



Analysis of Uber & Lyft Prices

Group 13

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- Context
- Data Overview
- Data Visualization
- Data Prediction



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- **Uber and Lyft are the two dominant business rideshare apps in the states.**
- **Prices for those two rideshare apps depends on numerous factors such as weather, time, destination and etc.**

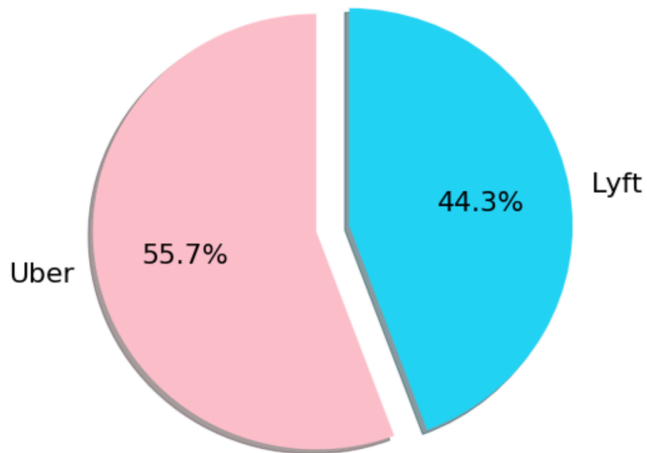
Data Overview



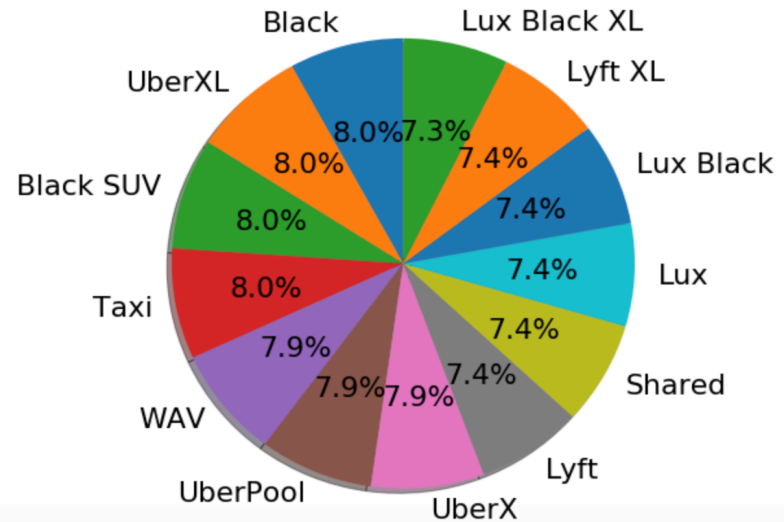
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Dataset of cab rides collected for a week in Nov - Dec 2018 in Boston.
Collected at a regular interval of 5 mins.

