

Course: Laboratory Practice III

Course Code: 410246

Name: Asmeeta Lalit Bardiya

Class: BE

Roll No. : 22

Div: A

Title: Predict the price of the Uber ride from a given pickup point to the agreed drop-off location.

Perform following tasks:

1. Pre-process the dataset.
2. Identify outliers.
3. Check the correlation.
4. Implement linear regression and random forest regression models.
5. Evaluate the models and compare their respective scores like R2, RMSE, etc. Dataset link:

<https://www.kaggle.com/datasets/yasserh/uber-fares-> dataset

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
df=pd.read_csv("/content/uber.csv")
```

```
df.head()
```

	Unnamed: 0	key	fare_amount	pickup_datetime	pickup_longitude	pickup_latitude
0	24238194	2015-05-07 19:52:06.0000003	7.5	2015-05-07 19:52:06 UTC	-73.999817	-88.262153
1	27835199	2009-07-17 20:04:56.0000002	7.7	2009-07-17 20:04:56 UTC	-73.994355	-88.262153
2	44984355	2009-08-24 21:45:00.00000061	12.9	2009-08-24 21:45:00 UTC	-74.005043	-88.262153
3	25894730	2009-06-26 08:22:21.0000001	5.3	2009-06-26 08:22:21 UTC	-73.976124	-88.262153
4	17610152	2014-08-28 17:47:00.000000188	16.0	2014-08-28 17:47:00 UTC	-73.925023	-88.262153

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200000 entries, 0 to 199999
Data columns (total 9 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Unnamed: 0            200000 non-null  int64
1   key                   200000 non-null  object
2   fare_amount           200000 non-null  float64
3   pickup_datetime       200000 non-null  object
4   pickup_longitude      200000 non-null  float64
5   pickup_latitude       200000 non-null  float64
6   dropoff_longitude     199999 non-null  float64
7   dropoff_latitude      199999 non-null  float64
8   passenger_count       200000 non-null  int64
dtypes: float64(5), int64(2), object(2)
memory usage: 13.7+ MB
```

```
df.isnull().sum()
```

```
Unnamed: 0      0
key             0
fare_amount     0
pickup_datetime 0
pickup_longitude 0
pickup_latitude 0
dropoff_longitude 1
dropoff_latitude 1
passenger_count 0
dtype: int64
```

```
df.dtypes
```

```
Unnamed: 0      int64
key            object
fare_amount    float64
pickup_datetime object
pickup_longitude float64
pickup_latitude float64
dropoff_longitude float64
dropoff_latitude float64
passenger_count int64
dtype: object
```

```
#df['pickup_datetime']=pd.to_datetime(df['pickup_datetime'])
```

```
df.dtypes
```

```
Unnamed: 0      int64
key            object
fare_amount    float64
pickup_datetime object
pickup_longitude float64
pickup_latitude float64
dropoff_longitude float64
dropoff_latitude float64
```

```

passenger_count    int64
dtype: object

df['dropoff_longitude'].fillna(value=df['dropoff_longitude'].mean(),inplace=True)

```

```
df.isnull().sum()
```

```

Unnamed: 0      0
key             0
fare_amount     0
pickup_datetime 0
pickup_longitude 0
pickup_latitude 0
dropoff_longitude 0
dropoff_latitude 1
passenger_count 0
dtype: int64

```

```
df['dropoff_latitude'].fillna(value=df['dropoff_latitude'].mean(),inplace=True)
```

```
df.isnull().sum()
```

```

Unnamed: 0      0
key             0
fare_amount     0
pickup_datetime 0
pickup_longitude 0
pickup_latitude 0
dropoff_longitude 0
dropoff_latitude 0
passenger_count 0
dtype: int64

```

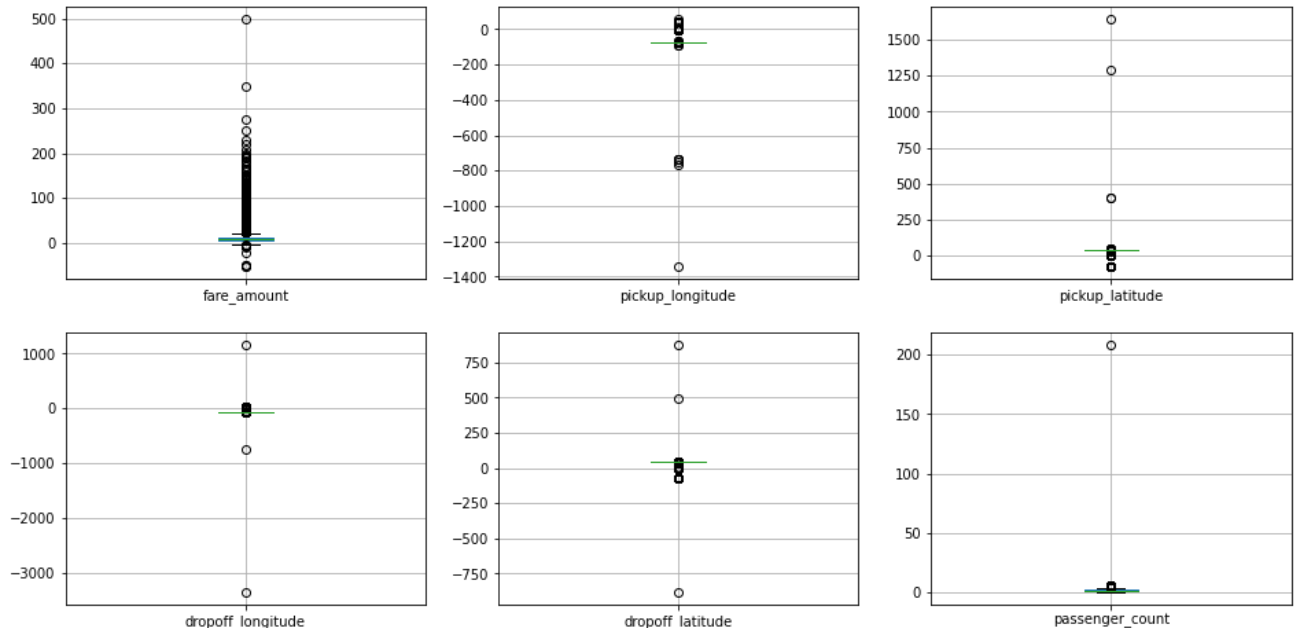
```
df.drop("Unnamed: 0",axis='columns',inplace=True)
```

```
df.head()
```

	key	fare_amount	pickup_datetime	pickup_longitude	pickup_latitude
0	2015-05-07 19:52:06.0000003	7.5	2015-05-07 19:52:06 UTC	-73.999817	40.738354
1	2009-07-17 20:04:56.0000002	7.7	2009-07-17 20:04:56 UTC	-73.994355	40.728221
2	2009-08-24 21:45:00.00000061	12.9	2009-08-24 21:45:00 UTC	-74.005043	40.740770

```
fig, axes = plt.subplots(2, 3, figsize=(16, 8))
df.boxplot(column="fare_amount", ax=axes[0,0])
df.boxplot(column="pickup_longitude", ax=axes[0,1])
df.boxplot(column="pickup_latitude", ax=axes[0,2])
df.boxplot(column="dropoff_longitude", ax=axes[1,0])
df.boxplot(column="dropoff_latitude", ax=axes[1,1])
df.boxplot(column="passenger_count", ax=axes[1,2])
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f96edb2c190>



