Realtime & Big Data Analytics Final Project Speaker: CHEN-YUAN HO





Big Data Analytics Symposium - Summer 2017

Analytics Project:

Differences between Taxi and Uber

Team:

Chen-Yuan Ho

Abstract:

- ✓ Goal: know differences between Uber and Taxi
- ✓ Tool: NYU HPC Dumbo, Hive
- ✓ Analytics: weather/month/hour vs. rides

Motivation

Who are the users of this analytic?

- ✓ Taxi
- ✓ Uber

Who will benefit from this analytic?

- ✓ Taxi
- ✓ Uber
- ✓ Startup companies who are interested in sharing economy

Why is this analytic important?

- ✓ Uber has posing great impact on shared transportation.
- ✓ Locate the competitive edges of both Uber and taxi, by identifying intrinsic factors of them.

Goodness

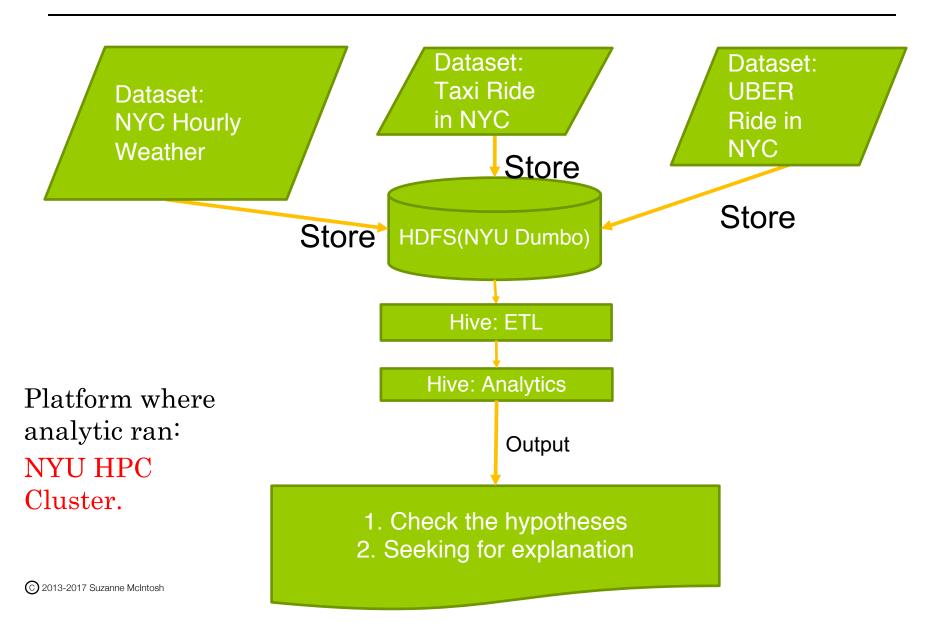
What steps were taken to assess the 'goodness' of the analytic?

- 1. Understood the shared transportation industry before collecting the datasets.
- 2. Source datasets are from official website.
- 3. Check the analytic results with relative researches.

Hypotheses

- Both occurred the most in rainy days.
- Both occurred the most in hot and cold temperature.
- Both occurred the most in rush hours and late night.
- ◆No significant differences between the distribution of both Uber and taxi rides on monthly basis.

Design Diagram



Data Source

Data Sources I

Name:

NYC TLC Trip Data 2014 – Yellow Taxi

Description:

Collected from NYC Open Data. Including pickup & drop-up date-time, and fare, distance, location of pickup & drop-up from April to September in 2014

Size of data:

12.9 GB

Data Source

Data Sources II

Name:

Uber Pickups in NYC 2014, from FiveThirtyEight

Description:

Collected from Fivethirtyeight.com. The dataset includes pickup date-time and location from April to September in 2014.

Size of data: 192 MB

Data Source

Data Sources III

Name:

Historical Weather Dataset in NYC of 2014

Description:

Parsed from UndergroundWeather.com. Including temperature and weather condition on hourly basis for everyday from April to September, 2014.