Cab type distribution



Cab name distribution

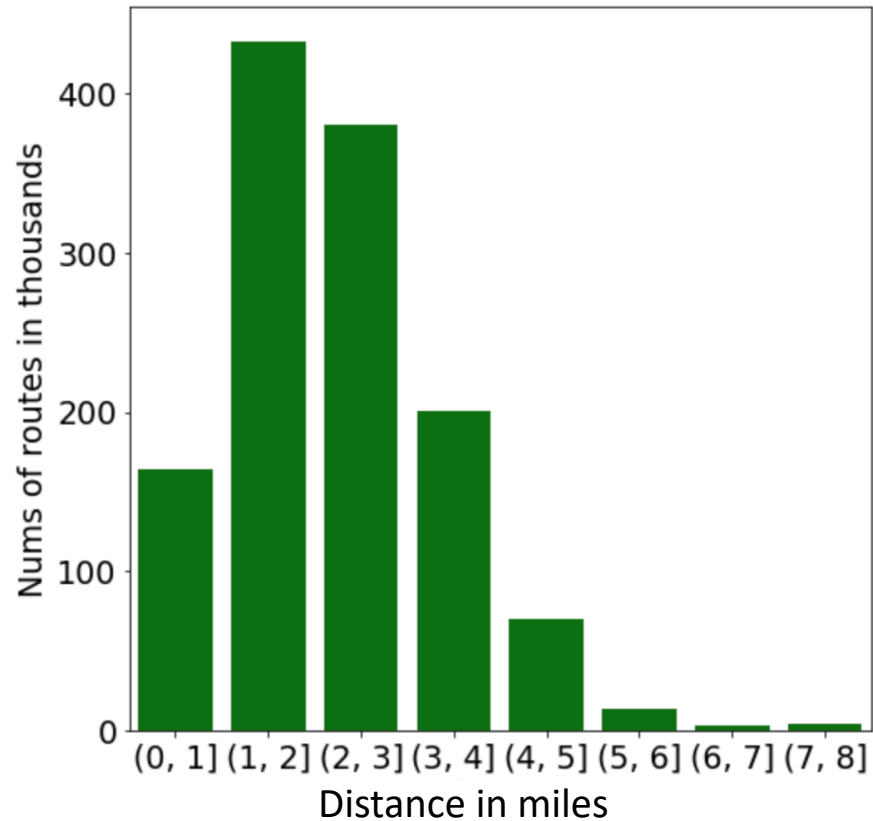


Data Overview

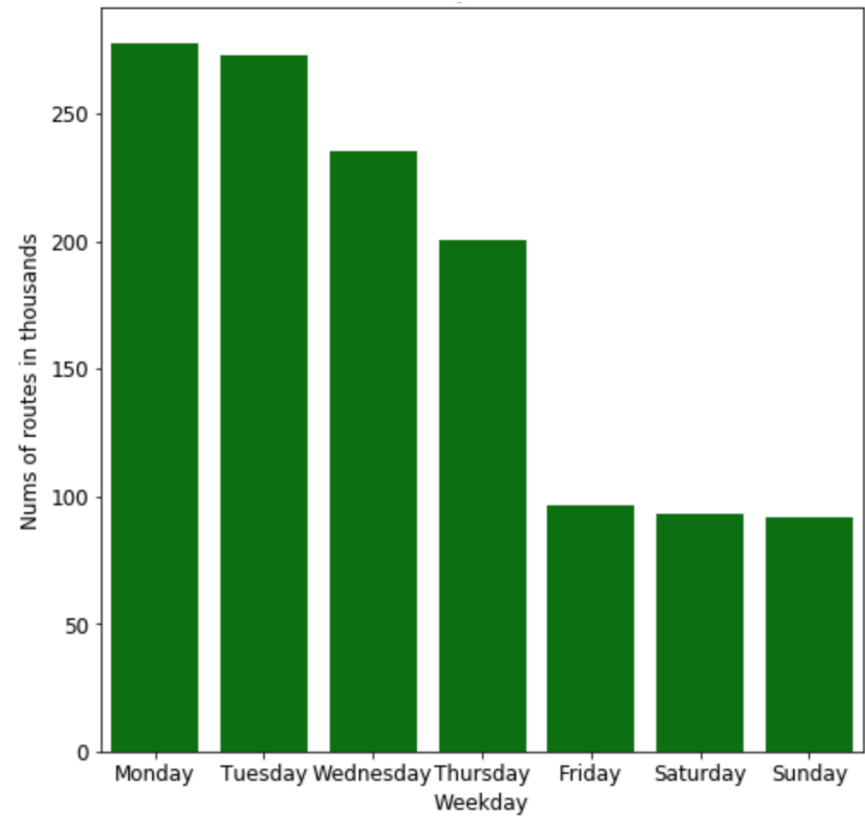


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Distance