

Delhivery_BCS

September 19, 2024

1 About Delhivery

Delhivery is the largest and fastest-growing fully integrated player in India by revenue in Fiscal 2021. They aim to build the operating system for commerce, through a combination of world-class infrastructure, logistics operations of the highest quality, and cutting-edge engineering and technology capabilities.

The Data team builds intelligence and capabilities using this data that helps them to widen the gap between the quality, efficiency, and profitability of their business versus their competitors.

2 Problem Statement

We have been given data on trips performed by parcels for Delhivery, which has attributes like trip_creation_time, routes, source and destination places, and open-source routing engine time. We need to clean, sanitize and manipulate data and get useful features and provide data to help them build forecasting models.

Importing required Python Libraries

```
[3]: import numpy as np
import pandas as pd
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels
from scipy.special import comb
from scipy.stats import binom
from scipy.stats import norm,t
from scipy.stats import poisson, expon,geom, ttest_1samp,
↳ttest_ind,ttest_ind_from_stats,boxcox
from scipy.stats import shapiro, levene, kruskal, chi2,
↳chi2_contingency,pearsonr, spearmanr
from statsmodels.graphics.gofplots import qqplot
from sklearn.preprocessing import LabelEncoder, StandardScaler, MinMaxScaler,
↳OneHotEncoder
from warnings import filterwarnings
filterwarnings('ignore')
```

```
[4]: df = pd.read_csv('delhivery_data.csv')
```

Observations

on shape of data, data types of all the attributes, conversion of categorical attributes to 'category', missing value detection, statistical summary

```
[8]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 144867 entries, 0 to 144866
Data columns (total 24 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   data                                  144867 non-null  object
1   trip_creation_time                   144867 non-null  object
2   route_schedule_uuid                 144867 non-null  object
3   route_type                           144867 non-null  object
4   trip_uuid                            144867 non-null  object
5   source_center                       144867 non-null  object
6   source_name                         144574 non-null  object
7   destination_center                  144867 non-null  object
8   destination_name                    144606 non-null  object
9   od_start_time                       144867 non-null  object
10  od_end_time                         144867 non-null  object
11  start_scan_to_end_scan               144867 non-null  float64
12  is_cutoff                           144867 non-null  bool
13  cutoff_factor                       144867 non-null  int64
14  cutoff_timestamp                    144867 non-null  object
15  actual_distance_to_destination       144867 non-null  float64
16  actual_time                         144867 non-null  float64
17  osrm_time                           144867 non-null  float64
18  osrm_distance                       144867 non-null  float64
19  factor                              144867 non-null  float64
20  segment_actual_time                 144867 non-null  float64
21  segment_osrm_time                   144867 non-null  float64
22  segment_osrm_distance               144867 non-null  float64
23  segment_factor                      144867 non-null  float64
dtypes: bool(1), float64(10), int64(1), object(12)
memory usage: 25.6+ MB
```

```
[9]: df.shape
```

```
[9]: (144867, 24)
```

```
[10]: df.head()
```

```

[10]:      data      trip_creation_time  \
0  training  2018-09-20 02:35:36.476840
1  training  2018-09-20 02:35:36.476840
2  training  2018-09-20 02:35:36.476840
3  training  2018-09-20 02:35:36.476840
4  training  2018-09-20 02:35:36.476840

      route_schedule_uuid route_type  \
0  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...  Carting
1  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...  Carting
2  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...  Carting
3  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...  Carting
4  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...  Carting

      trip_uuid source_center      source_name  \
0  trip-153741093647649320  IND388121AAA  Anand_VUNagar_DC (Gujarat)
1  trip-153741093647649320  IND388121AAA  Anand_VUNagar_DC (Gujarat)
2  trip-153741093647649320  IND388121AAA  Anand_VUNagar_DC (Gujarat)
3  trip-153741093647649320  IND388121AAA  Anand_VUNagar_DC (Gujarat)
4  trip-153741093647649320  IND388121AAA  Anand_VUNagar_DC (Gujarat)

      destination_center      destination_name  \
0      IND388620AAB  Khambhat_MotvdDPP_D (Gujarat)
1      IND388620AAB  Khambhat_MotvdDPP_D (Gujarat)
2      IND388620AAB  Khambhat_MotvdDPP_D (Gujarat)
3      IND388620AAB  Khambhat_MotvdDPP_D (Gujarat)
4      IND388620AAB  Khambhat_MotvdDPP_D (Gujarat)

      od_start_time  ...      cutoff_timestamp  \
0  2018-09-20 03:21:32.418600  ...      2018-09-20 04:27:55
1  2018-09-20 03:21:32.418600  ...      2018-09-20 04:17:55
2  2018-09-20 03:21:32.418600  ...  2018-09-20 04:01:19.505586
3  2018-09-20 03:21:32.418600  ...      2018-09-20 03:39:57
4  2018-09-20 03:21:32.418600  ...      2018-09-20 03:33:55

      actual_distance_to_destination  actual_time  osrm_time  osrm_distance  \
0      10.435660      14.0      11.0      11.9653
1      18.936842      24.0      20.0      21.7243
2      27.637279      40.0      28.0      32.5395
3      36.118028      62.0      40.0      45.5620
4      39.386040      68.0      44.0      54.2181

      factor  segment_actual_time  segment_osrm_time  segment_osrm_distance  \
0  1.272727      14.0      11.0      11.9653
1  1.200000      10.0      9.0      9.7590
2  1.428571      16.0      7.0      10.8152
3  1.550000      21.0      12.0      13.0224