```
q1=df['fare_amount'].quantile(0.25)
q3=df['fare_amount'].quantile(0.75)
IQR=q3-q1
LL=q1-1.5*(IQR)
UL=q3+1.5*(IQR)
print(UL,LL)
```

22.25 -3.75

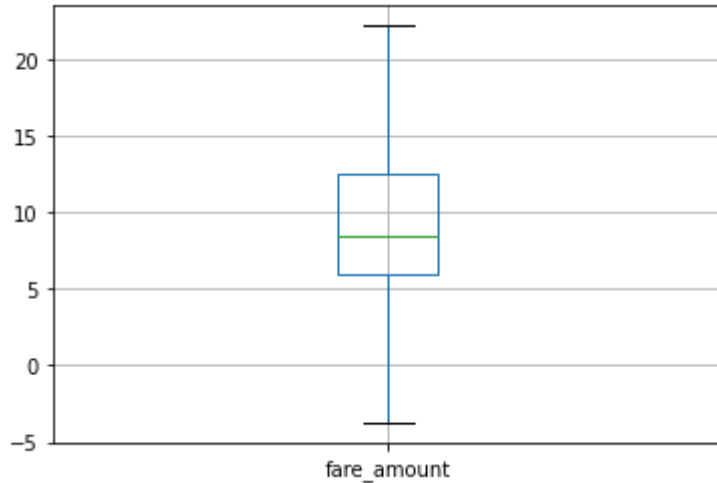
```
outlier =[]
for x in df['fare_amount']:
    if ((x> UL) or (x<LL)):
        outlier.append(x)
print(' outlier in the dataset is', outlier)
```

outlier in the dataset is [24.5, 25.7, 39.5, 29.0, 56.8, 26.1, 49.57, 30.9, 26.9, 43.5, 28.5, 27.5, 26.5, 25.5, 24.5, 23.5, 22.5, 21.5, 20.5, 19.5, 18.5, 17.5, 16.5, 15.5, 14.5, 13.5, 12.5, 11.5, 10.5, 9.5, 8.5, 7.5, 6.5, 5.5, 4.5, 3.5, 2.5, 1.5, 0.5, 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1.0, 1.1, 1.2, 1.3, 1.4, 1.5, 1.6, 1.7, 1.8, 1.9, 2.0, 2.1, 2.2, 2.3, 2.4, 2.5, 2.6, 2.7, 2.8, 2.9, 3.0, 3.1, 3.2, 3.3, 3.4, 3.5, 3.6, 3.7, 3.8, 3.9, 4.0, 4.1, 4.2, 4.3, 4.4, 4.5, 4.6, 4.7, 4.8, 4.9, 5.0, 5.1, 5.2, 5.3, 5.4, 5.5, 5.6, 5.7, 5.8, 5.9, 6.0, 6.1, 6.2, 6.3, 6.4, 6.5, 6.6, 6.7, 6.8, 6.9, 7.0, 7.1, 7.2, 7.3, 7.4, 7.5, 7.6, 7.7, 7.8, 7.9, 8.0, 8.1, 8.2, 8.3, 8.4, 8.5, 8.6, 8.7, 8.8, 8.9, 9.0, 9.1, 9.2, 9.3, 9.4, 9.5, 9.6, 9.7, 9.8, 9.9, 10.0, 10.1, 10.2, 10.3, 10.4, 10.5, 10.6, 10.7, 10.8, 10.9, 11.0, 11.1, 11.2, 11.3, 11.4, 11.5, 11.6, 11.7, 11.8, 11.9, 12.0, 12.1, 12.2, 12.3, 12.4, 12.5, 12.6, 12.7, 12.8, 12.9, 13.0, 13.1, 13.2, 13.3, 13.4, 13.5, 13.6, 13.7, 13.8, 13.9, 14.0, 14.1, 14.2, 14.3, 14.4, 14.5, 14.6, 14.7, 14.8, 14.9, 15.0, 15.1, 15.2, 15.3, 15.4, 15.5, 15.6, 15.7, 15.8, 15.9, 16.0, 16.1, 16.2, 16.3, 16.4, 16.5, 16.6, 16.7, 16.8, 16.9, 17.0, 17.1, 17.2, 17.3, 17.4, 17.5, 17.6, 17.7, 17.8, 17.9, 18.0, 18.1, 18.2, 18.3, 18.4, 18.5, 18.6, 18.7, 18.8, 18.9, 19.0, 19.1, 19.2, 19.3, 19.4, 19.5, 19.6, 19.7, 19.8, 19.9, 20.0, 20.1, 20.2, 20.3, 20.4, 20.5, 20.6, 20.7, 20.8, 20.9, 21.0, 21.1, 21.2, 21.3, 21.4, 21.5, 21.6, 21.7, 21.8, 21.9, 22.0, 22.1, 22.2, 22.3, 22.4, 22.5, 22.6, 22.7, 22.8, 22.9, 23.0, 23.1, 23.2, 23.3, 23.4, 23.5, 23.6, 23.7, 23.8, 23.9, 24.0, 24.1, 24.2, 24.3, 24.4, 24.5, 24.6, 24.7, 24.8, 24.9, 25.0, 25.1, 25.2, 25.3, 25.4, 25.5, 25.6, 25.7, 25.8, 25.9, 26.0, 26.1, 26.2, 26.3, 26.4, 26.5, 26.6, 26.7, 26.8, 26.9, 27.0, 27.1, 27.2, 27.3, 27.4, 27.5, 27.6, 27.7, 27.8, 27.9, 28.0, 28.1, 28.2, 28.3, 28.4, 28.5, 28.6, 28.7, 28.8, 28.9, 29.0, 29.1, 29.2, 29.3, 29.4, 29.5, 29.6, 29.7, 29.8, 29.9, 30.0, 30.1, 30.2, 30.3, 30.4, 30.5, 30.6, 30.7, 30.8, 30.9, 31.0, 31.1, 31.2, 31.3, 31.4, 31.5, 31.6, 31.7, 31.8, 31.9, 32.0, 32.1, 32.2, 32.3, 32.4, 32.5, 32.6, 32.7, 32.8, 32.9, 33.0, 33.1, 33.2, 33.3, 33.4, 33.5, 33.6, 33.7, 33.8, 33.9, 34.0, 34.1, 34.2, 34.3, 34.4, 34.5, 34.6, 34.7, 34.8, 34.9, 35.0, 35.1, 35.2, 35.3, 35.4, 35.5, 35.6, 35.7, 35.8, 35.9, 36.0, 36.1, 36.2, 36.3, 36.4, 36.5, 36.6, 36.7, 36.8, 36.9, 37.0, 37.1, 37.2, 37.3, 37.4, 37.5, 37.6, 37.7, 37.8, 37.9, 38.0, 38.1, 38.2, 38.3, 38.4, 38.5, 38.6, 38.7, 38.8, 38.9, 39.0, 39.1, 39.2, 39.3, 39.4, 39.5, 39.6, 39.7, 39.8, 39.9, 40.0, 40.1, 40.2, 40.3, 40.4, 40.5, 40.6, 40.7, 40.8, 40.9, 41.0, 41.1, 41.2, 41.3, 41.4, 41.5, 41.6, 41.7, 41.8, 41.9, 42.0, 42.1, 42.2, 42.3, 42.4, 42.5, 42.6, 42.7, 42.8, 42.9, 43.0, 43.1, 43.2, 43.3, 43.4, 43.5, 43.6, 43.7, 43.8, 43.9, 44.0, 44.1, 44.2, 44.3, 44.4, 44.5, 44.6, 44.7, 44.8, 44.9, 45.0, 45.1, 45.2, 45.3, 45.4, 45.5, 45.6, 45.7, 45.8, 45.9, 46.0, 46.1, 46.2, 46.3, 46.4, 46.5, 46.6, 46.7, 46.8, 46.9, 47.0, 47.1, 47.2, 47.3, 47.4, 47.5, 47.6, 47.7, 47.8, 47.9, 48.0, 48.1, 48.2, 48.3, 48.4, 48.5, 48.6, 48.7, 48.8, 48.9, 49.0, 49.1, 49.2, 49.3, 49.4, 49.5, 49.6, 49.7, 49.8, 49.9, 50.0, 50.1, 50.2, 50.3, 50.4, 50.5, 50.6, 50.7, 50.8, 50.9, 51.0, 51.1, 51.2, 51.3, 51.4, 51.5, 51.6, 51.7, 51.8, 51.9, 52.0, 52.1, 52.2, 52.3, 52.4, 52.5, 52.6, 52.7, 52.8, 52.9, 53.0, 53.1, 53.2, 53.3, 53.4, 53.5, 53.6, 53.7, 53.8, 53.9, 54.0, 54.1, 54.2, 54.3, 54.4, 54.5, 54.6, 54.7, 54.8, 54.9, 55.0, 55.1, 55.2, 55.3, 55.4, 55.5, 55.6, 55.7, 55.8, 55.9, 56.0, 56.1, 56.2, 56.3, 56.4, 56.5, 56.6, 56.7, 56.8, 