Distribution



Weekday Distribution

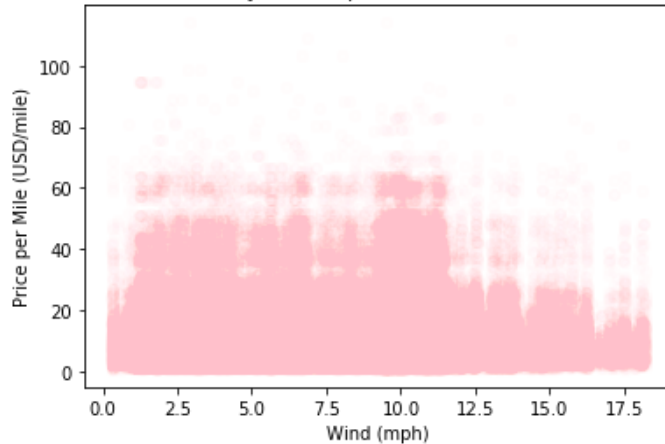




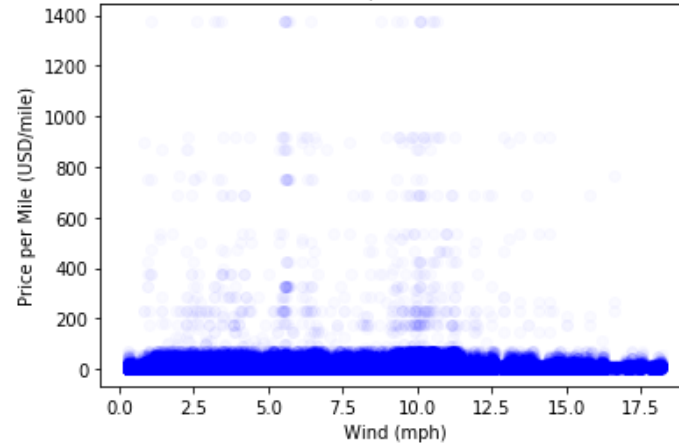
Uber vs Lyft: Effect of Weather on Price per Mile

