



## **UBER TAXI FARE ANALYSIS**

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# Problem Definition

- ❑ Aim is to **compare and analyze fare data** of large number of trips for Taxi Companies mainly New York Yellow Cabs and Uber.
- ❑ We will also try **to predict the value of surcharge** at a given location and time.
- ❑ The analysis is focused on New York region only.



# Dataset



For this work we used two datasets. NYC Yellow taxi data and Uber Data.

## ❑ NYC Dataset

Pickup (Lat,Lon)	Dropoff (Lat,Lon)	Distance (miles)	Duration (seconds)	Total Fare (USD)	Time stamp
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## ❑ Uber Dataset

- ❑ We selected **popular origin, destination (OD) pairs** from the **NYC dataset** for different hour bucket and get fare data from Uber API.
- ❑ For comparison, we focused on **four** different one hour time buckets, **6** [5:30am-6:30am], **10** [9:30am-10:30am], **16** [3:30pm-4:30pm], **20** [7:30pm-8:30pm] and collected uber data for **425308 OD pairs**.

Min Fare (USD)	Low-High Estimated Fare (USD)	Distance (miles)	Duration (seconds)	Surcharge Multiplier
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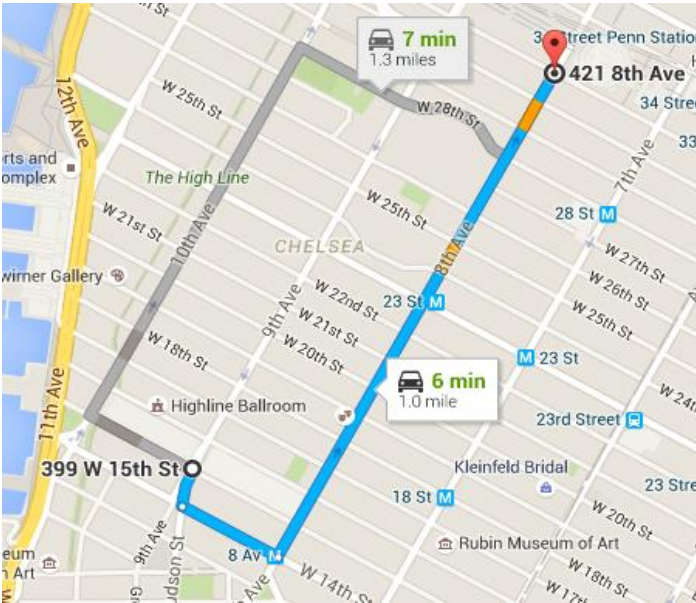
**Pickup(lat,lon) = 399 W 15th St, New York**      **Dropoff(lat,lon) = 421 8th Ave, New York**  
(40.7416561,-74.0048858)      (40.7502935,-73.9948451)

# UberX

Hour	Surcharge	Min Fare(\$)	Low-High Estimate(\$)	Distance (miles)	Duration (minutes)
6	1.5	12	12-13	1.04	5
10	1.0	8	8-9	1.06	9
16	2.2	18	18-21	1.04	11
20	1.0	8	8-10	1.30	9

