

# New Taxi Agencies

HES-SO, Big Data Analytics

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# Dataset

Provided by **Chris Whong**, Urbanist, New York City

Trips and fares taxis in New York City (2013)

2x 12 CSV files (one per month), total of **173'179'759** records

Trips dataset: Taxi identification, start and end time, GPS coordinates

Fares dataset: Taxi identification, start time, amount, payment type, tip

# Goals

In order to open new taxi agencies and increase customer satisfaction (and profit):

- (1) Find the best spots to find customers
- (2) Identify the most profitable spots

# Preprocessing

Removed out of bound GPS coordinates (1.75% of the records)

Removed invalid fares (1% of the records)

# Machine Learning

PySpark

K-means, centroids for clusters of best locations for new agencies

Customers satisfaction:

- Using pickups coordinates

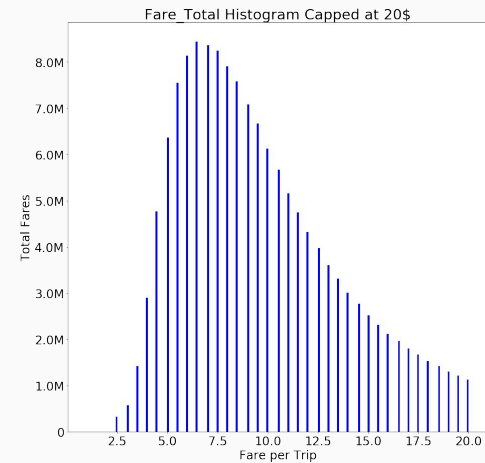
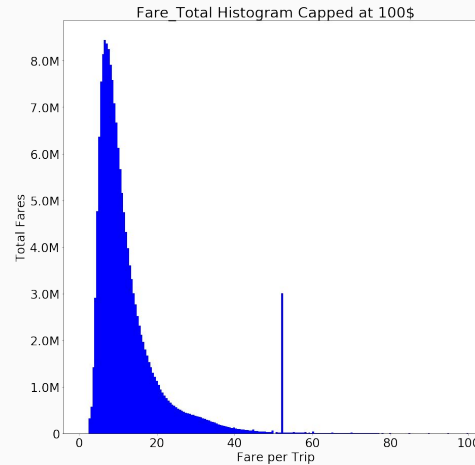
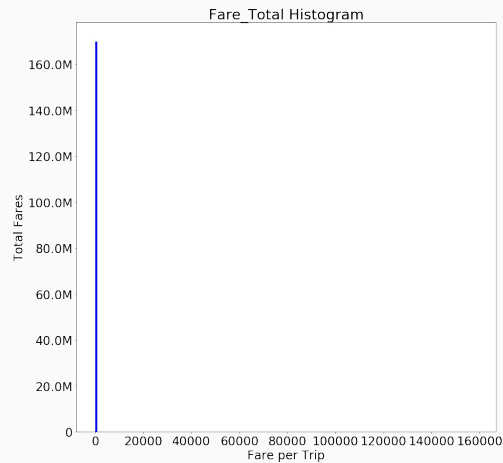
Profit:

- Using pickups coordinates
- Weighted with fare

# Machine Learning

total fare = amount of the fare + surcharge

maximum fare: 158'995.8125 \$



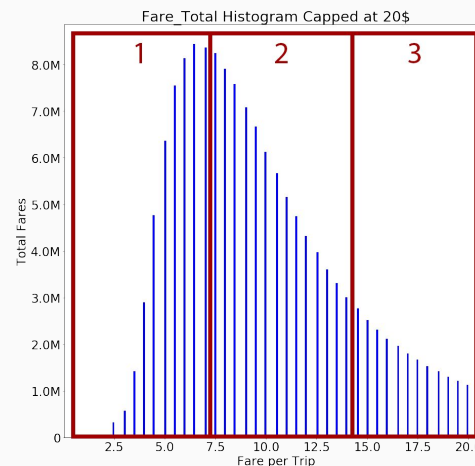
# Optimization

Fare weight: proportionally duplicated records to add attraction

- Increased the dataset size of 2'575 %

Sampling fares in groups, example:

- 05.0\$ → not cloned
- 11.5\$ → cloned once
- 20.0\$ → 3 cloned twice



# Testing and evaluation

- Custom softwares for data transformations
- Our own Spark cluster
- Try, Fails and Meta Parameterizing