Weather does not affect the prices of rides

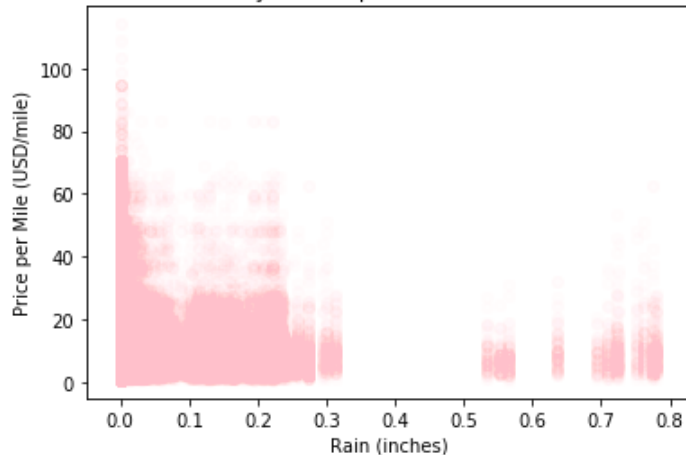
Lyft: Price per Mile vs Wind



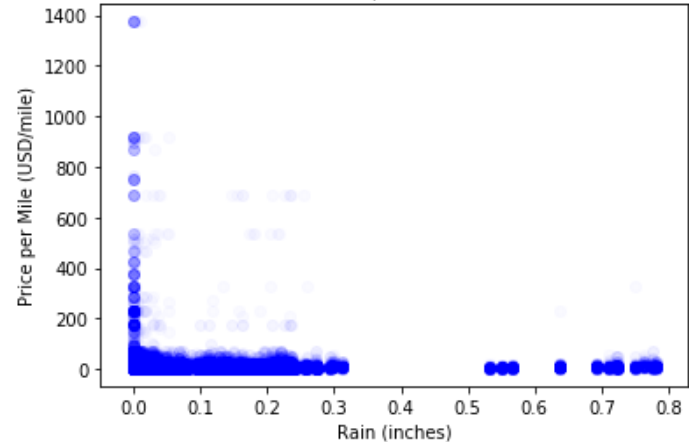
Uber: Price per Mile vs Wind



Lyft: Price per Mile vs Rain



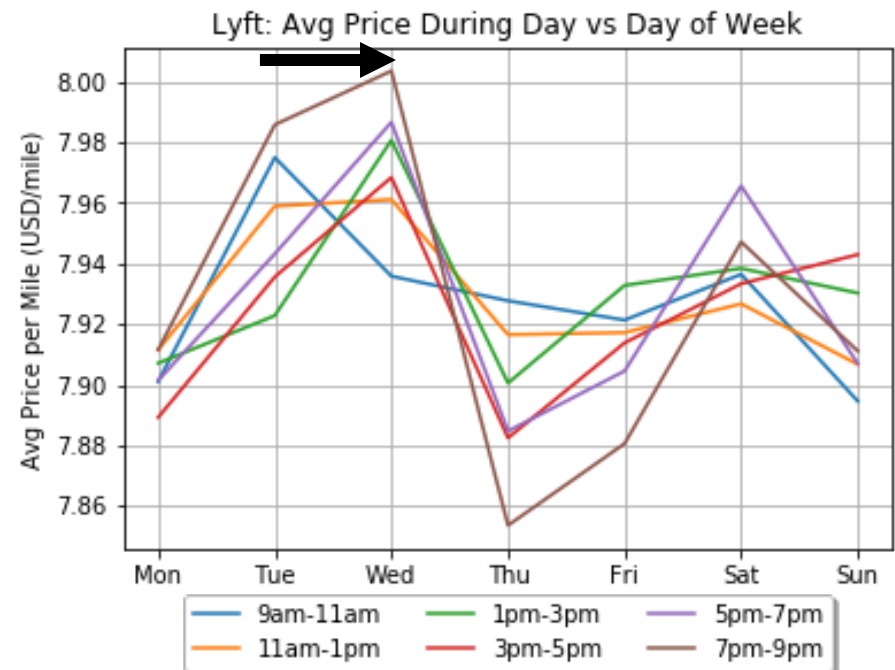
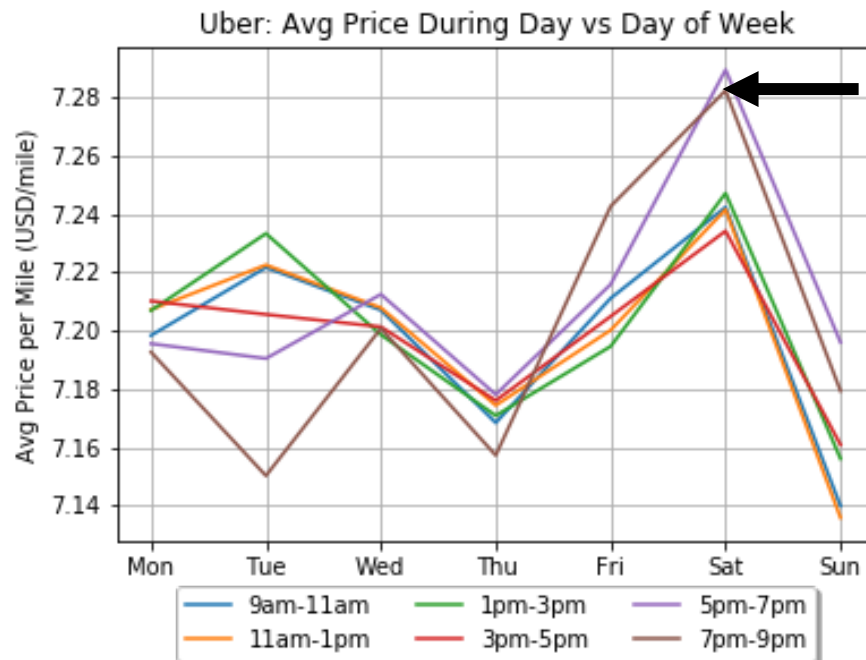
Uber: Price per Mile vs Rain