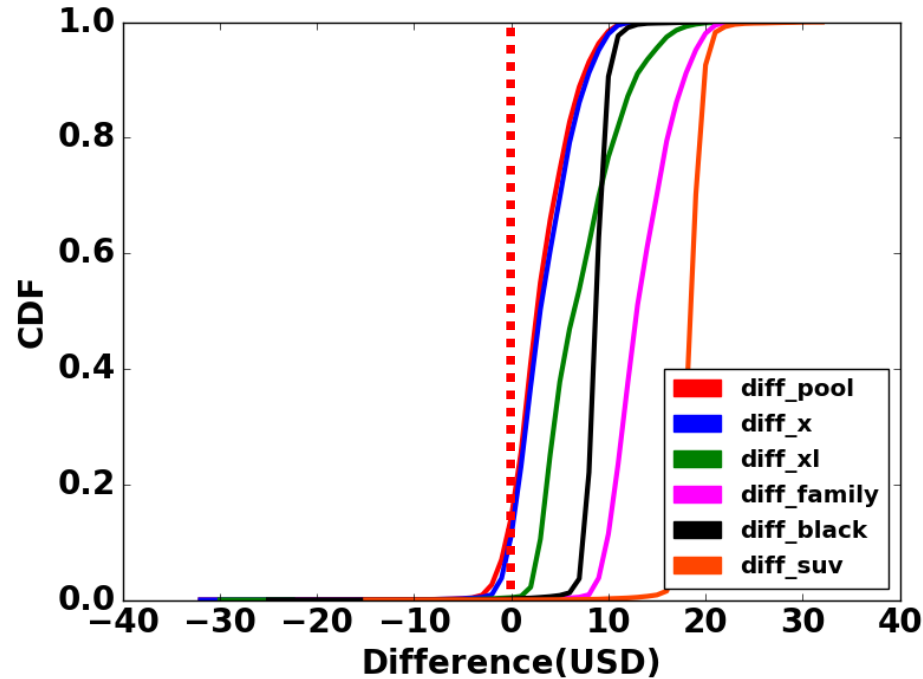
**NYC**

Hour	Trip Count	Avg. Fare(\$)	Avg. Dist (miles)	Avg. Duration (minutes)
6	31	6.2	0.95	5.5
10	105	8.5	1.02	8.2
16	121	8.9	1.05	7.0
20	239	8.1	1.06	6.2