```

```
4  1.545455          6.0          5.0          3.9153
```

```
    segment_factor
0      1.272727
1      1.111111
2      2.285714
3      1.750000
4      1.200000
```

```
[5 rows x 24 columns]
```

```
[11]: df.nunique() # number of unique values in columns
```

```
[11]: data                2
trip_creation_time      14817
route_schedule_uuid     1504
route_type              2
trip_uuid              14817
source_center           1508
source_name             1498
destination_center      1481
destination_name        1468
od_start_time          26369
od_end_time            26369
start_scan_to_end_scan  1915
is_cutoff               2
cutoff_factor           501
cutoff_timestamp        93180
actual_distance_to_destination 144515
actual_time             3182
osrm_time              1531
osrm_distance          138046
factor                 45641
segment_actual_time      747
segment_osrm_time        214
segment_osrm_distance    113799
segment_factor           5675
dtype: int64
```

```
[12]: df.isna().sum() #missing values in columns
```

```
[12]: data                0
trip_creation_time      0
route_schedule_uuid     0
route_type              0
trip_uuid              0
source_center           0
```

```

source_name                293
destination_center          0
destination_name            261
od_start_time               0
od_end_time                 0
start_scan_to_end_scan      0
is_cutoff                   0
cutoff_factor               0
cutoff_timestamp            0
actual_distance_to_destination 0
actual_time                 0
osrm_time                   0
osrm_distance               0
factor                      0
segment_actual_time         0
segment_osrm_time           0
segment_osrm_distance        0
segment_factor              0
dtype: int64

```

```
[13]: df.describe() #Statistical summary of the dataset
```

```
[13]:
```

	start_scan_to_end_scan	cutoff_factor	actual_distance_to_destination \
count	144867.000000	144867.000000	144867.000000
mean	961.262986	232.926567	234.073372
std	1037.012769	344.755577	344.990009
min	20.000000	9.000000	9.000045
25%	161.000000	22.000000	23.355874
50%	449.000000	66.000000	66.126571
75%	1634.000000	286.000000	286.708875
max	7898.000000	1927.000000	1927.447705

	actual_time	osrm_time	osrm_distance	factor \
count	144867.000000	144867.000000	144867.000000	144867.000000
mean	416.927527	213.868272	284.771297	2.120107
std	598.103621	308.011085	421.119294	1.715421
min	9.000000	6.000000	9.008200	0.144000
25%	51.000000	27.000000	29.914700	1.604264
50%	132.000000	64.000000	78.525800	1.857143
75%	513.000000	257.000000	343.193250	2.213483
max	4532.000000	1686.000000	2326.199100	77.387097

	segment_actual_time	segment_osrm_time	segment_osrm_distance \
count	144867.000000	144867.000000	144867.000000
mean	36.196111	18.507548	22.82902
std	53.571158	14.775960	17.86066
min	-244.000000	0.000000	0.00000

25%	20.000000	11.000000	12.07010
50%	29.000000	17.000000	23.51300
75%	40.000000	22.000000	27.81325
max	3051.000000	1611.000000	2191.40370

	segment_factor
count	144867.000000
mean	2.218368
std	4.847530
min	-23.444444
25%	1.347826
50%	1.684211
75%	2.250000
max	574.250000

```
[14]: df.describe(include=object)
```

```
[14]:
```

	data	trip_creation_time	\
count	144867	144867	
unique	2	14817	
top	training	2018-09-28 05:23:15.359220	
freq	104858	101	

	route_schedule_uuid	route_type	\
count	144867	144867	
unique	1504	2	
top	thanos::sroute:4029a8a2-6c74-4b7e-a6d8-f9e069f...	FTL	
freq	1812	99660	

	trip_uuid	source_center	source_name	\
count	144867	144867	144574	
unique	14817	1508	1498	
top	trip-153811219535896559	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	
freq	101	23347	23347	

	destination_center	destination_name	\
count	144867	144606	
unique	1481	1468	
top	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	
freq	15192	15192	

	od_start_time	od_end_time	\
count	144867	144867	
unique	26369	26369	
top	2018-09-21 18:37:09.322207	2018-09-24 09:59:15.691618	
freq	81	81	

```

        cutoff_timestamp
count          144867
unique          93180
top    2018-09-24 05:19:20
freq              40

```

```

[15]: #Checking for source center value for which source name is null
df[(df["source_name"].notnull()) & (df["source_center"].
    ↳isin(df[df["source_name"].isnull()]["source_center"]))]

```

```

[15]: Empty DataFrame
Columns: [data, trip_creation_time, route_schedule_uuid, route_type, trip_uuid,
source_center, source_name, destination_center, destination_name, od_start_time,
od_end_time, start_scan_to_end_scan, is_cutoff, cutoff_factor, cutoff_timestamp,
actual_distance_to_destination, actual_time, osrm_time, osrm_distance, factor,
segment_actual_time, segment_osrm_time, segment_osrm_distance, segment_factor]
Index: []

[0 rows x 24 columns]

```

```

[16]: #Checking for destination center value for which destination name is null
df[(df["destination_name"].notnull()) & (df["destination_center"].
    ↳isin(df[df["destination_name"].isnull()]["destination_center"]))]

```

```

[16]: Empty DataFrame
Columns: [data, trip_creation_time, route_schedule_uuid, route_type, trip_uuid,
source_center, source_name, destination_center, destination_name, od_start_time,
od_end_time, start_scan_to_end_scan, is_cutoff, cutoff_factor, cutoff_timestamp,
actual_distance_to_destination, actual_time, osrm_time, osrm_distance, factor,
segment_actual_time, segment_osrm_time, segment_osrm_distance, segment_factor]
Index: []

[0 rows x 24 columns]

```

```

[17]: #Here we can observe that minimum value of segment_actual_time and
    ↳segment_factor is negative,
    #which seems false values as time can not be negative, so we will drop that data
df.drop(df[df["segment_actual_time"]<0].index, inplace=True)