Uber vs Lyft: Price per Mile vs Day of Week

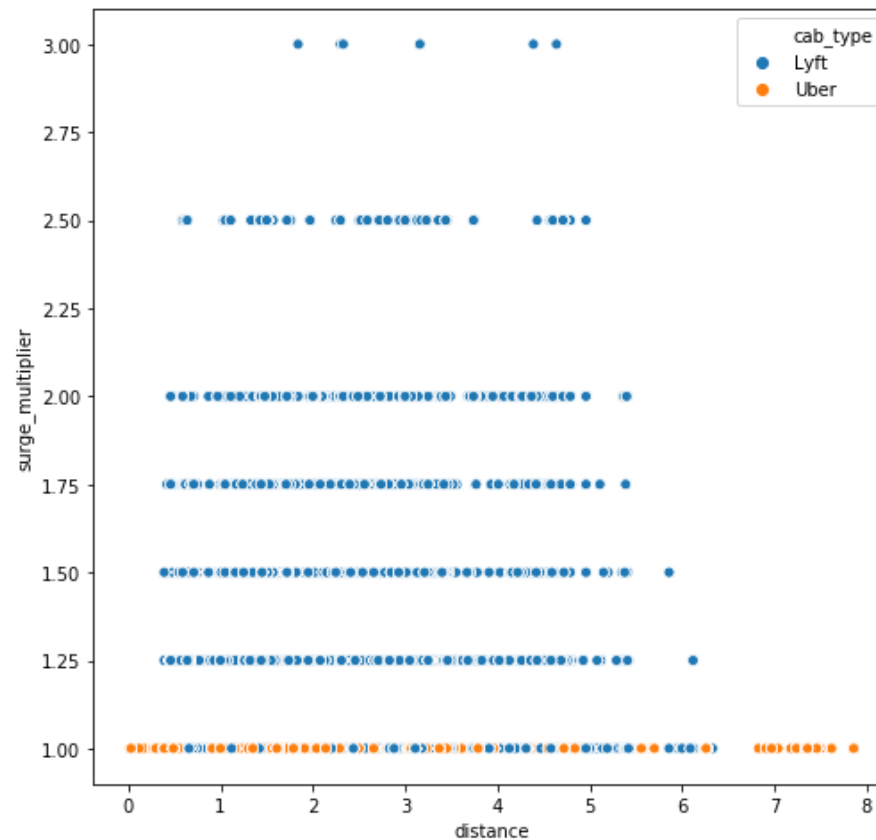
- Uber is generally, slightly less expensive than Lyft
- Uber's prices fluctuate less than Lyft's
- Prices spike from 5pm-9pm





