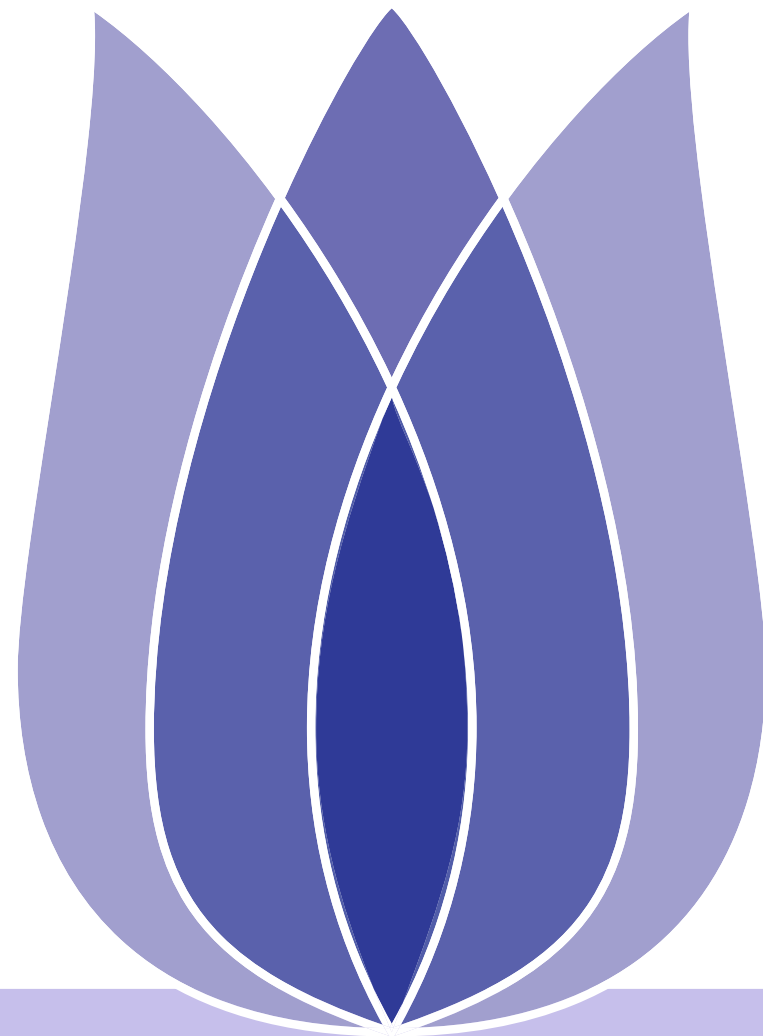


New York City Taxi Fare Prediction

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Problem

- Predicting the taxi fare according to the known features of the passengers taking taxis, which including longitude coordinate and latitude coordinate of where the taxi ride started and so on..
- Do better than this using Machine Learning techniques: an RMSE of 5–8

Evaluation

RMSE measures the difference between the predictions of a model, and the corresponding ground truth.

- RMSE is given by:

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n (\hat{y}_i - y_i)^2}$$



TULIP

Team for Universal Learning and Intelligent Processing

Dataset Description

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- Files: train.csv; test.csv; sample_submission.csv
- Data fields
 - ◆ ID: **key** - Comprised of pickup_datetime plus a unique integer.
 - ◆ Features
 - **pickup_datetime** - timestamp value indicating when the taxi ride started.
 - **pickup_longitude** - float for longitude coordinate of where the taxi ride started.
 - **pickup_latitude** - float for latitude coordinate of where the taxi ride started.
 - **dropoff_longitude** - float for longitude coordinate of where the taxi ride ended.
 - **dropoff_latitude** - float for latitude coordinate of where the taxi ride ended.
 - **passenger_count** - integer indicating the number of passengers in the taxi ride.
 - ◆ Target: **fare_amount** - float dollar amount of the cost of the taxi ride.





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Exploration and Preprocessing



Exploratory data analysis

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- Data import and preliminary processing.
 - ◆ Import the data and view it.
 - ◆ Calculating eigenvalues.minimum value, etc.
 - ◆ Preliminary processing of date types and sorting.