# Fare Comparison(Cumulative Distribution Function)

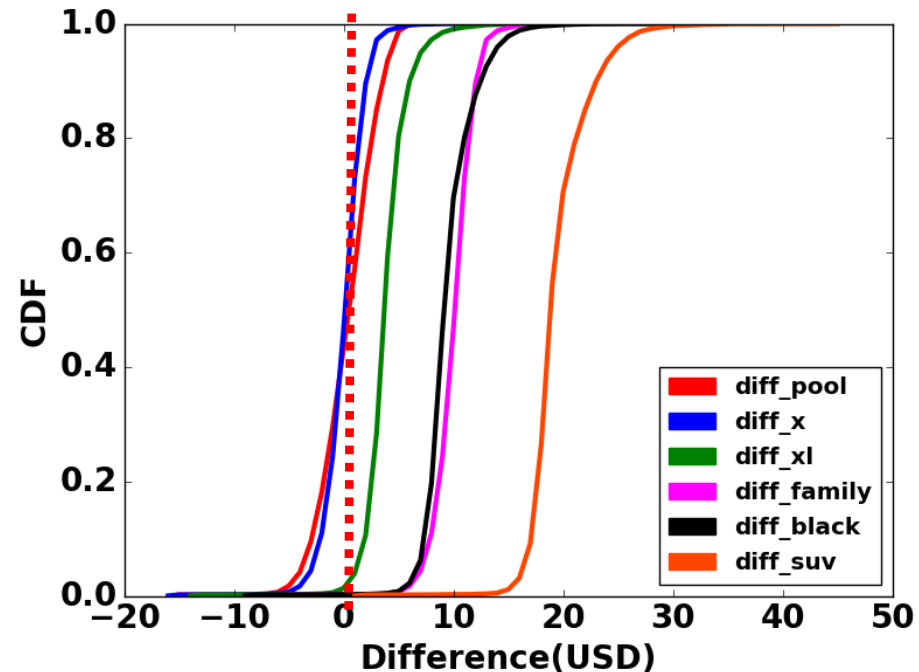
## HOUR 6



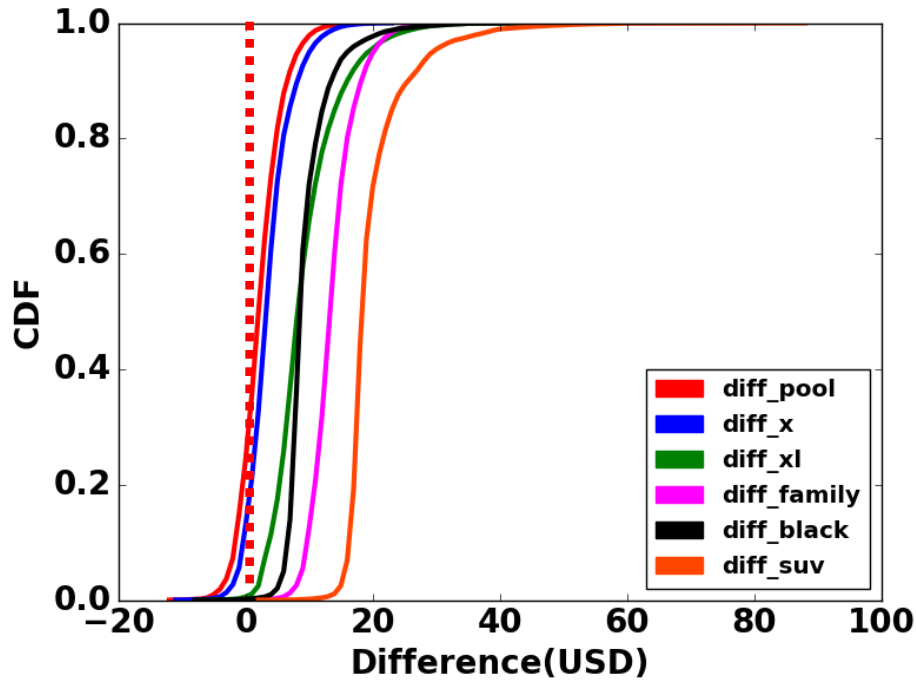
UberX costs compared to NYC cabs  
For ~50% trips UberX < NYC  
For ~50% trips UberX > NYC