56.9, 57.0, 57.1, 57.2, 57.3, 57.4, 57.5, 57.6, 57.7, 57.8, 57.9, 58.0, 58.1, 58.2, 58.3, 58.4, 58.5, 58.6, 58.7, 58.8, 58.9, 59.0, 59.1, 59.2, 59.3, 59.4, 59.5, 59.6, 59.7, 59.8, 59.9, 60.0, 60.1, 60.2, 60.3, 60.4, 60.5, 60.6, 60.7, 60.8, 60.9, 61.0, 61.1, 61.2, 61.3, 61.4, 61.5, 61.6, 61.7, 61.8, 61.9, 62.0, 62.1, 62.2, 62.3, 62.4, 62.5, 62.6, 62.7, 62.8, 62.9, 63.0, 63.1, 63.2, 63.3, 63.4, 63.5, 63.6, 63.7, 63.8, 63.9, 64.0, 64.1, 64.2, 64.3, 64.4, 64.5, 64.6, 64.7, 64.8, 64.9, 65.0, 65.1, 65.2, 65.3, 65.4, 65.5, 65.6, 65.7, 65.8, 65.9, 66.0, 66.1, 66.2, 66.3, 66.4, 66.5, 66.6, 66.7, 66.8, 66.9, 67.0, 67.1, 67.2, 67.3, 67.4, 67.5, 67.6, 67.7, 67.8, 67.9, 68.0, 68.1, 68.2, 68.3, 68.4, 68.5, 68.6, 68.7, 68.8, 68.9, 69.0, 69.1, 69.2, 69.3, 69.4, 69.5, 69.6, 69.7, 69.8, 69.9, 70.0, 70.1, 70.2, 70.3, 70.4, 70.5, 70.6, 70.7, 70.8, 70.9, 71.0, 71.1, 71.2, 71.3, 71.4, 71.5, 71.6, 71.7, 71.8, 71.9, 72.0, 72.1, 72.2, 72.3, 72.4, 72.5, 72.6, 72.7, 72.8, 72.9, 73.0, 73.1, 73.2, 73.3, 73.4, 73.5, 73.6, 73.7, 73.8, 73.9, 74.0, 74.1, 74.2, 74.3, 74.4, 74.5, 74.6, 74.7, 74.8, 74.9, 75.0, 75.1, 75.2, 75.3, 75.4, 75.5, 75.6, 75.7, 75.8, 75.9, 76.0, 76.1, 76.2, 76.3, 76.4, 76.5, 76.6, 76.7, 76.8, 76.9, 77.0, 77.1, 77.2, 77.3, 77.4, 77.5, 77.6, 77.7, 77.8, 77.9, 78.0, 78.1, 78.2, 78.3, 78.4, 78.5, 78.6, 78.7, 78.8, 78.9, 79.0, 79.1, 79.2, 79.3, 79.4, 79.5, 79.6, 79.7, 79.8, 79.9, 80.0, 80.1, 80.2, 80.3, 80.4, 80.5, 80.6, 80.7, 80.8, 80.9, 81.0, 81.1, 81.2, 81.3, 81.4, 81.5, 81.6, 81.7, 81.8, 81.9, 82.0, 82.1, 82.2, 82.3, 82.4, 82.5, 82.6, 82.7, 82.8, 82.9, 83.0, 83.1, 83.2, 83.3, 83.4, 83.5, 83.6, 83.7, 83.8, 83.9, 84.0, 84.1, 84.2, 84.3, 84.4, 84.5, 84.6, 84.7, 84.8, 84.9, 85.0, 85.1, 85.2, 85.3, 85.4, 85.5, 85.6, 85.7, 85.8, 85.9, 86.0, 86.1, 86.2, 86.3, 86.4, 86.5, 86.6, 86.7, 86.8, 86.9, 87.0, 87.1, 87.2, 87.3, 87.4, 87.5, 87.6, 87.7, 87.8, 87.9, 88.0, 88.1, 88.2, 88.3, 88.4, 88.5, 88.6, 88.7, 88.8, 88.9, 89.0, 89.1, 89.2, 89.3, 89.4, 89.5, 89.6, 89.7, 89.8, 89.9, 90.0, 90.1, 90.2, 90.3, 90.4, 90.5, 90.6, 90.7, 90.8, 90.9, 91.0, 91.1, 91.2, 91.3, 91.4, 91.5, 91.6, 91.7, 91.8, 91.9, 92.0, 92.1, 92.2, 92.3, 92.4, 92.5, 92.6, 92.7, 92.8, 92.9, 93.0, 93.1, 93.2, 93.3, 93.4, 93.5, 93.6, 93.7, 93.8, 93.9, 94.0, 94.1, 94.2, 94.3, 94.4, 94.5, 94.6, 94.7, 94.8, 94.9, 95.0, 95.1, 95.2, 95.3, 95.4, 95.5, 95.6, 95.7, 95.8, 95.9, 96.0, 96.1, 96.2, 96.3, 96.4, 96.5, 96.6, 96.7, 96.8, 96.9, 97.0, 97.1, 97.2, 97.3, 97.4, 97.5, 97.6, 97.7, 97.8, 97.9, 98.0, 98.1, 98.2, 98.3, 98.4, 98.5, 98.6, 98.7, 98.8, 98.9, 99.0, 99.1, 99.2, 99.3, 99.4, 99.5, 99.6, 99.7, 99.8, 99.9, 100.0, 100.1, 100.2, 100.3, 100.4, 100.5, 100.6, 100.7, 100.8, 100.9, 101.0, 101.1, 101.2, 101.3, 101.4, 101.5, 101.6, 101.7, 101.8, 101.9, 102.0, 102.1, 102.2, 102.3, 102.4, 102.5, 102.6, 102.7, 102.8, 102.9, 103.0, 103.1, 103.2, 103.3, 103.4, 103.5, 103.6, 103.7, 103.8, 103.9, 104.0, 104.1, 104.2, 104.3, 104.4, 104.5, 104.6, 104.7, 104.8, 104.9, 105.0, 105.1, 105.2, 105.3, 105.4, 105.5, 105.6, 105.7, 105.8, 105.9, 106.0, 106.1, 106.2, 106.3, 106.4, 106.5, 106.6, 106.7, 106.8, 106.9, 107.0, 107.1, 107.2, 107.3, 107.4, 107.5, 107.6, 107.7, 107.8, 107.9, 108.0, 108.1, 108.2, 108.3, 108.4, 108.5, 108.6, 108.7, 108.8, 108.9, 109.0, 109.1, 109.2, 109.3, 109.4, 109.5, 109.6, 109.7, 109.8, 109.9, 110.0, 110.1, 110.2, 110.3, 110.4, 110.5, 110.6, 110.7, 110.8, 110.9, 111.0, 111.1, 111.2, 111.3, 111.4, 111.5, 111.6, 111.7, 111.8, 111.9, 112.0, 112.1, 112.2, 112.3, 112.4, 112.5, 112.6, 112.7, 112.8, 112.9, 113.0, 113.1, 113.2, 113.3, 113.4, 113.5, 113.6, 113.7, 113.8, 113.9, 114.0, 114.1, 114.2, 114.3, 114.4, 114.5, 114.6, 114.7, 114.8, 114.9, 115.0, 115.1, 115.2, 115.3, 115.4, 115.5, 115.6, 115.7, 115.8, 115.9, 116.0, 116.1, 116.2, 116.3, 116.4, 116.5, 116.6, 116.7, 116.8, 116.9, 117.0, 117.1, 117.2, 117.3, 117.4, 117.5, 117.6, 117.7, 117.8, 117.9, 118.0, 118.1, 118.2, 118.3, 118.4, 118.5, 118.6, 118.7, 