```

```

[18]: df.describe()

```

```

[18]:
      start_scan_to_end_scan  cutoff_factor  actual_distance_to_destination \
count          144846.000000    144846.000000          144846.000000
mean             961.226537      232.911057            234.057171
std            1036.993595      344.740981            344.974984
min              20.000000        9.000000             9.000045
25%            161.000000      22.000000            23.354927

```

50%	449.000000	66.000000	66.126234
75%	1634.000000	286.000000	286.706673
max	7898.000000	1927.000000	1927.447705

	actual_time	osrm_time	osrm_distance	factor \
count	144846.000000	144846.000000	144846.000000	144846.000000
mean	416.908724	213.853002	284.750969	2.120190
std	598.085058	307.997702	421.101831	1.715508
min	9.000000	6.000000	9.008200	0.144000
25%	51.000000	27.000000	29.909925	1.604288
50%	132.000000	64.000000	78.524600	1.857143
75%	513.000000	257.000000	343.062075	2.213589
max	4532.000000	1686.000000	2326.199100	77.387097

	segment_actual_time	segment_osrm_time	segment_osrm_distance \
count	144846.000000	144846.000000	144846.000000
mean	36.207427	18.507304	22.828528
std	53.561259	14.775870	17.860268
min	0.000000	0.000000	0.000000
25%	20.000000	11.000000	12.070100
50%	29.000000	17.000000	23.513000
75%	40.000000	22.000000	27.812975
max	3051.000000	1611.000000	2191.403700

	segment_factor
count	144846.000000
mean	2.219084
std	4.847144
min	-1.000000
25%	1.347826
50%	1.684211
75%	2.250000
max	574.250000

Dataset Information data: It contains whether the data is testing or training type

trip_creation_time: It is the timestamp of trip_creation. It ranges from '2018-09-12 00:25:19.499696' to '2018-10-03 23:59:42.701692'

oute_schedule_uuid: it is unique_id for particular route schedule

route type: It contains whether the route is Full Truck Load or Carting type

trip_uuid: It is a unique id associated with a particular trip

source_center: It is the ID of the origin of the trip

source_name: Its the name of the origin of the trip

destination_center: It is the ID of the destination of the trip

destination_name: It is the name of the destination of the trip

od_start_time: It is the trip start time

od_end_time: It is the trip end time

Start_scan_to_end_scan: It gives the time taken to deliver from source to destination. It ranges from 20 to 7898.

is_cutoff: It is an unknown field, which is boolean

cutoff_factor: It is the rounded value of the actual_distance_to_destination, it ranges from 9 to 1927

cutoff_timestamp: It is an unknown field

actual_distance_to_destination: It is the distance between the source and destination warehouses, it ranges from 9.00 to 1927.44

actual_time: It contains the actual time taken to complete the delivery (cumulative), it ranges from 9 to 4532.

osrm_time: It is an open-source routing engine time calculator which computes the shortest path between points in a given map and gives the time (cumulative), it ranges from 6 to 1686

osrm_distance: It contains the distance to the destination based on osrm, it ranges from 9.00 to 2326.199

factor: It is a ratio of actual_time to osrm_time, it ranges from 0.144 to 77.38.

segment_actual_time: It is a segment time, a time taken by a subset of package delivery, It ranges from -244 to 3051

segment_osrm_time: It contains the osrm time taken by a subset of the package delivery. It ranges from 0 to 1611

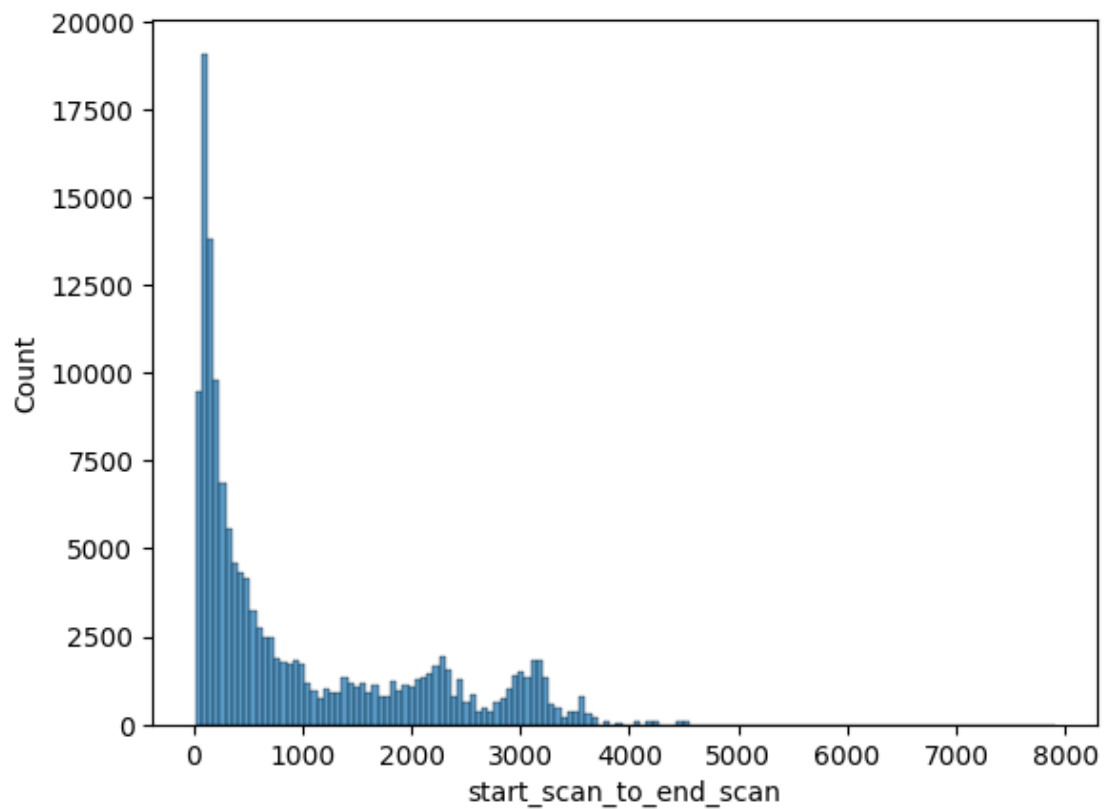
segment_osrm_distance: It contains OSRM distance, the distance covered by a subset of package delivery, it ranges from 0 to 2191.40

segment_factor: It is a ratio between segment_actual_time to segment_osrm_time, it ranges from -23.544 to 574.25