Cheapest UberX is costlier than NYC  
For 90% trips UberX > NYC

## HOUR 10



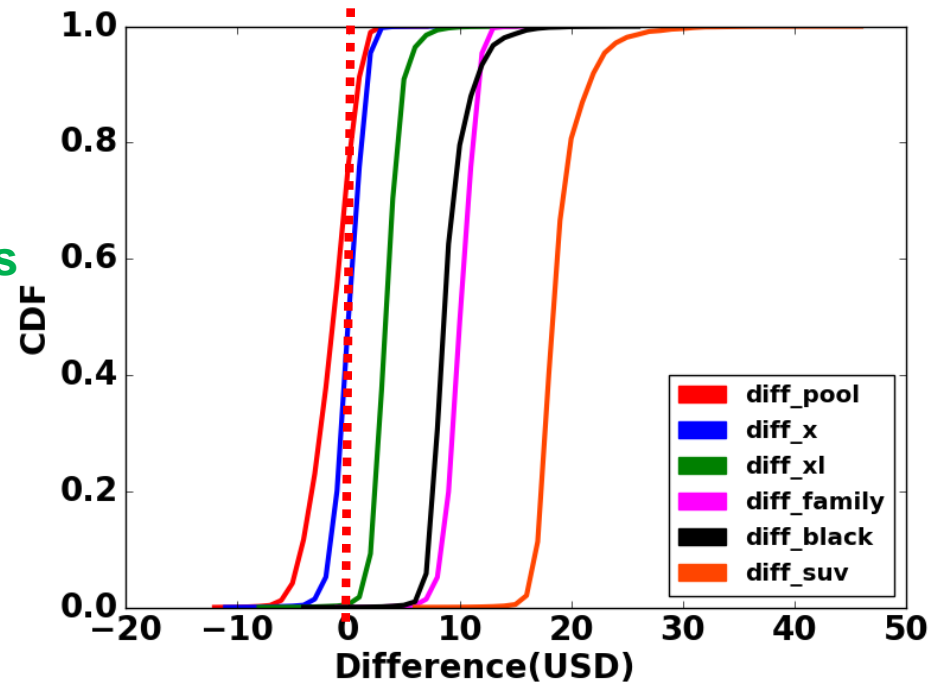
## HOURL 16



Cheapest UberX is costlier than NYC  
For ~80% trips UberX > NYC

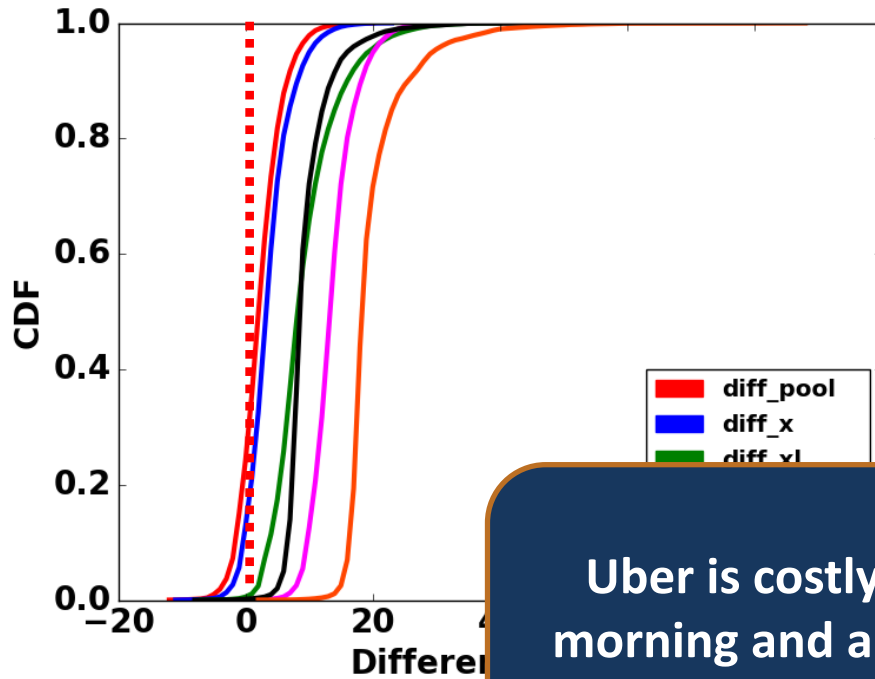
UberX costs almost equal to NYC cabs

## HOURL 20





## HOURL 16

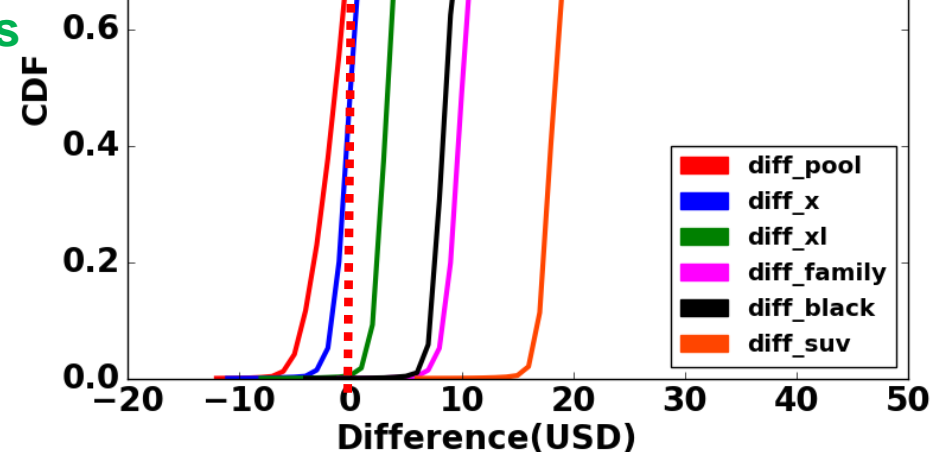


Cheapest UberX is costlier than NYC  
For ~80% trips UberX > NYC

Uber is costly during early morning and around evening time

UberX costs almost equal to NYC cabs

## HOURL 20



# Surcharge Estimator

- ➔ After doing the analysis ,we found out that uber earns most of the its money by its **dynamically changing Surcharge values**.
- ➔ Surcharge Multiplier =  
(Demand By Customers)/(Supply of Drivers).
- ➔ **Can we make a model that will predict the surcharge value giving the following inputs**
  - 1. pickup location
  - 2. dropoff location
  - 3. Hour
  - 4. cab service
- ➔ Assuming **uniform supply** of drivers. We consider, according to *Anastasios et al* [1], that surcharge only depends on the **Popularity** of the **pickup location**.

••••• AT&T LTE 10:03 PM

SURGE PRICING ✕

Demand is off the charts! Fares have increased to get more Ubers on the road.