118.8, 118.9, 119.0, 119.1, 119.2, 119.3, 119.4, 119.5, 119.6, 119.7, 119.8, 119.9, 120.0, 120.1, 120.2, 120.3, 120.4, 120.5, 120.6, 120.7, 120.8, 120.9, 121.0, 121.1, 121.2, 121.3, 121.4, 121.5, 121.6, 121.7, 121.8, 121.9, 122.0, 122.1, 122.2, 122.3, 122.4, 122.5, 122.6, 122.7, 122.8, 122.9, 123.0, 123.1, 123.2, 123.3, 123.4, 123.5, 123.6, 123.7, 123.8, 123.9, 124.0, 124.1, 124.2, 124.3, 124.4, 124.5, 124.6, 124.7, 124.8, 124.9, 125.0, 125.1, 125.2, 125.3, 125.4, 125.5, 125.6, 125.7, 125.8, 125.9, 126.0, 126.1, 126.2, 126.3, 126.4, 126.5, 126.6, 126.7, 126.8, 126.9, 127.0, 127.1, 127.2, 127.3, 127.4, 127.5, 127.6, 127.7, 127.8, 127.9, 128.0, 128.1, 128.2, 128.3, 128.4, 128.5, 128.6, 128.7, 128.8, 128.9, 129.0, 129.1, 129.2, 129.3, 129.4, 129.5, 129.6, 129.7, 129.8, 129.9, 130.0, 130.1, 130.2, 130.3, 130.4, 130.5, 130.6, 130.7, 130.8, 130.9, 131.0, 131.1, 131.2, 131.3, 131.4, 131.5, 131.6, 131.7, 131.8, 131.9, 132.0, 132.1, 132.2, 132.3, 132.4, 132.5, 132.6, 132.7, 132.8, 132.9, 133.0, 133.1, 133.2, 133.3, 133.4, 133.5, 133.6, 133.7, 133.8, 133.9, 134.0, 134.1, 134.2, 134.3, 134.4, 134.5, 134.6, 134.7, 134.8, 134.9, 135.0, 135.1, 135.2, 135.3, 135.4, 135.5, 135.6, 135.7, 135.8, 135.9, 136.0, 136.1, 136.2, 136.3, 136.4, 136.5, 136.6, 136.7, 136.8, 136.9, 137.0, 137.1, 137.2, 137.3, 137.4, 137.5, 137.6, 137.7, 137.8, 137.9, 138.0, 138.1, 138.2, 138.3, 138.4, 138.5, 138.6, 138.7, 138.8, 138.9, 139.0, 139.1, 139.2, 139.3, 139.4, 139.5, 139.6, 139.7, 139.8, 139.9, 140.0, 140.1, 140.2, 140.3, 140.4, 140.5, 140.6, 140.7, 140.8, 140.9, 141.0, 141.1, 141.2, 141.3, 141.4, 141.5, 141.6, 141.7, 141.8, 141.9, 142.0, 142.1, 142.2, 142.3, 142.4, 142.5, 142.6, 142.7, 142.8, 142.9, 143.0, 143.1, 143.2, 143.3, 143.4, 143.5, 143.6, 143.7, 143.8, 143.9, 144.0, 144.1, 144.2, 144.3, 144.4, 144.5, 144.6, 144.7, 144.8, 144.9, 145.0, 145.1, 145.2, 145.3, 145.4, 145.5, 145.6, 145.7, 145.8, 145.9, 146.0, 146.1, 146.2, 146.3, 146.4, 146.5, 146.6, 146.7, 146.8, 146.9, 147.0, 147.1, 147.2, 147.3, 147.4, 147.5, 147.6, 147.7, 147.8, 147.9, 148.0, 148.1, 148.2, 148.3, 148.4, 148.5, 148.6, 148.7, 148.8, 148.9, 149.0, 149.1, 149.2, 149.3, 149.4, 149.5, 149.6, 149.7, 149.8, 149.9, 150.0, 150.1, 150.2, 150.3, 150.4, 150.5, 150.6, 150.7, 150.8, 150.9, 151.0, 151.1, 151.2, 151.3, 151.4, 151.5, 151.6, 151.7, 151.8, 151.9, 152.0, 152.1, 152.2, 152.3, 152.4, 152.5, 152.6, 152.7, 152.8, 152.9, 153.0, 153.1, 153.2, 153.3, 153.4, 153.5, 153.6, 153.7, 153.8, 153.9, 154.0, 154.1, 154.2, 154.3, 154.4, 154.5, 154.6, 154.7, 154.8, 154.9, 155.0, 155.1, 155.2, 155.3, 155.4, 155.5, 155.6, 155.7, 155.8, 155.9, 156.0, 156.1, 156.2, 156.3, 156.4, 156.5, 156.6, 156.7, 156.8, 156.9, 157.0, 157.1, 157.2, 157.3, 157.4, 157.5, 157.6, 157.7, 157.8, 157.9, 158.0, 158.1, 158.2, 158.3, 158.4, 158.5, 158.6, 158.7, 158.8, 158.9, 159.0, 159.1, 159.2, 159.3, 159.4, 159.5, 159.6, 159.7, 159.8, 159.9, 160.0, 160.1, 160.2, 160.3, 160.4, 160.5, 160.6, 160.7, 160.8, 160.9, 161.0, 161.1, 161.2, 161.3, 161.4, 161.5, 161.6, 161.7, 161.8, 161.9, 162.0, 162.1, 162.2, 162.3, 162.4, 162.5, 162.6, 162.7, 162.8, 162.9, 163.0, 163.1, 163.2, 163.3, 163.4, 163.5, 163.6, 163.7, 163.8, 163.9, 164.0, 164.1, 164.2, 164.3, 164.4, 164.5, 164.6, 164.7, 164.8, 164.9, 165.0, 165.1, 165.2, 165.3, 165.4, 165.5, 165.6, 165.7, 165.8, 165.9, 166.0, 166.1, 166.2, 166.3, 166.4, 166.5, 166.6, 166.7, 166.8, 166.9, 167.0, 167.1, 167.2, 167.3, 167.4, 167.5, 167.6, 167.7, 167.8, 167.9, 168.0, 168.1, 168.2, 168.3, 168.4, 168.5, 168.6, 168.7, 168.8, 168.9, 169.0, 169.1, 169.2, 169.3, 169.4, 169.5, 169.6, 169.7, 169.8, 169.9, 170.0, 170.1, 170.2, 170.3, 170.4, 170.5, 170.6, 170.7, 170.8, 170.9, 171.0, 171.1, 171.2, 171.3, 171.4, 171.5, 171.6, 171.7, 171.8, 171.9, 172.0, 172.1, 172.2, 172.3, 172.4, 172.5, 172.6, 172.7, 172.8, 172.9, 173.0, 173.1, 173.2, 173.3, 173.4, 173.5, 173.6, 173.7, 173.8, 173.9, 174.0, 174.1, 174.2, 174.3, 174.4, 174.5,