Size of data:

176KB

Analysis: Check the Hypothesis I

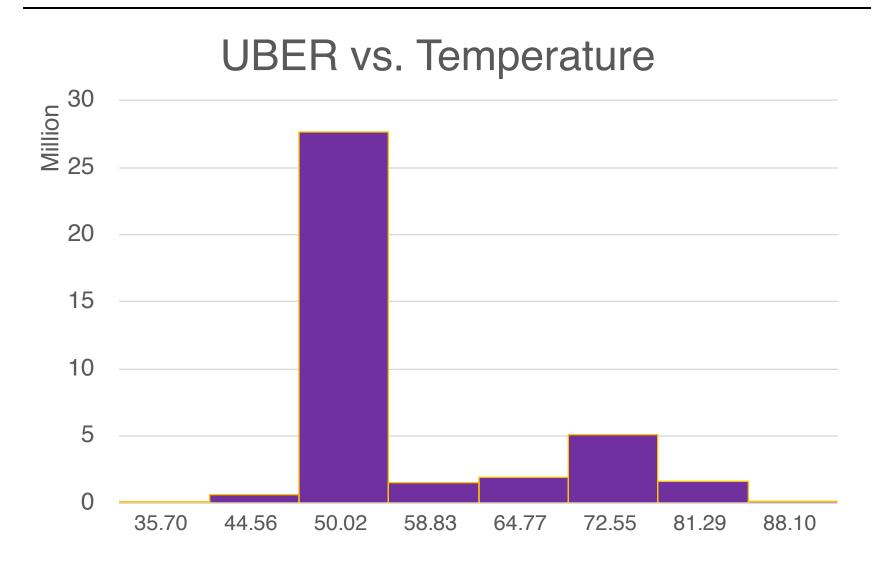
UBER vs Weather

count
18453802
9438232
4492004
2481666
1232854
947512
679642
591626
238254
105230
9952
5222

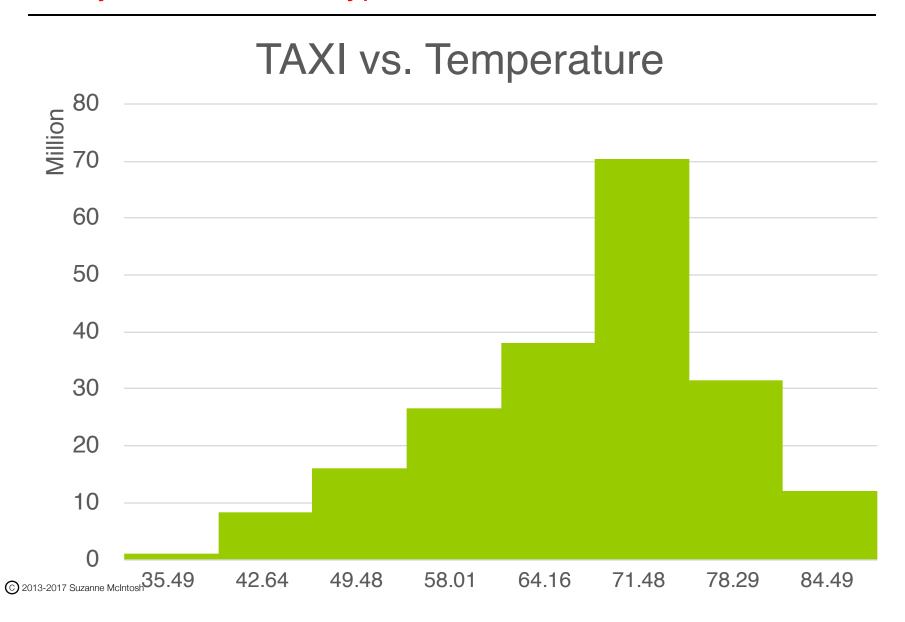
TAXI vs Weather

count
86077294
43968102
21161340
14853332
11629956
10434858
5491188
4316672
4059788
1719996
434262
102052