I ACCEPT HIGHER FARE

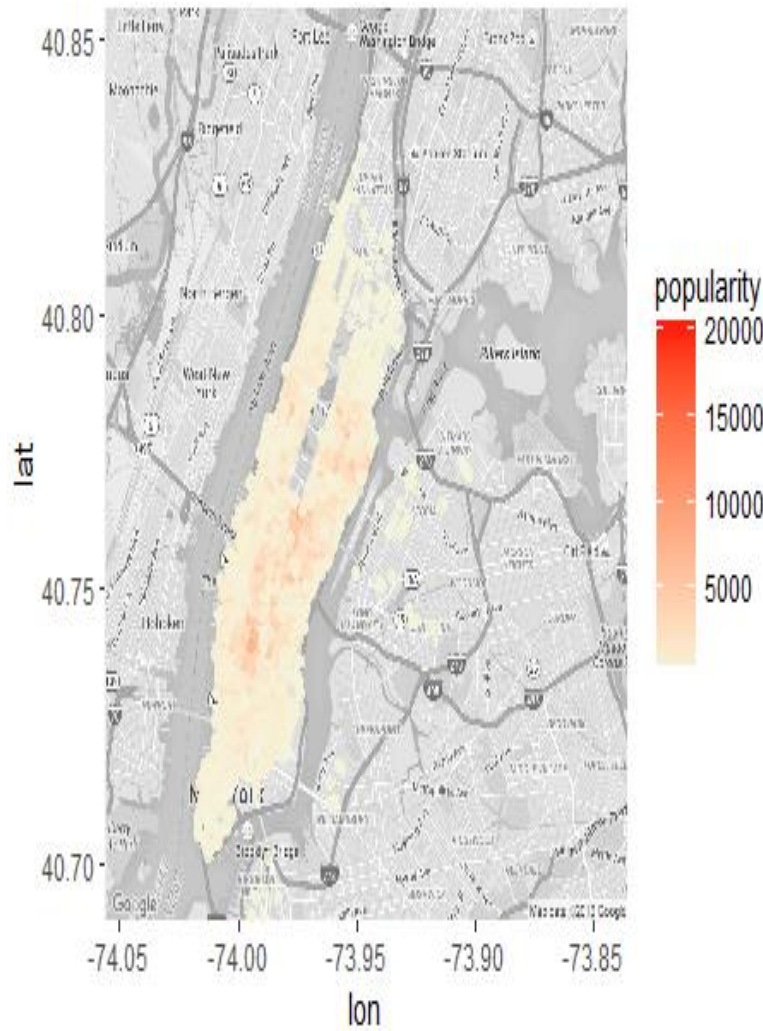
OR

NOTIFY ME IF SURGE ENDS

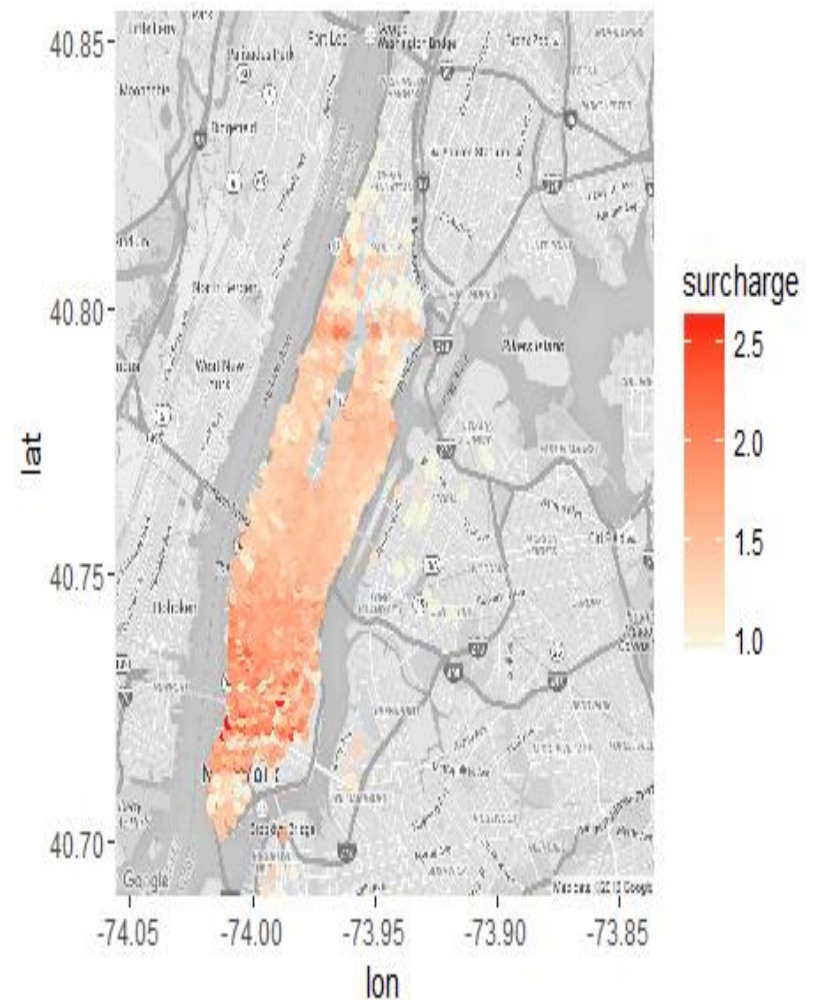
THIS RATE EXPIRES IN 2 MIN



# How the data looks at hour 16 ?



Original Popularity from NYC data

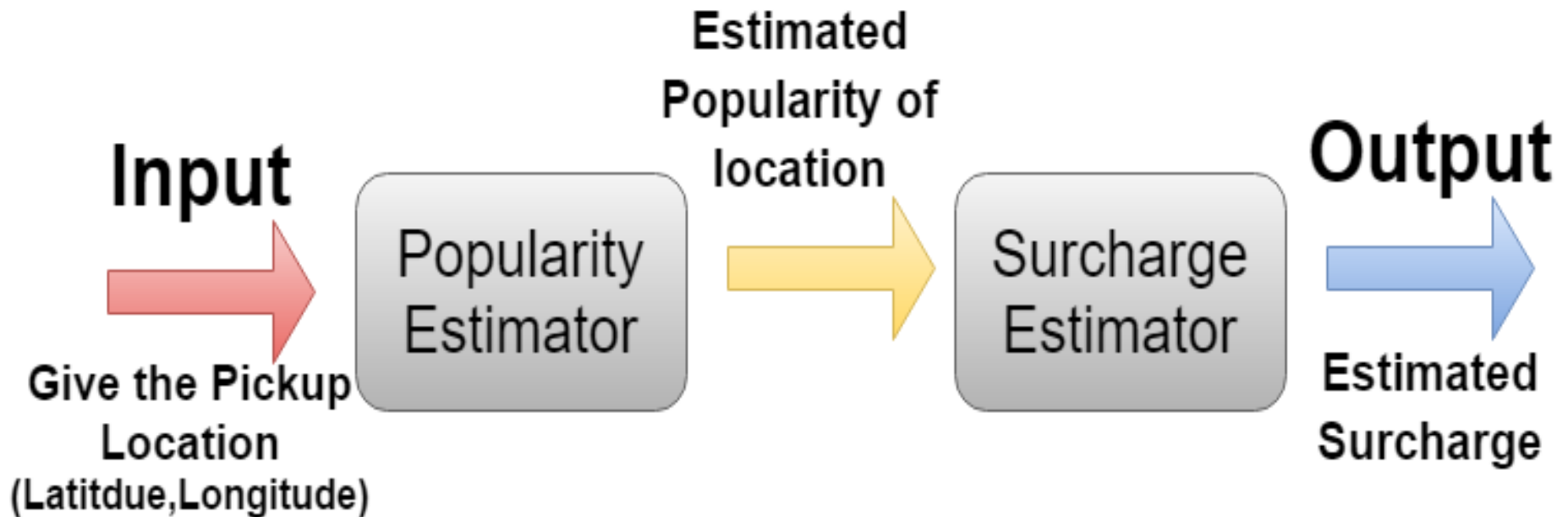


Original Surcharge from Uber

# Proposed System



**For a given Hour and Cab Service**



# PROCESS FLOW

**Dataset**

Pickup Latitude	Pickup Longitude	Popularity	Average Surcharge

Training(80%)

Test(20%)

**Popularity Estimator**

Pickup Latitude	Pickup Longitude	Popularity

1

Estimated Popularity



**Surcharge Estimator**

Pickup Latitude	Pickup Longitude	Popularity	Average Surcharge

Training(80%)

Test(20%)