3 Univariate Analysis

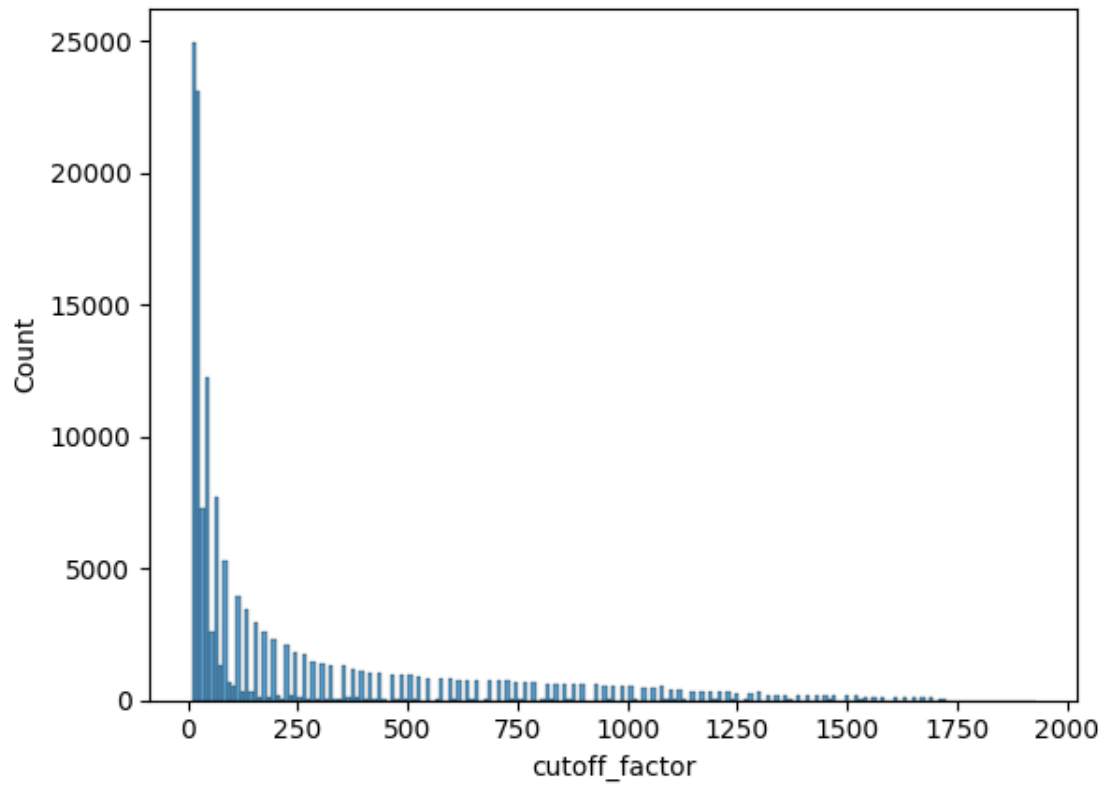
```
[19]: #Histogram for start_to_scan_to_end_scan attribute
sns.histplot(df["start_scan_to_end_scan"])
```

```
[19]: <Axes: xlabel='start_scan_to_end_scan', ylabel='Count'>
```



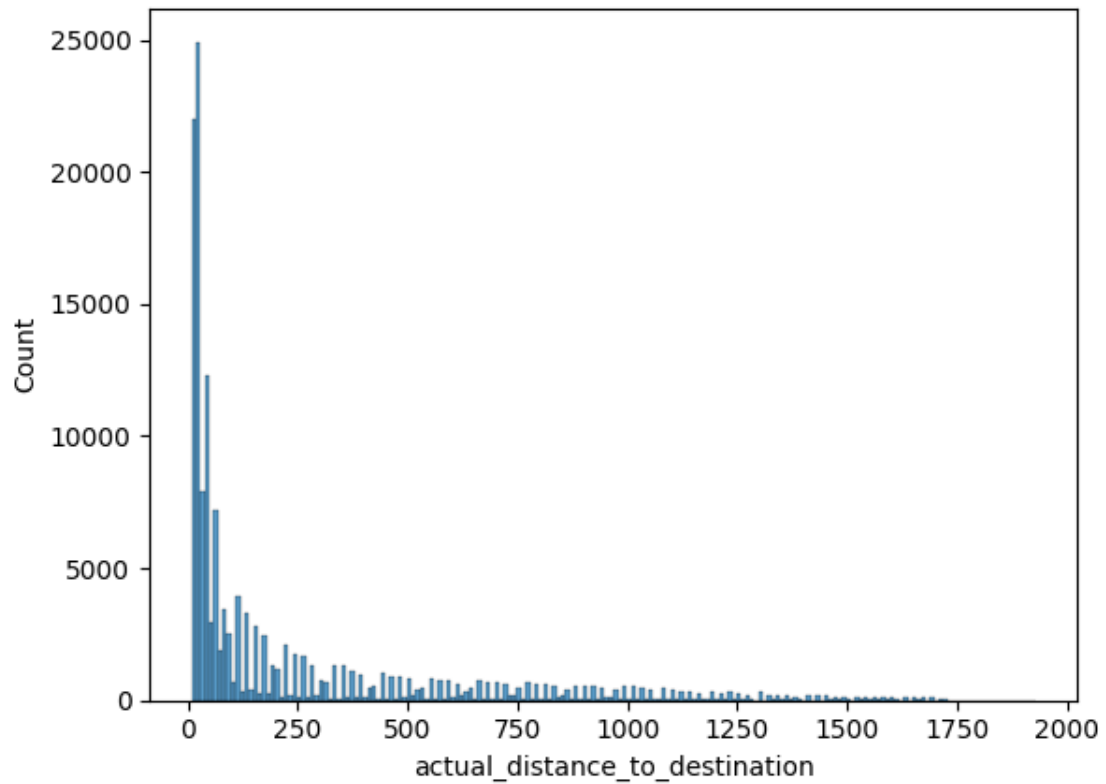
```
[20]: #Histogram for cutoff_factor attribute  
sns.histplot(df["cutoff_factor"])
```

```
[20]: <Axes: xlabel='cutoff_factor', ylabel='Count'>
```