- Handling missing values.
 - ◆ Deleting 14 rows with missing values, resulting in 1999986 remaining rows.

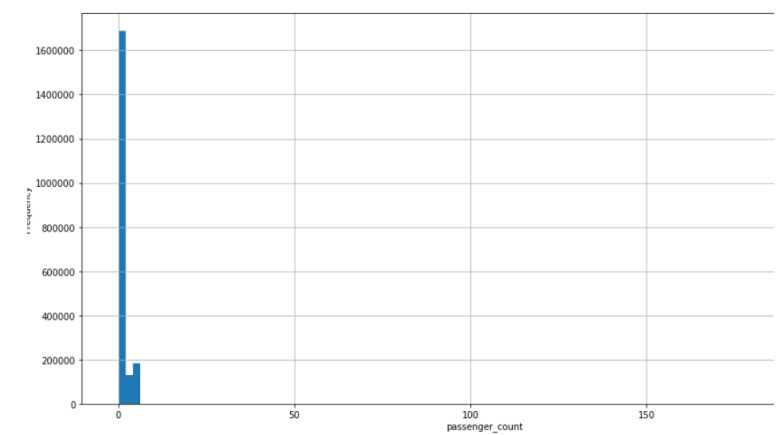
- Handling outliers.Remaining 1999822 rows.
 - ◆ The fare should be positive.Delete 77 lines with negative fares.
 - ◆ latitude:[-90,90],longitude:[-180,180].Deleting rows outside the range.
 - ◆ The number of passengers has an outlier of 208, which should be removed.



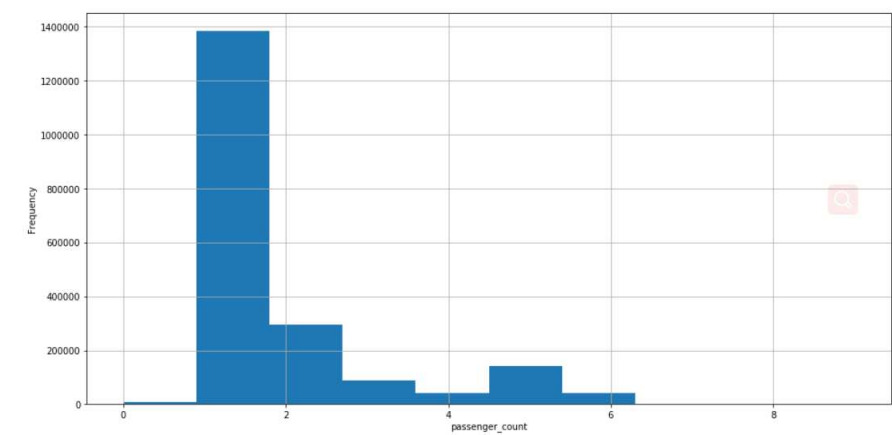


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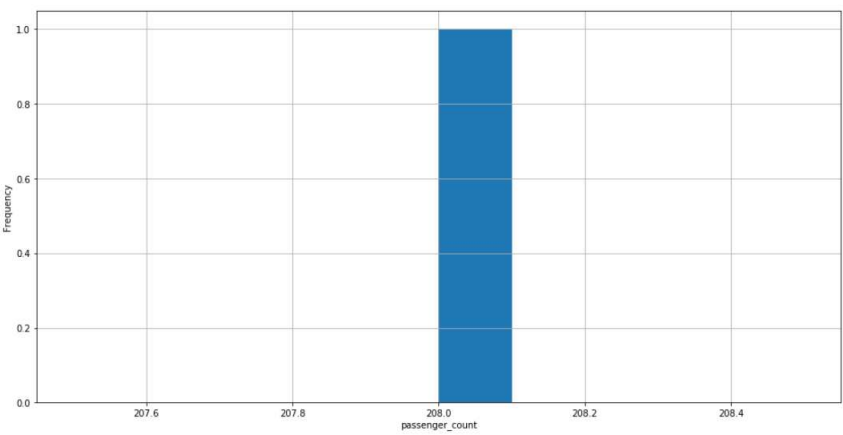
- Plot the passenger fare histogram.
Combined with the data description results of the training set, the histogram of passenger consumption in the training set was made, and the outliers of passenger consumption were screened and removed.