2

3

Estimated Avg. Surcharge



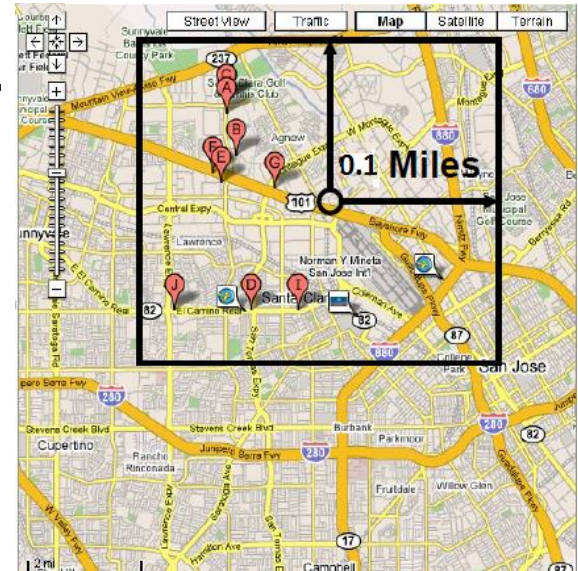
# Popularity Estimator

- The demand (popularity) at a test location depends on the popularity of neighbour.
- Start with a distance of 0.1 mile and go on searching upto 1 mile till we find at least one neighbouring location.
- We can take the average of demand of all locations within the bounding box. *This method did not give us good results.*
- Even within the bounding box ,the location nearer should have more effect or more weight.

So we use Inverse Distance Weighting Average(IDW)

$$w_i(\mathbf{x}) = \frac{1}{d(\mathbf{x}, \mathbf{x}_i)^2}$$

$$y.(\mathbf{x}) = \sum_{i=1}^d w_i x_i$$





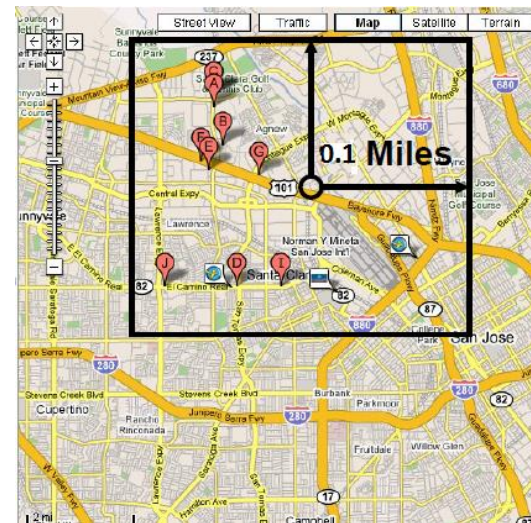
# Surcharge Estimator



We give **popularity as of a location as input** and get **estimated surcharge as output** for a fixed hour.

## Locally Weighted Regression(LWR)

- Estimated value of surcharge should depend on local regions more.
- Let  $X_0$  be the popularity of the new location.
- Find the squared difference between  $X_0$  and its neighbors. We will use this distance for the weight function.



- Find diagonal weight matrix.
- $X$  is estimated popularity
- $Y$  is surcharge

$$w_i(\mathbf{x}) = \frac{1}{d(\mathbf{x}, \mathbf{x}_i)^2}$$

$$w(x) = ae^{-\frac{(x-b)^2}{2c^2}}$$

$$\beta = (X'WX)^{-1}X'WY$$

$$\hat{y} = \beta X_0$$

# MSE for Surcharge Estimation



Hour	6	10	16	20
Linear Regression	0.0287	0.0221	0.0422	0.00239
Decision Tree Regression	0.02713	0.01980	0.03089	0.00231
KNN Regression (k- nearest neighbor)	0.02806	0.02050	0.03680	0.002720
<b>Locally Weighted Regression(LWR)</b>	<b>0.0091</b>	<b>0.0081</b>	<b>0.0101</b>	<b>0.00168</b>

For the above MSE the original range of surcharge values lie from 1 to 2.0



# Conclusion



- ✓ The proposed model estimates the average surcharge at a location with **maximum Mean Square Error of 0.01**.
- ✓ This model helps to predict the surcharge of a location for **any hour bucket** using history data. So the user can plan accordingly.
- ✓ This model also helps to compare the estimated surcharge given by Uber against the estimated surcharge of the location using history data.

*Thank You*