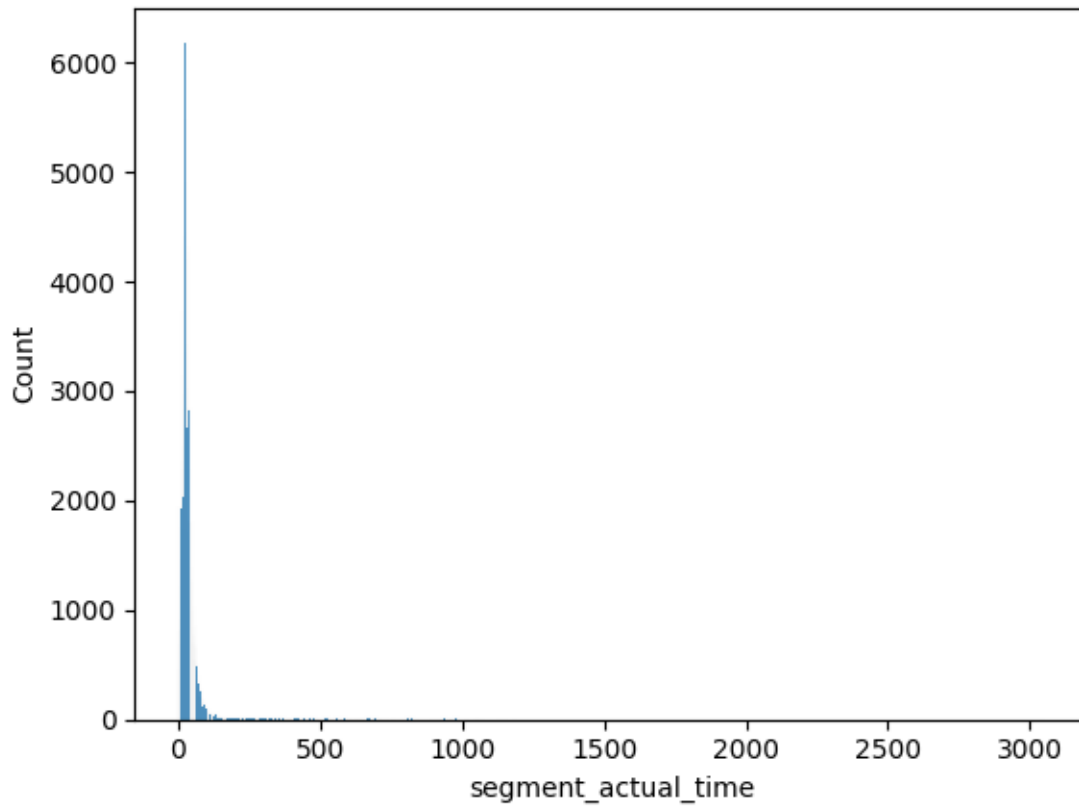
```
[21]: #Histogram for actual_distance_to_destination attribute  
sns.histplot(df["actual_distance_to_destination"])
```

```
[21]: <Axes: xlabel='actual_distance_to_destination', ylabel='Count'>
```