```
df['fare_amount']=np.where(df['fare_amount'] <=LL, LL,df['fare_amount'])
df['fare_amount']=np.where(df['fare_amount'] >=UL, UL,df['fare_amount'])
```

```
df.boxplot(column='fare_amount')
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f96ecce3e50>

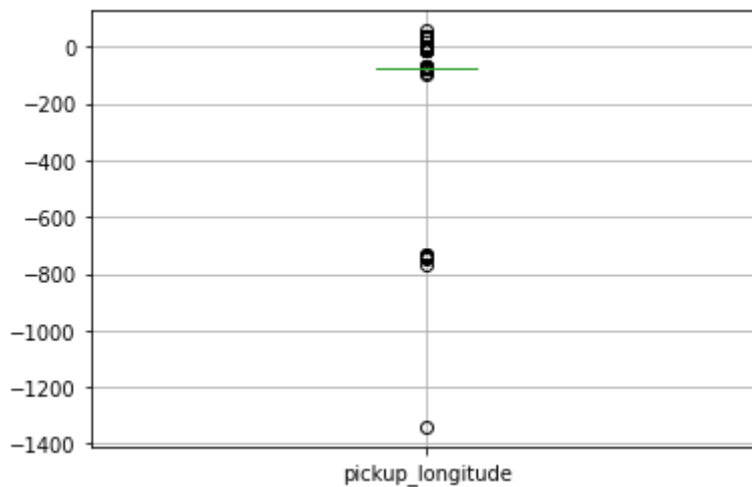


```
q1=df['pickup_longitude'].quantile(0.25)
q3=df['pickup_longitude'].quantile(0.75)
IQR=q3-q1
LL=q1-1.5*(IQR)
UL=q3+1.5*(IQR)
print(UL,LL)
```

-73.92978625000003 -74.02943224999999

```
df.boxplot(column='pickup_longitude')
```

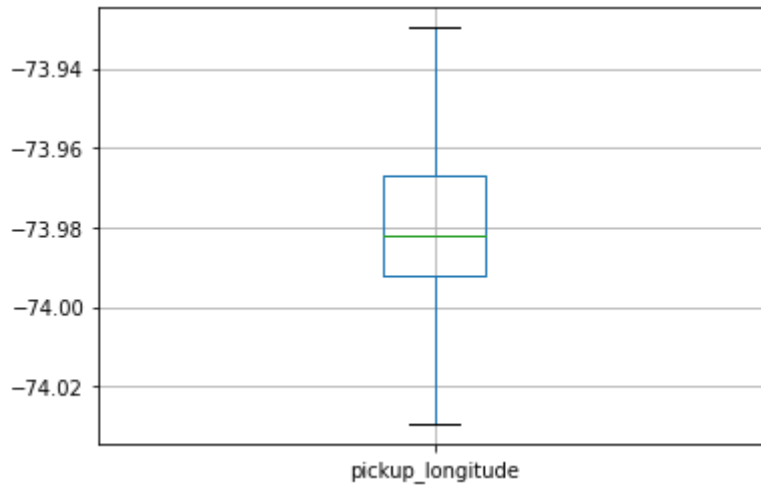
<matplotlib.axes.\_subplots.AxesSubplot at 0x7f96f045a9d0>



```
df['pickup_longitude']=np.where(df['pickup_longitude'] <=LL, LL,df['pickup_longitude'])
df['pickup_longitude']=np.where(df['pickup_longitude'] >=UL, UL,df['pickup_longitude'])
```

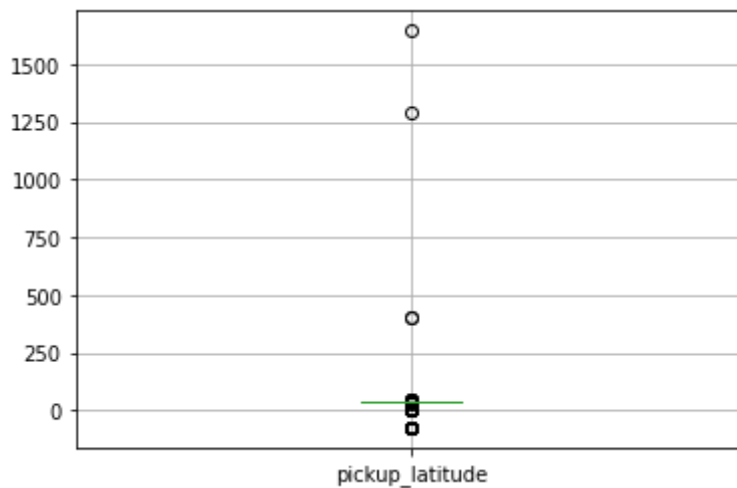
```
df.boxplot(column='pickup_longitude')
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f96eb341a50>



```
df.boxplot(column='pickup_latitude')
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f96eb2ffc50>



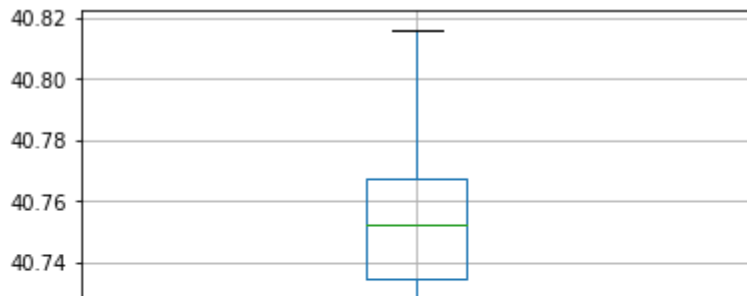
```
q1=df['pickup_latitude'].quantile(0.25)
q3=df['pickup_latitude'].quantile(0.75)
IQR=q3-q1
LL=q1-1.5*(IQR)
UL=q3+1.5*(IQR)
print(UL,LL)
```

```
40.815701375 40.68625237500001
```

```
df['pickup_latitude']=np.where(df['pickup_latitude'] <=LL, LL,df['pickup_latitude'])
df['pickup_latitude']=np.where(df['pickup_latitude'] >=UL, UL,df['pickup_latitude'])
```