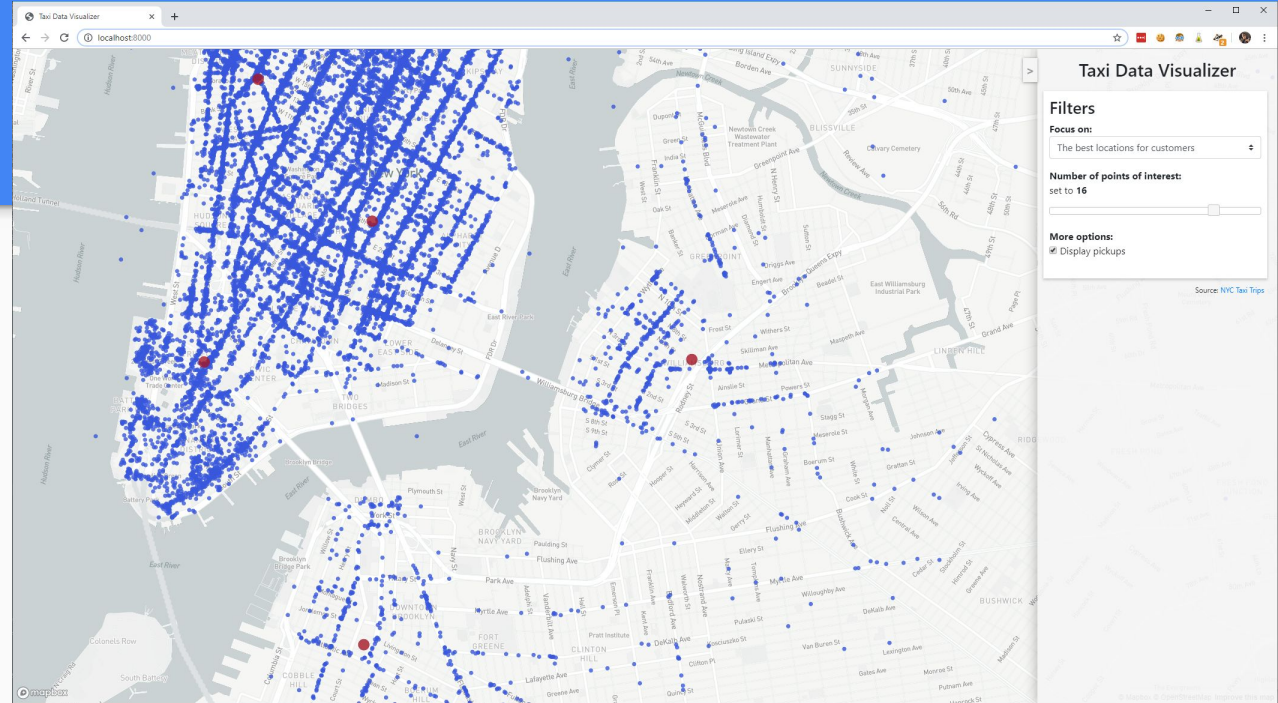
**AVERAGE DAY WITH SPARK**

```
1  [|||||] 100.0% 13 [|||||] 100.0% 25 [|||||] 100.0% 37 [|||||] 100.0%
2  [|||||] 100.0% 14 [|||||] 100.0% 26 [|||||] 100.0% 38 [|||||] 100.0%
3  [|||||] 100.0% 15 [|||||] 100.0% 27 [|||||] 100.0% 39 [|||||] 100.0%
4  [|||||] 100.0% 16 [|||||] 100.0% 28 [|||||] 100.0% 40 [|||||] 100.0%
5  [|||||] 100.0% 17 [|||||] 100.0% 29 [|||||] 100.0% 41 [|||||] 100.0%
6  [|||||] 100.0% 18 [|||||] 100.0% 30 [|||||] 100.0% 42 [|||||] 100.0%
7  [|||||] 100.0% 19 [|||||] 100.0% 31 [|||||] 100.0% 43 [|||||] 100.0%
8  [|||||] 100.0% 20 [|||||] 100.0% 32 [|||||] 100.0% 44 [|||||] 100.0%
9  [|||||] 100.0% 21 [|||||] 100.0% 33 [|||||] 100.0% 45 [|||||] 100.0%
10 [|||||] 100.0% 22 [|||||] 100.0% 34 [|||||] 100.0% 46 [|||||] 100.0%
11 [|||||] 100.0% 23 [|||||] 100.0% 35 [|||||] 100.0% 47 [|||||] 100.0%
12 [|||||] 100.0% 24 [|||||] 100.0% 36 [|||||] 100.0% 48 [|||||] 100.0%
Mem[|||||]
Swp[|||||]
10.6G/189G Tasks: 77, 394 thr, 481 kthr; 67 running
0K/28.0G Load average: 64.18 56.35 46.04
Uptime: 9 days, 06:31:08
```