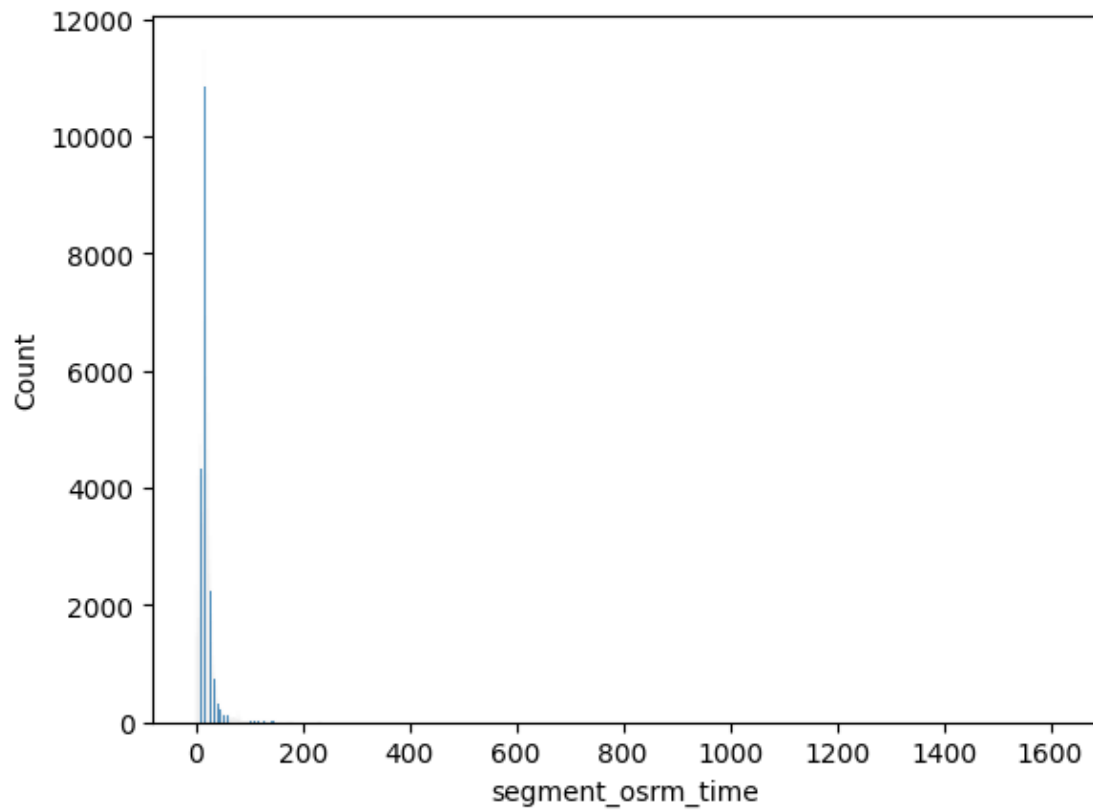
```
[22]: #Histogram for segment_actual_time attribute  
sns.histplot(df["segment_actual_time"])
```

```
[22]: <Axes: xlabel='segment_actual_time', ylabel='Count'>
```



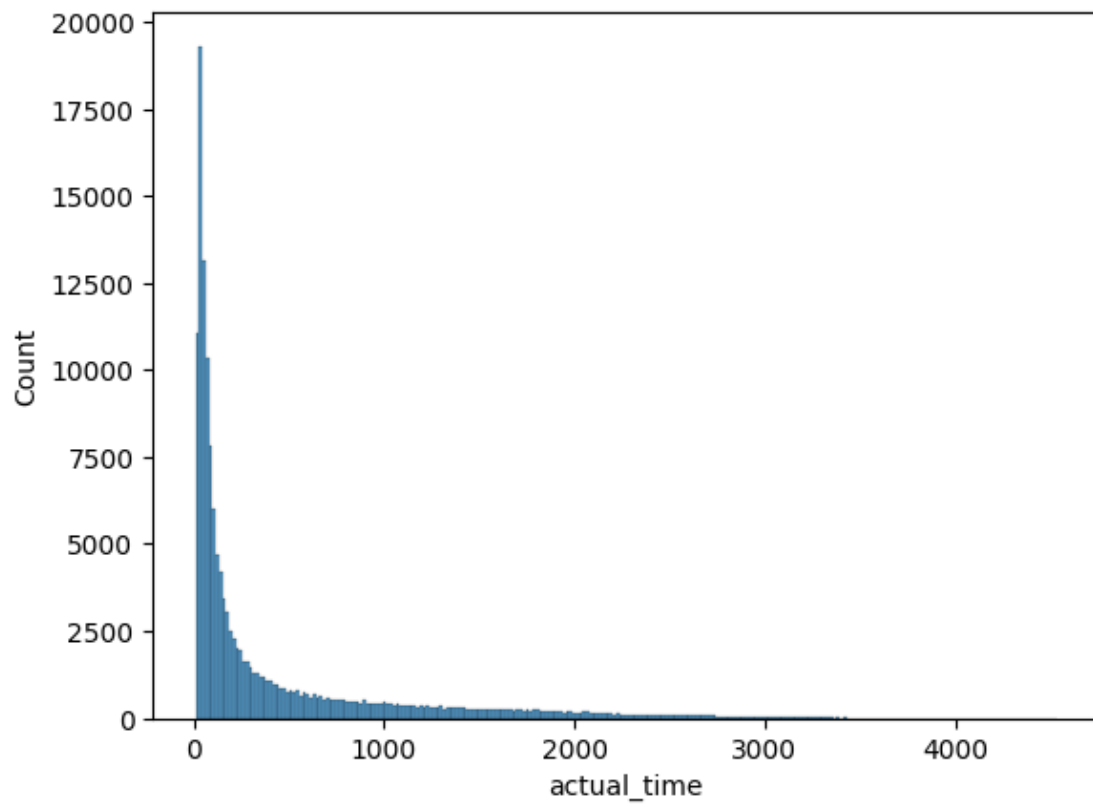
```
[23]: #Histogram for segment_osrm_time attribute  
sns.histplot(df["segment_osrm_time"])
```

```
[23]: <Axes: xlabel='segment_osrm_time', ylabel='Count'>
```



