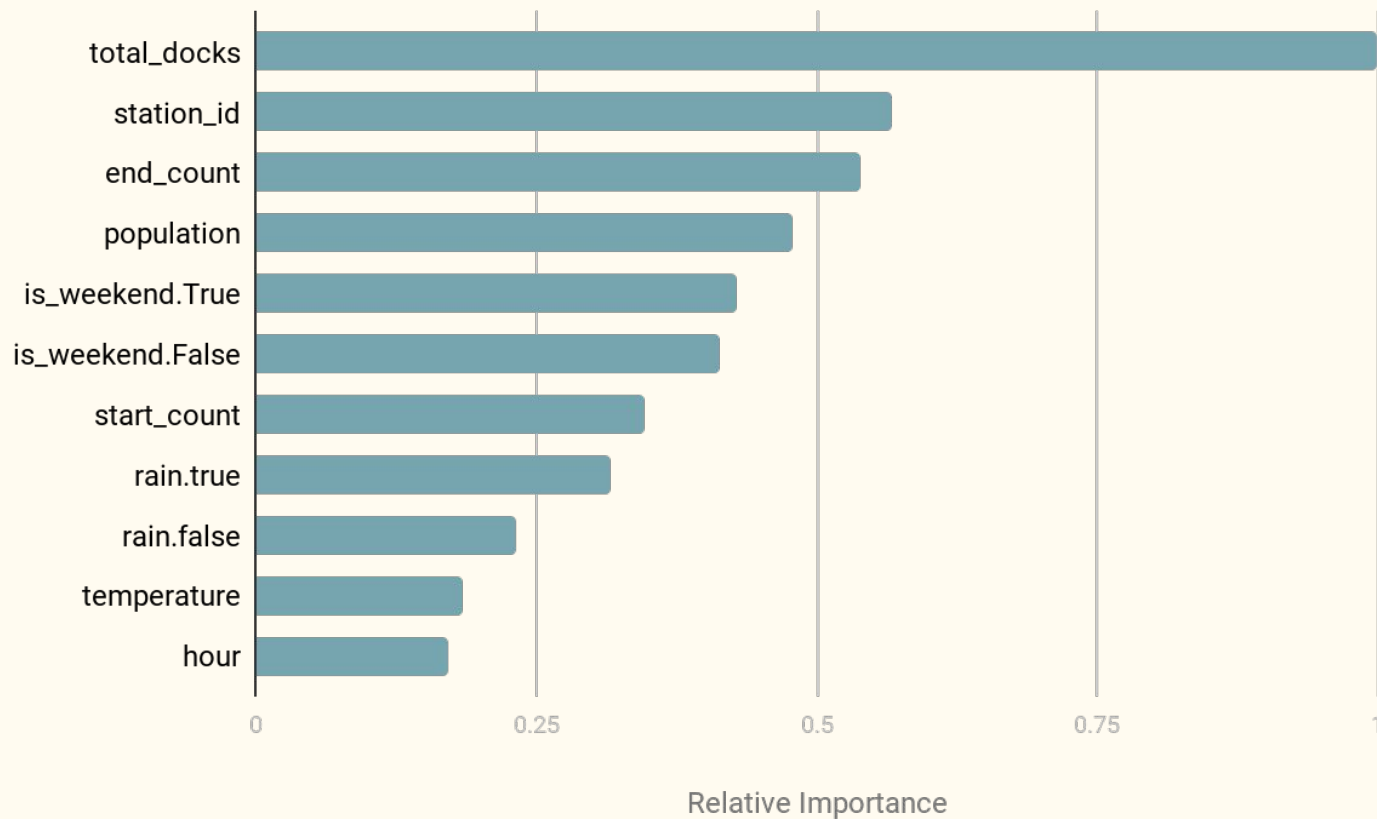
XGBoost Regressor has the lowest RMSE



Random Forest Predictions



Feature Importance



H2O

Deep Learning:

Train RMSE

Test RMSE

3.48

3.37

Auto ML - XGBoost:

Train RMSE

Test RMSE

2.22

2.37

Runtime Comparison

	r5a.8xlarge (memory optimized)	r5a.12xlarge (memory optimized)	c4.8xlarge (compute optimized)	c4.8xlarge (compute optimized)
Nodes	5 nodes	3 nodes	4 nodes	5 nodes
Time	14m 12s	13m 2s	12m 15s	11m 43s

Conclusion & Lessons Learned

- For spark ML: Random Forest performed the best on our data
- For H2O: XGboosting performed the best
- H2O was slower on EMR clusters when operating on data
- Population of the station area is correlated to the number of bikes available