(a) figure 1:100 groups



(b) figure 2:10 groups,x<10



(c) figure 3:10 groups,x>=10

Figure 1: Histogram of different groups

■ Data processing and computing.

- ◆ Frame the training set range. Find the longitude and latitude range of the test set. The training dataset contains **1957,917** records after being processed.
- ◆ Calculate the distance between pick-up and drop-off points according to the Haversine Equation.

The records with zero distance between taxi fare and pick-up and drop-off location are invalid, and **1957913** training set data are obtained after deletion.





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Data Modeling

■ Data modeling.

- ◆ multiple linear regression models matrix representation:

$$Y = X\beta + \mu$$

The multiple regression equation is obtained as follows:

$$Y_i = 4.46 + 2.10X_{0_i} - 0.05X_{1_i} + 0.04X_{2_i}$$

where X_{0_i} is the value of "H_Distance", X_{1_i} is the value of "weekday+1", X_{2_i} is the value of "passenger_count".





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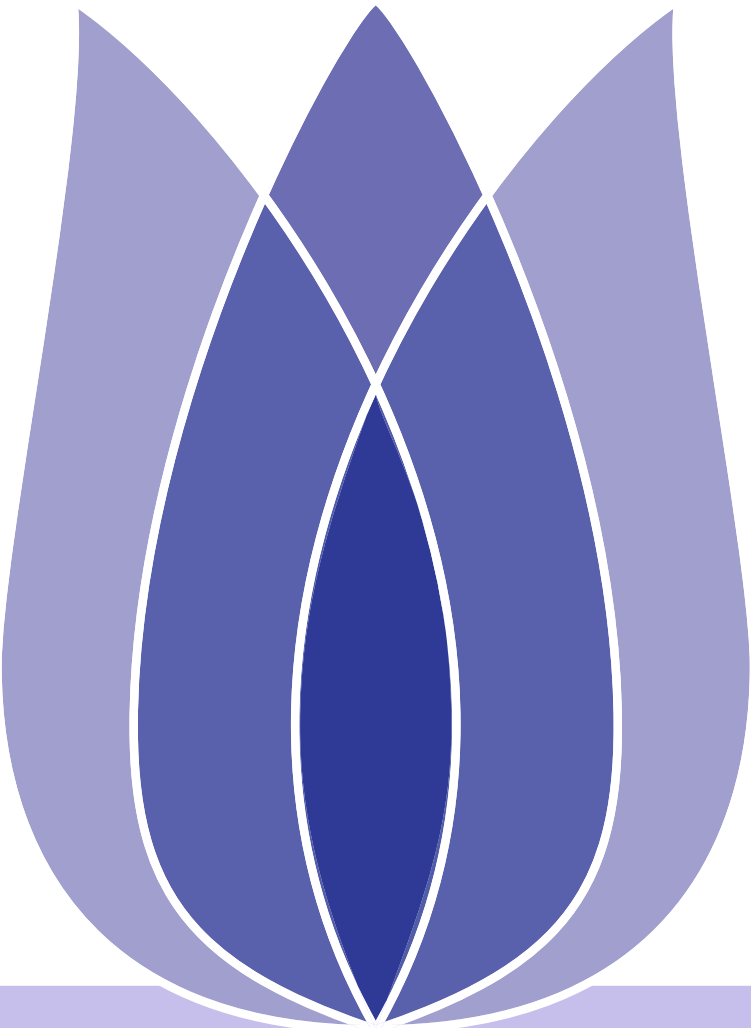
- Data to predict.
 - ◆ Putting the relevant fields of the prediction set into the regression equation yields the predicted value of "fare_amount" :

Table 1: Prediction for the field "fare_amount"

			key	fare_amount
0	2015-01-27	13:08:24.000000	200	9.29
1	2015-01-27	13:08:24.000000	300	9.50
2	2011-10-08	11:53:44.000000	200	5.51
...		



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