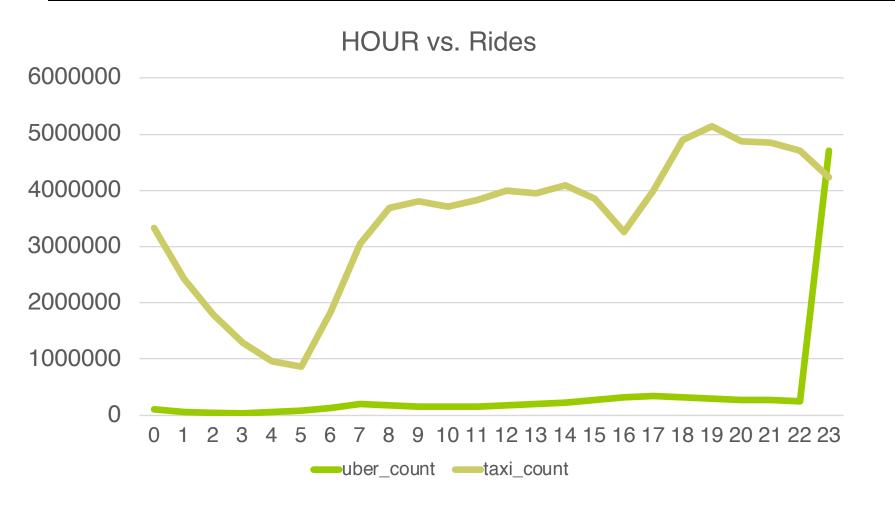
Analysis: Check the Hypothesis II



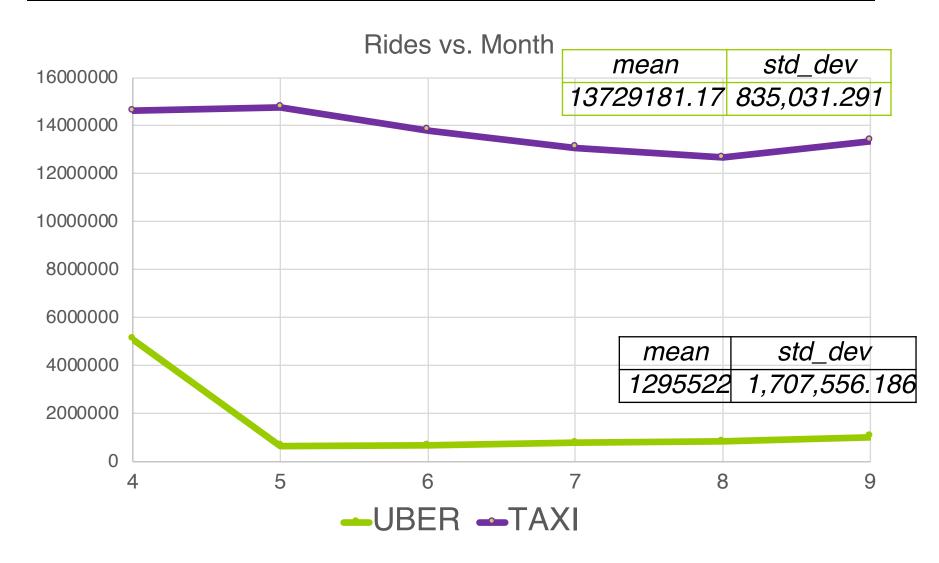
Analysis: Check the Hypothesis II



Analysis: Check the Hypothesis III



Analysis: Check the Hypothesis IV



Obstacles

- 1. <Obstacle1> Limitation of the Scope
 - > The content of Uber Dataset limits the scope of the topic.

- 2. <Obstacle 2> Factor Identifying
 - > Cannot collect all possible factors in the beginning
 - ➤ Lack of thorough domain knowledge

Evaluation

Summary

- 1. Most of hypotheses in the beginning are challenged.
- 2. There are some intrinsic factors behind the pattern.
- 3. Identified patterns can be a reference of strategy modification for both Uber and taxi companies.
- 4. Unknown factors behind the pattern: business opportunities.

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Acknowledgement

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- Santhosh Konda, High Performance Computing Center, New York University.

References

- [1] https://data.cityofnewyork.us/view/gn7m-em8n
- [2] https://github.com/fivethirtyeight/uber-tlc-foil-response/tree/master/uber-trip-dataList the work you referenced [3]
- https://www.wunderground.com/history/airport/KJFK/2014/1/2/D ailyHistory.html
- [4] Hian Chye Koh, Gerald Tan. Data Mining Applications in Healthcare. Nanyang Business School, Nanyang Technological University, Singapore. April, 2005.
- [5] Chris Rygielski, Jyun-Cheng Wang, David C. Yen. Data Mining Techniques for Customer Relationship Management. Department of DSC & MIS, Miami University, Oxford, OH, USA. Department of Information Management, National Chung-Cheng University, Taiwan. November, 2002>

Thank you!