Uber vs Lyft: Surge Multiplier vs Distance

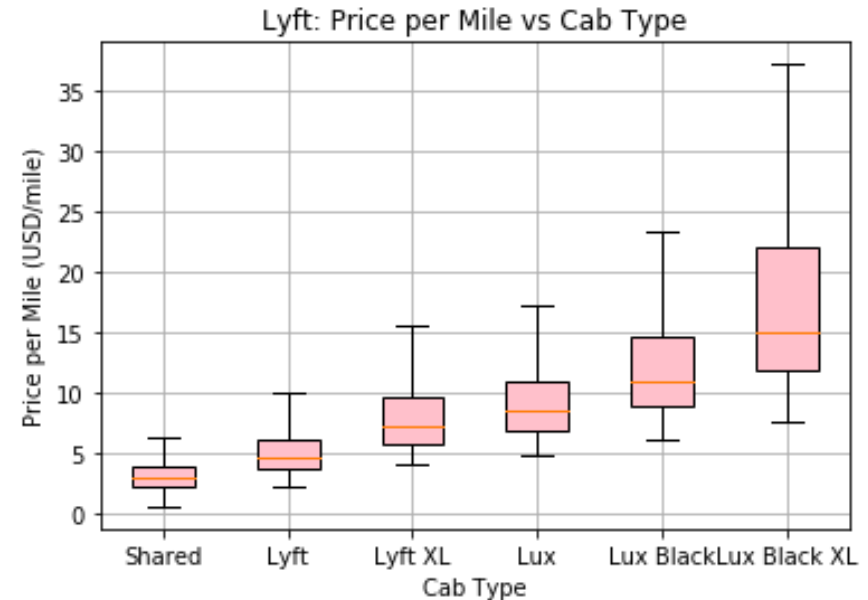
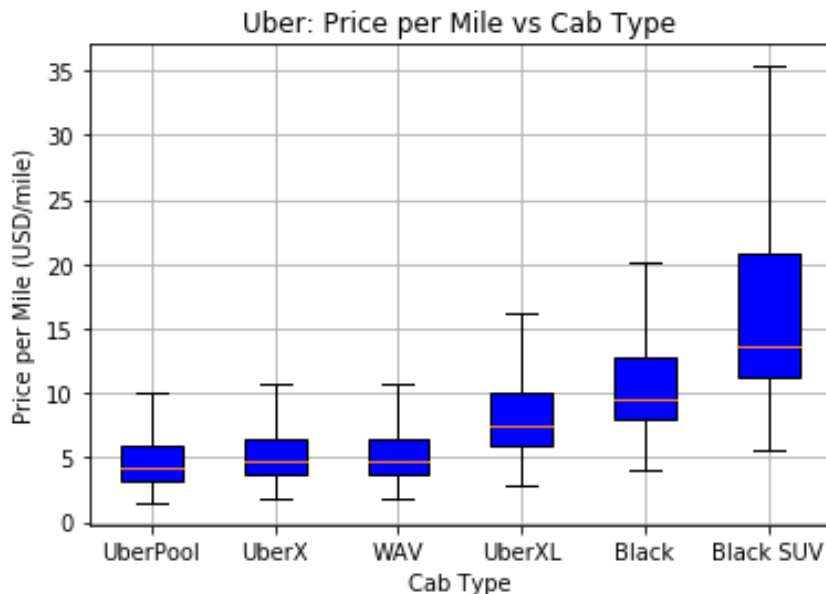
Uber Surge Multiplier is always 1





Uber vs Lyft: Price per Mile vs Cab Type

- Uber is generally cheaper than Lyft
- Lyft has much more price variation
- Better cab type leads higher price