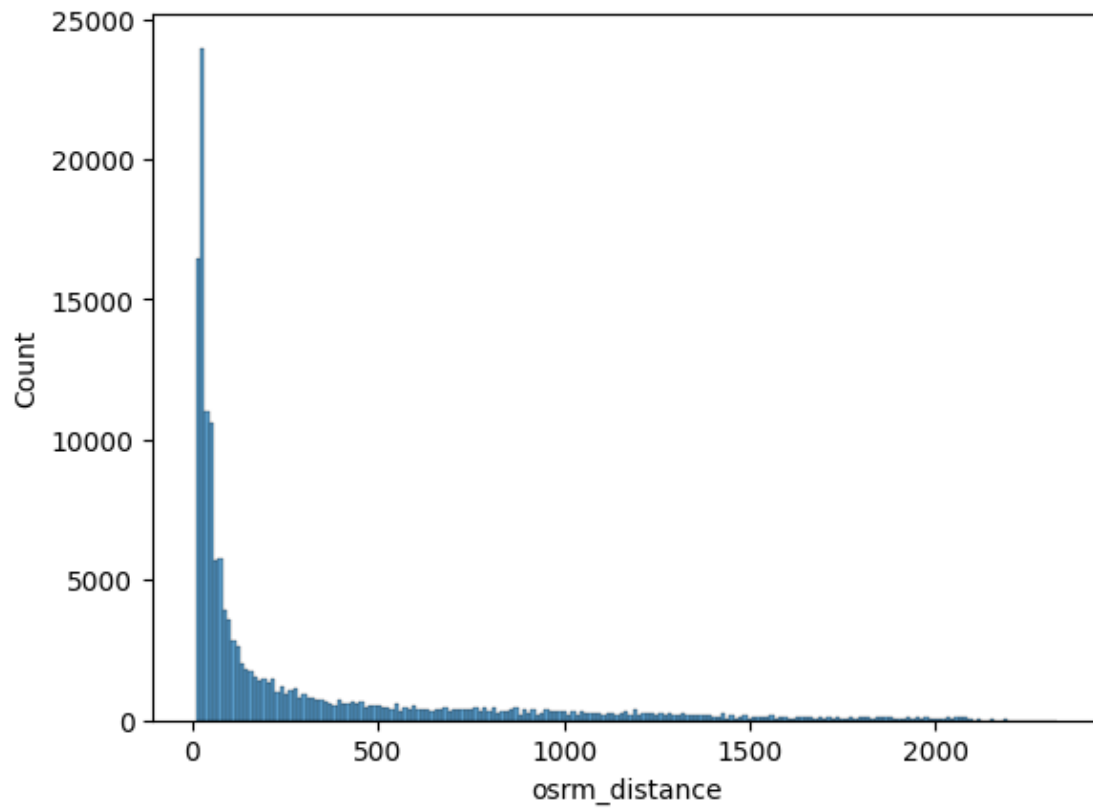
```
[24]: #Histogram for actual_time attribute  
sns.histplot(df["actual_time"])
```

```
[24]: <Axes: xlabel='actual_time', ylabel='Count'>
```



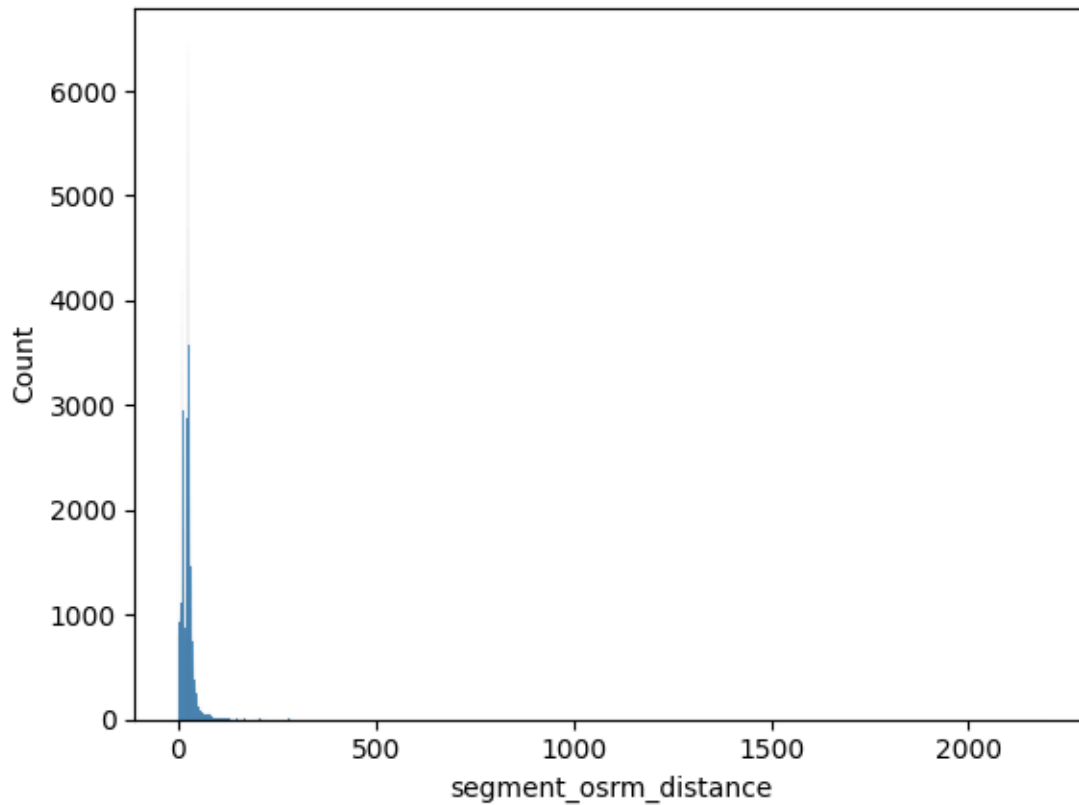
```
[25]: #Histogram for osrm_distance attribute  
sns.histplot(df["osrm_distance"])
```

```
[25]: <Axes: xlabel='osrm_distance', ylabel='Count'>
```



```
[26]: #Histogram for segment_osrm_distance attribute  
sns.histplot(df["segment_osrm_distance"])
```