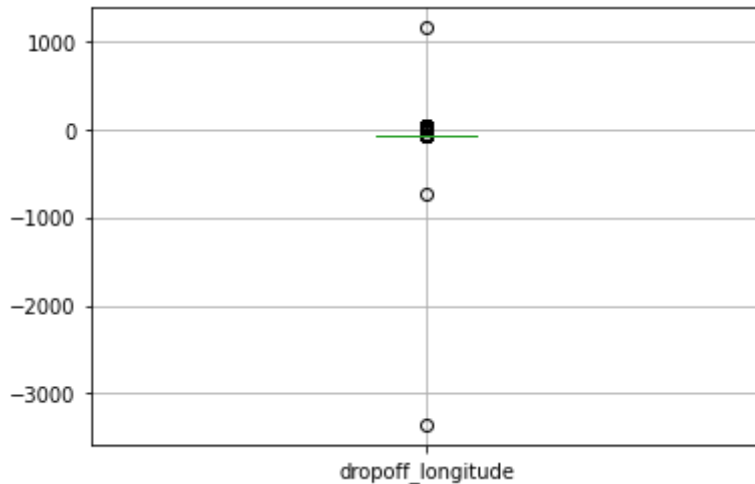
```
df.boxplot(column='pickup_latitude')
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f96eb187c10>



```
df.boxplot('dropoff_longitude')
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f96eb18c090>



```
q1=df['dropoff_longitude'].quantile(0.25)
q3=df['dropoff_longitude'].quantile(0.75)
IQR=q3-q1
LL=q1-1.5*(IQR)
UL=q3+1.5*(IQR)
print(UL,LL)
```

```
-73.9220345 -74.0330305
```

```
df['dropoff_longitude']=np.where(df['dropoff_longitude'] <=LL, LL,df['dropoff_longitude'])
df['dropoff_longitude']=np.where(df['dropoff_longitude'] >=UL, UL,df['dropoff_longitude'])
```

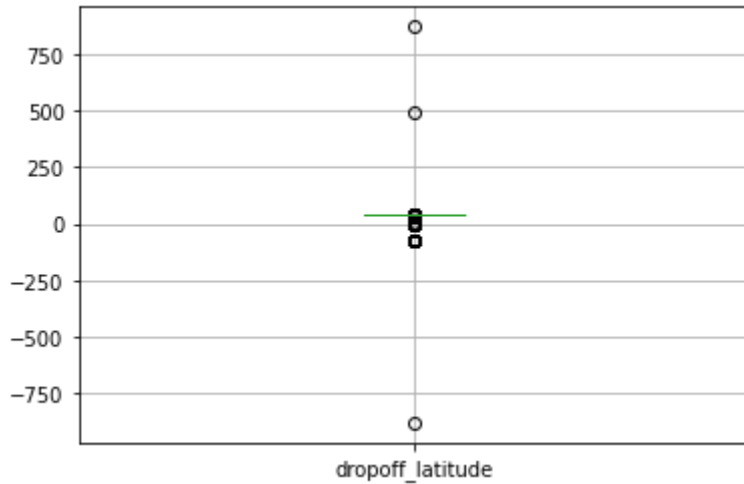
```
df.boxplot('dropoff_longitude')
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f96eb158fd0>
```



```
df.boxplot('dropoff_latitude')
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f96eaf8be90>
```



```
q1=df['dropoff_latitude'].quantile(0.25)
```

```
q3=df['dropoff_latitude'].quantile(0.75)
```

```
IQR=q3-q1
```

```
LL=q1-1.5*(IQR)
```

```
UL=q3+1.5*(IQR)
```

```
print(UL,LL)
```

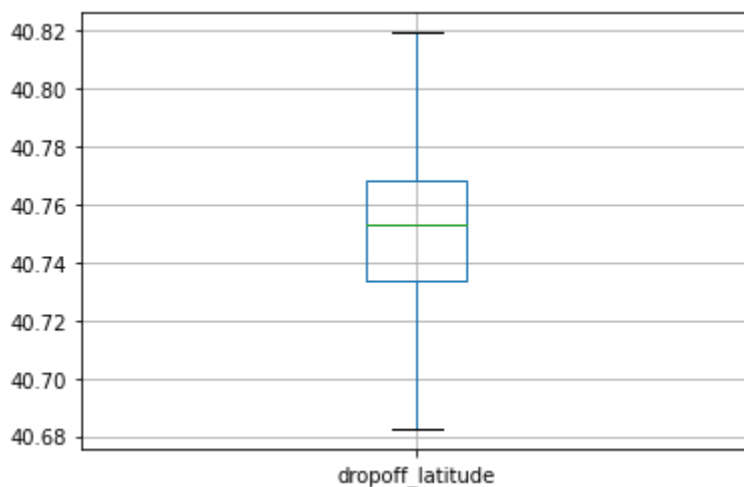
```
40.819268347747794 40.68255579135132
```

```
df['dropoff_latitude']=np.where(df['dropoff_latitude'] <=LL, LL,df['dropoff_latitude'])
```

```
df['dropoff_latitude']=np.where(df['dropoff_latitude'] >=UL, UL,df['dropoff_latitude'])
```

```
df.boxplot('dropoff_latitude')
```

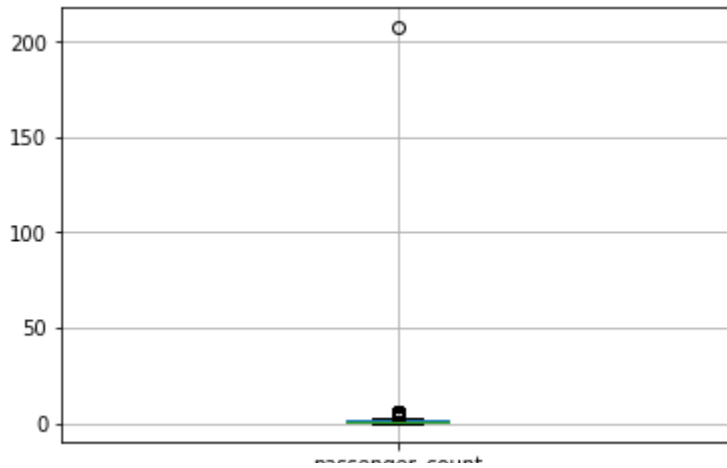
```
<matplotlib.axes._subplots.AxesSubplot at 0x7f96eaf830d0>
```



```
df.boxplot(column='passenger_count')
```



<matplotlib.axes.\_subplots.AxesSubplot at 0x7f96ed1adf10>



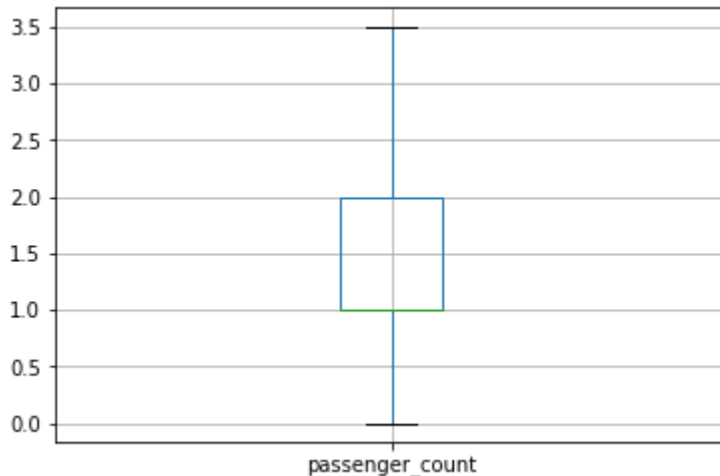
```
q1=df['passenger_count'].quantile(0.25)
q3=df['passenger_count'].quantile(0.75)
IQR=q3-q1
LL=q1-1.5*(IQR)
UL=q3+1.5*(IQR)
print(UL,LL)
```

3.5 -0.5

```
df['passenger_count']=np.where(df['passenger_count'] <=LL, LL,df['passenger_count'])
df['passenger_count']=np.where(df['passenger_count'] >=UL, UL,df['passenger_count'])
```