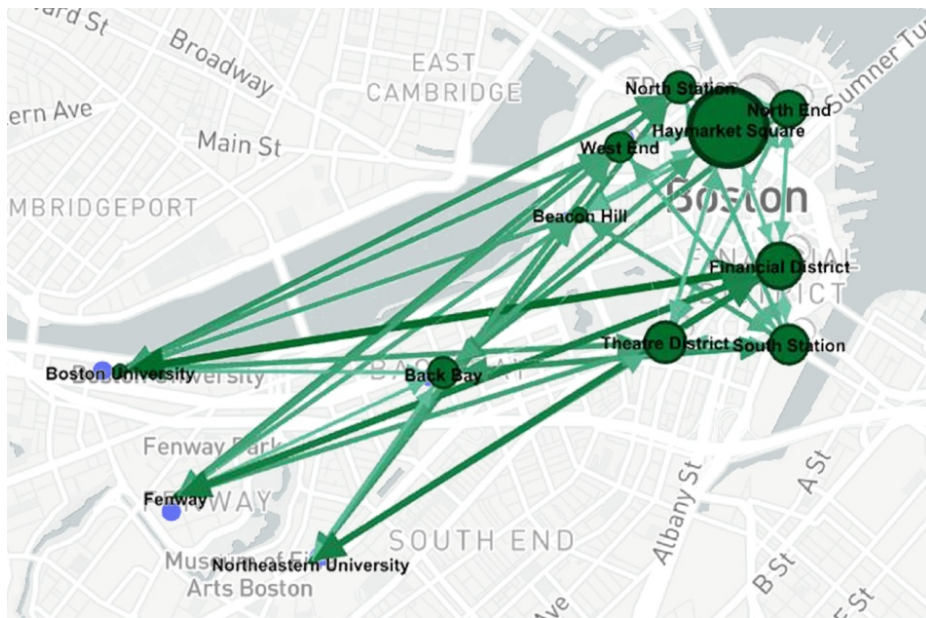




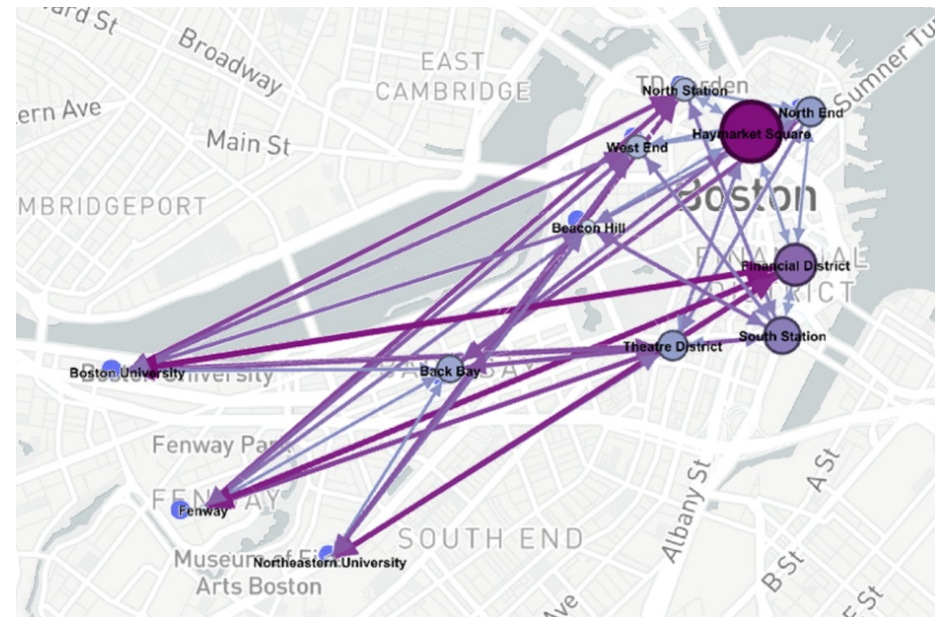
Price Networks for Uber & Lyft

- Edge weight = mean price
- Node weight = price/distance
- Suburb area is cheaper
- Downtown area node is quite large
- Generally: Lyft < Uber

Uber Price Network



Lyft Price Network





- Goal
 - Predict future Uber & Lyft rides price
- Training set
 - Splitting original dataset into 80% training set and 20% validation set.
- Test set
 - Manually collected 3000 entries from Uber & Lyft App.
- Model
 - Random Forest
 - SVM
 - Decision Tree