# Results

Demo time



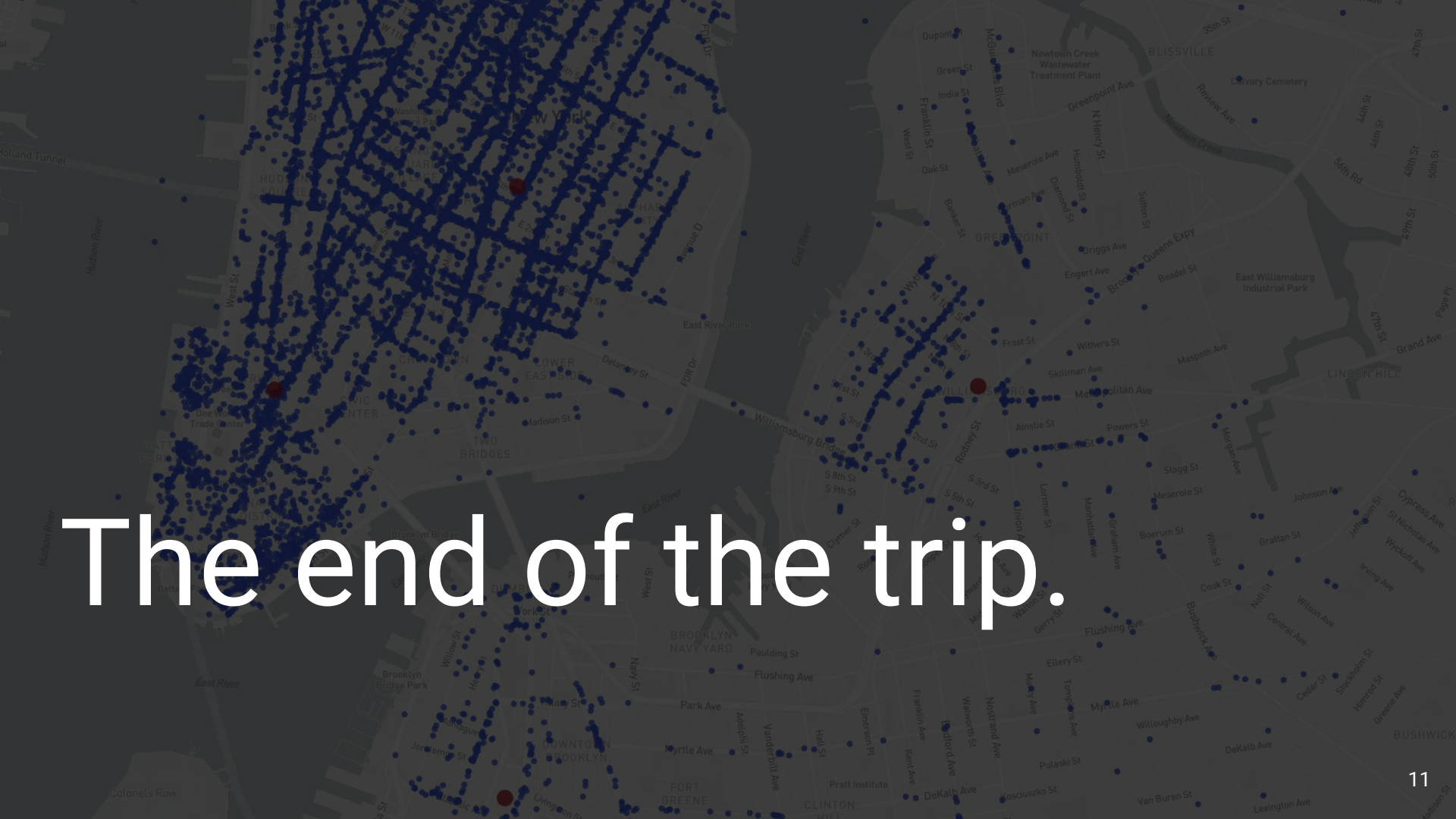
<http://bit.do/nyc-taxis>

# Conclusion

Best spots for customers are also the most profitable ones :)

Next steps:

- More processing time
- Zoom in data
- Prediction
- More data



The end of the trip.

<https://github.com/ZenLulz/hesso-bigdata-analytics>