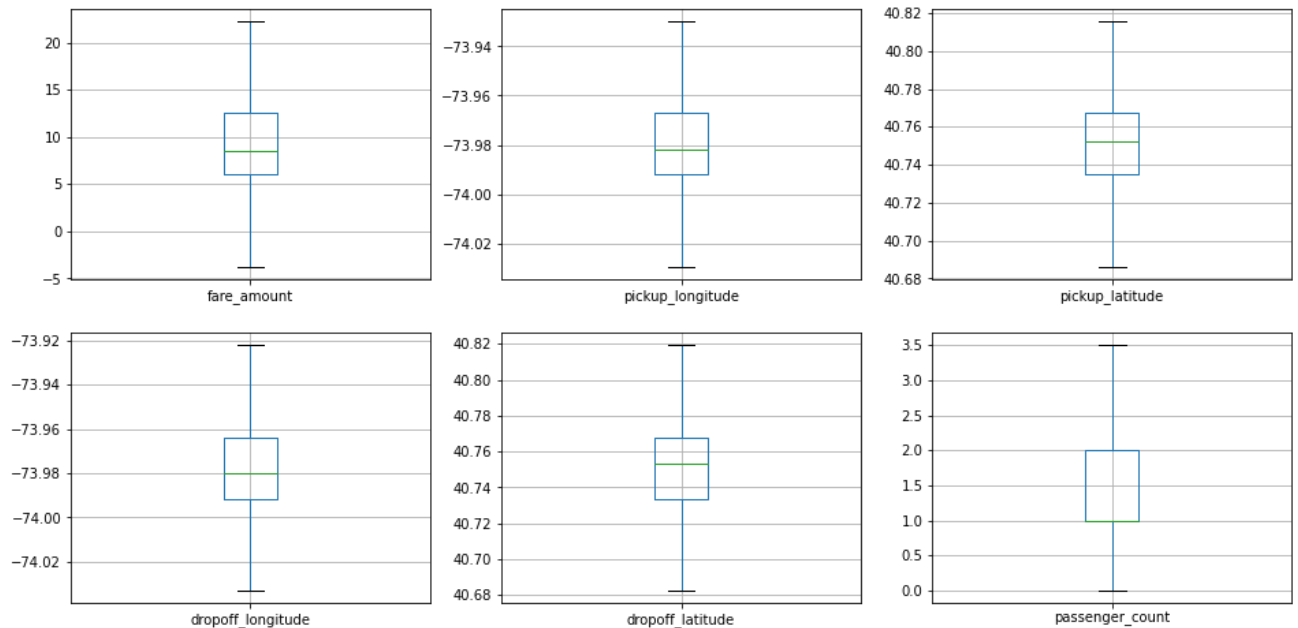
```
df.boxplot(column='passenger_count')
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f96eaf83c10>

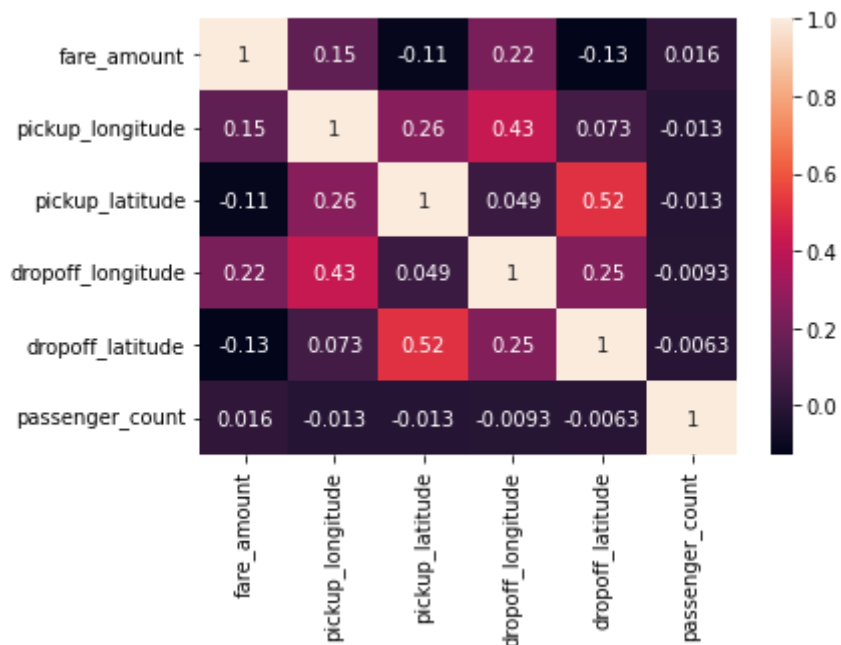


```
fig, axes = plt.subplots(2, 3, figsize=(16, 8))
df.boxplot(column="fare_amount",ax=axes[0,0])
df.boxplot(column="pickup_longitude",ax=axes[0,1])
df.boxplot(column="pickup_latitude",ax=axes[0,2])
df.boxplot(column="dropoff_longitude",ax=axes[1,0])
df.boxplot(column="dropoff_latitude",ax=axes[1,1])
df.boxplot(column="passenger_count",ax=axes[1,2])
```

&lt;matplotlib.axes.\_subplots.AxesSubplot at 0x7f96eacca1d0&gt;



```
sns.heatmap(df.corr(),annot=True)
plt.show()
```



```
x= df.iloc[:, :-1]
y= df.iloc[:, -1:]
```

```
from sklearn.model_selection import train_test_split
```

```
xtrain, xtest, ytrain, ytest = train_test_split(df.drop(labels=['fare_amount','pickup_date
```

```
print("xtrain shape : ", xtrain.shape)
print("xtest shape : ", xtest.shape)
print("ytrain shape : ", ytrain.shape)
print("ytest shape : ", ytest.shape)
```

```
xtrain shape : (150000, 5)
xtest shape : (50000, 5)
ytrain shape : (150000,)
ytest shape : (50000,)
```

```
from sklearn.linear_model import LinearRegression
lm=LinearRegression()
lm.fit(xtrain,ytrain)
```

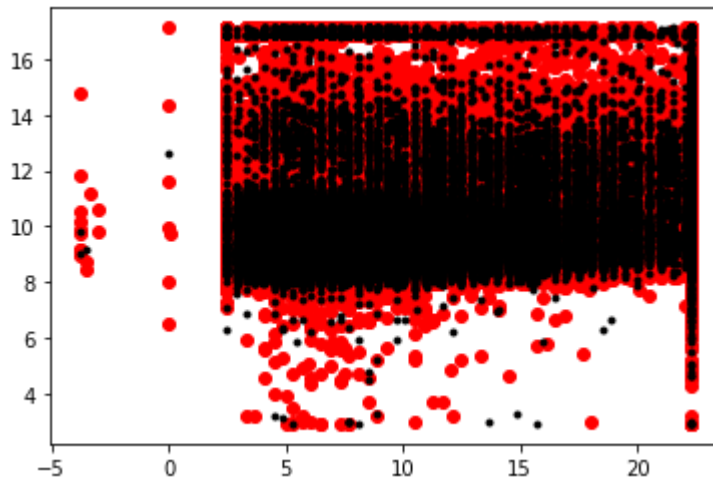
```
LinearRegression()
```

```
ytest_pred=lm.predict(xtest)
ytrain_pred=lm.predict(xtrain)
```

```
import matplotlib.pyplot as plt
```

```
plt.scatter(ytrain,ytrain_pred,c='red',marker='o',label="Training data")
plt.scatter(ytest,ytest_pred,c="black",marker=".",label="Testing data")
```