Data Prediction

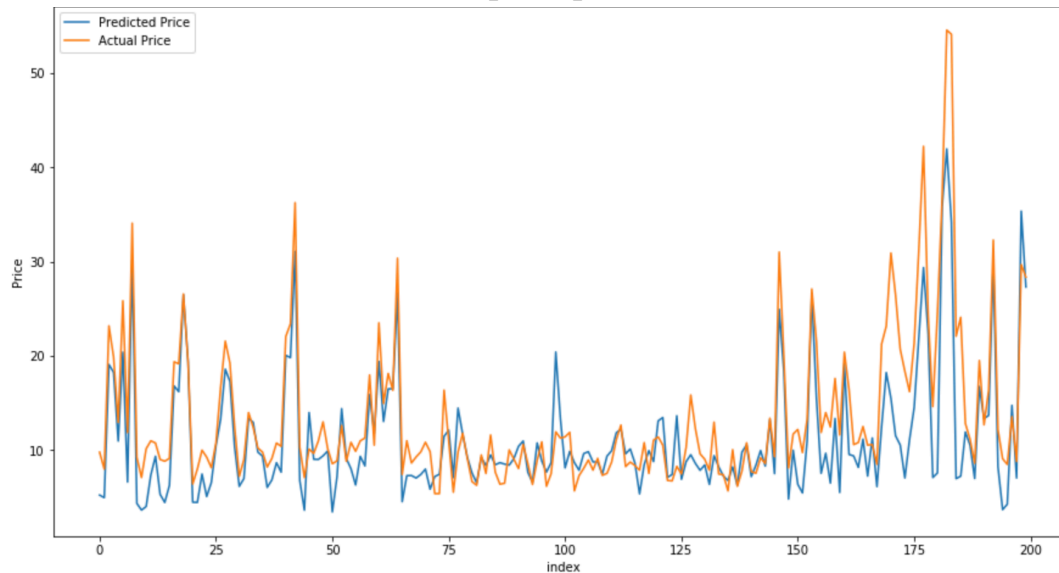


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Accuracy	Training set	Validation set	Test set
Random Forest	95.95 %	89.4 %	81.99 %
Decision Tree	98.67 %	93.01 %	79.58 %
SVM	84.56 %	59.77 %	61.56 %

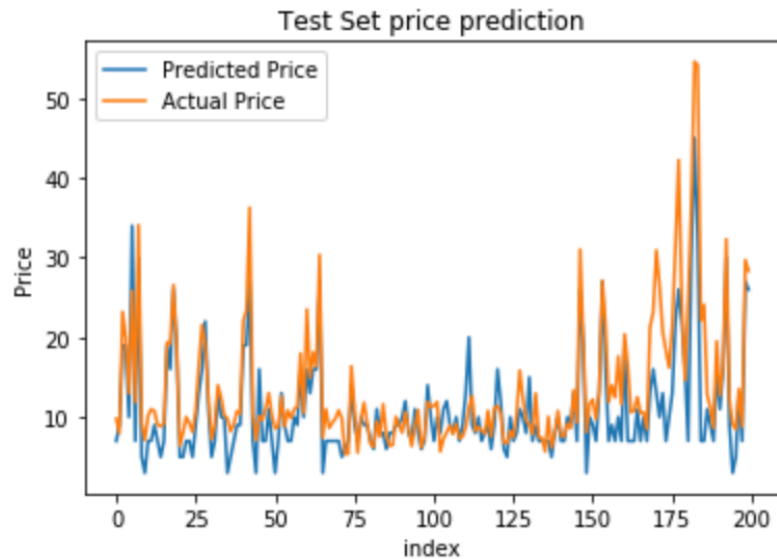
Random Forest:

Test set price prediction

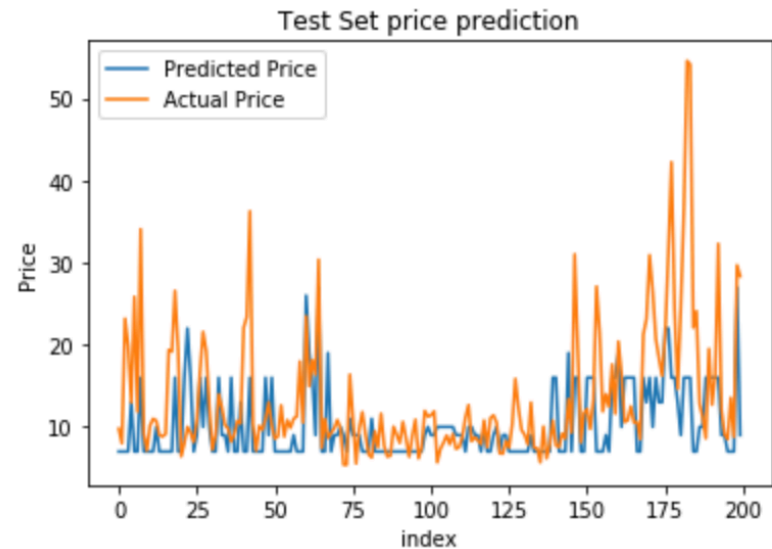




Decision Tree:



SVM:



Conclusion:

Random Forest achieves the best results.



Thank You