```
<matplotlib.collections.PathCollection at 0x7f96e6bebbd0>
```



```
from sklearn.metrics import mean_squared_error,r2_score
mse = mean_squared_error(ytest,ytest_pred)
print("mseTest = ", mse)
print("rmseTest = ", np.sqrt(mse))
mse = mean_squared_error(ytrain,ytrain_pred)
print("mseTrain = ", mse)
print("rmseTrain = ", np.sqrt(mse))
```

```
mseTest = 27.04690874482973
rmseTest = 5.200664259960426
mseTrain = 26.936549296136462
rmseTrain = 5.19004328461107
```

```
corr_matrix = df.corr()
```

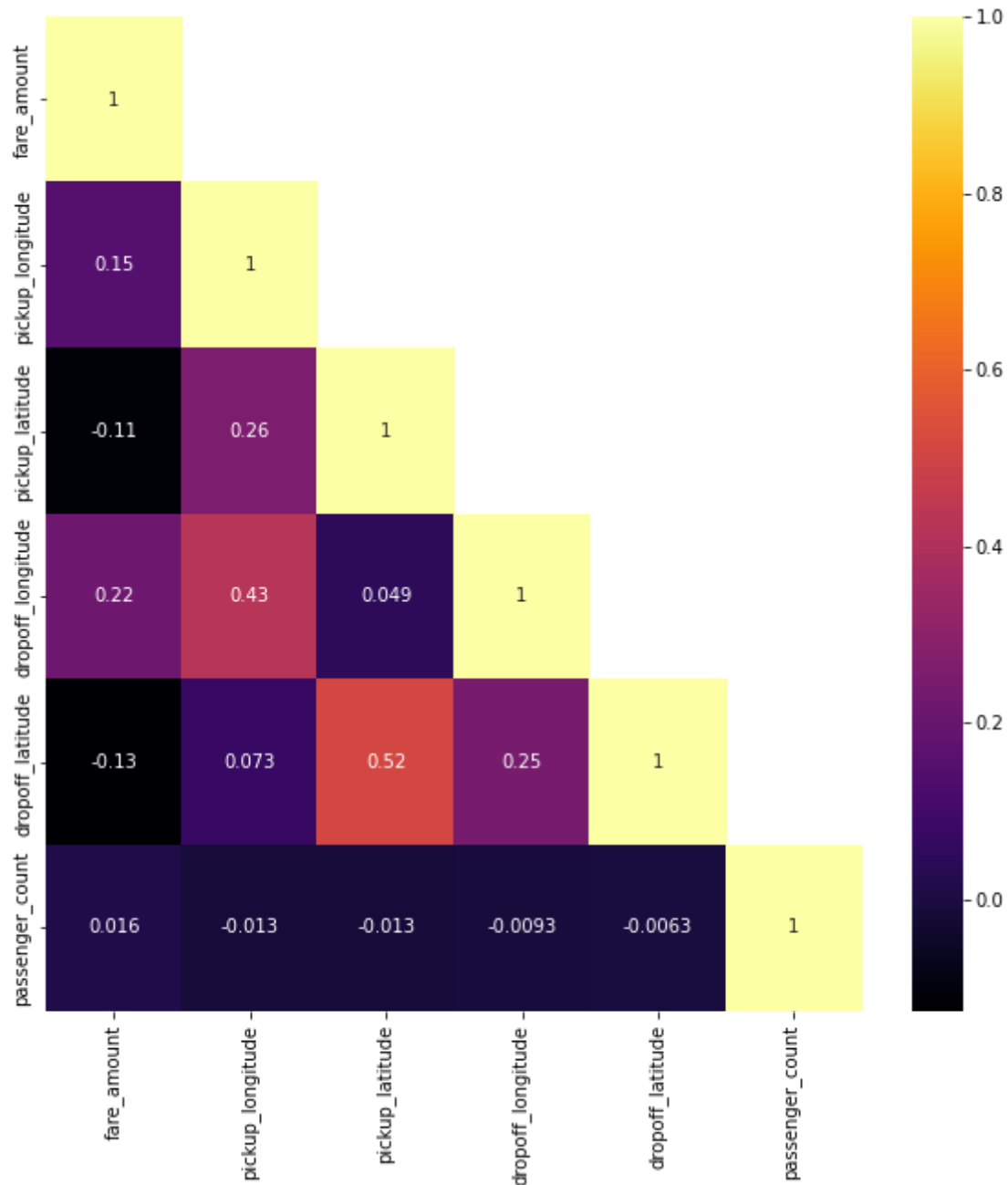
```
print(corr_matrix["fare_amount"].sort_values(ascending=False))
```

```
fare_amount      1.000000
dropoff_longitude 0.218704
pickup_longitude  0.154069
passenger_count   0.015778
pickup_latitude  -0.110842
dropoff_latitude  -0.125898
Name: fare_amount, dtype: float64
```

```
plt.figure(figsize=(10,10))
```

```
sns.heatmap(df.corr(), annot=True, cmap='inferno', mask=np.triu(df.corr(), k=1))
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f96e6ba73d0>
```



```
from sklearn.ensemble import RandomForestRegressor
```

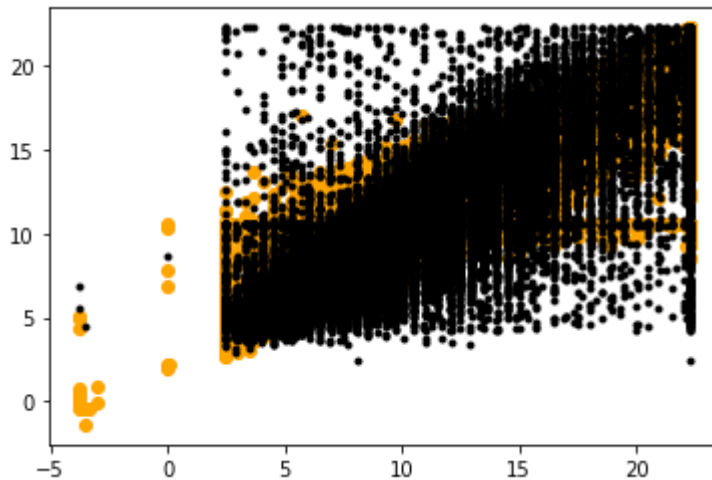
```
rf = RandomForestRegressor()  
rf.fit(xtrain,ytrain)
```

```
RandomForestRegressor()
```

```
ytest_pred2=rf.predict(xtest)  
ytrain_pred2=rf.predict(xtrain)
```

```
plt.scatter(ytrain,ytrain_pred2,c='orange',marker='o',label="Training data")  
plt.scatter(ytest,ytest_pred2,c="black",marker=".",label="Testing data")
```

<matplotlib.collections.PathCollection at 0x7f96e6a9b090>



```
from sklearn.metrics import mean_squared_error,r2_score, mean_absolute_error  
mse = mean_squared_error(ytest,ytest_pred2)  
print("mseTest = ", mse)  
print("rmseTest = ", np.sqrt(mse))
```

```
mse = mean_squared_error(ytrain,ytrain_pred2)  
print("mseTrain = ", mse)  
print("rmseTrain = ", np.sqrt(mse))
```

```
mseTest = 7.109194055504839  
rmseTest = 2.6663071945116976  
mseTrain = 1.6670899194229358  
rmseTrain = 1.291158363417492
```

[Colab paid products](#) - [Cancel contracts here](#)