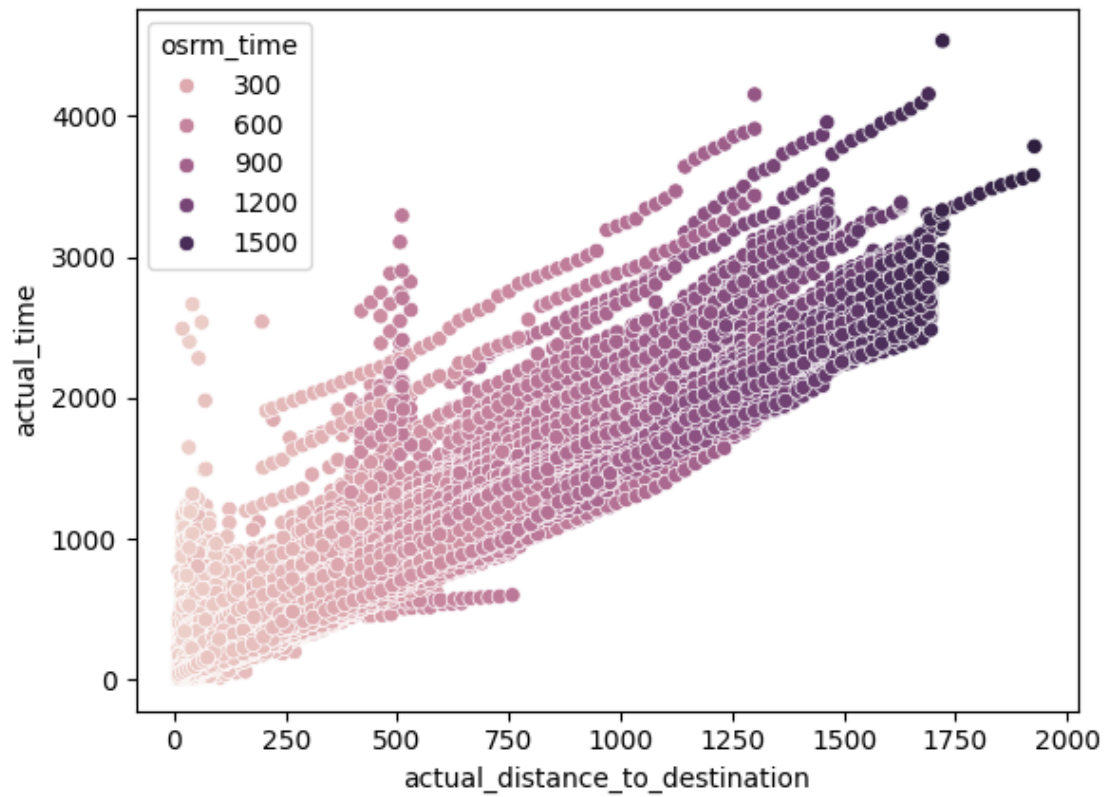
```
[26]: <Axes: xlabel='segment_osrm_distance', ylabel='Count'>
```

4 Bivariate Analysis

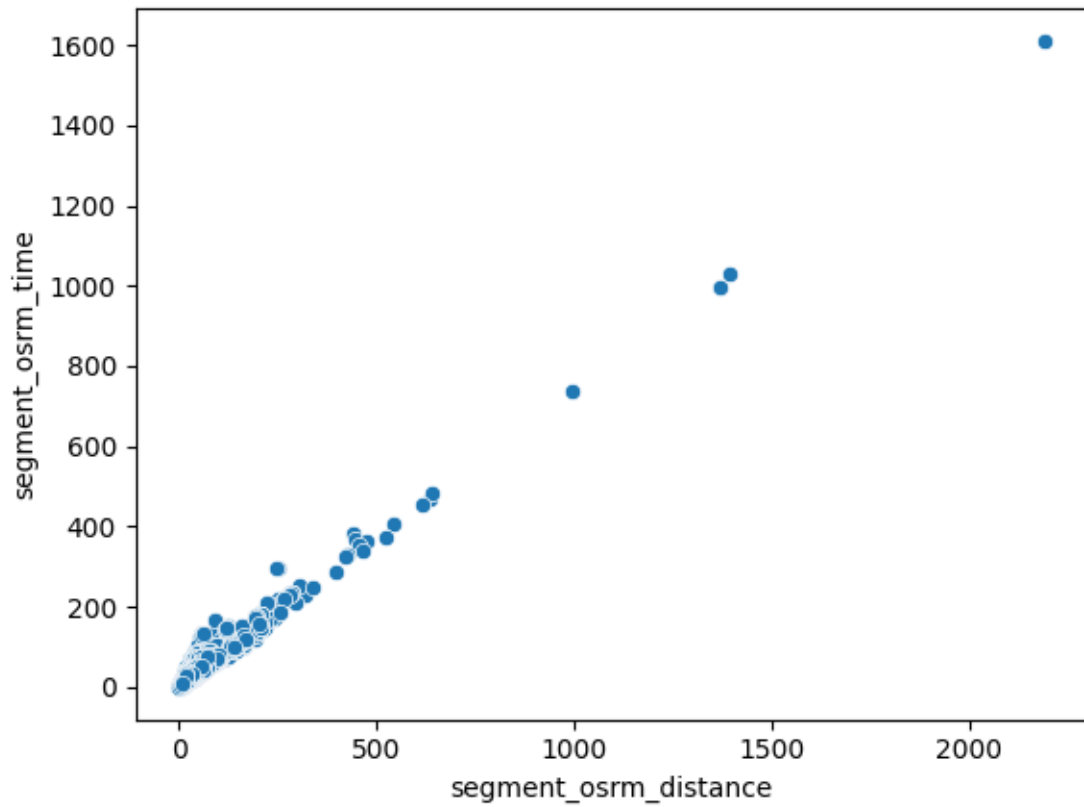
```
[27]: #Scatterplot between actual-distance_to_destination, actual_time and osrm_time  
sns.scatterplot(data=df, x="actual_distance_to_destination", y="actual_time",  
               ↪hue="osrm_time")
```

```
[27]: <Axes: xlabel='actual_distance_to_destination', ylabel='actual_time'>
```



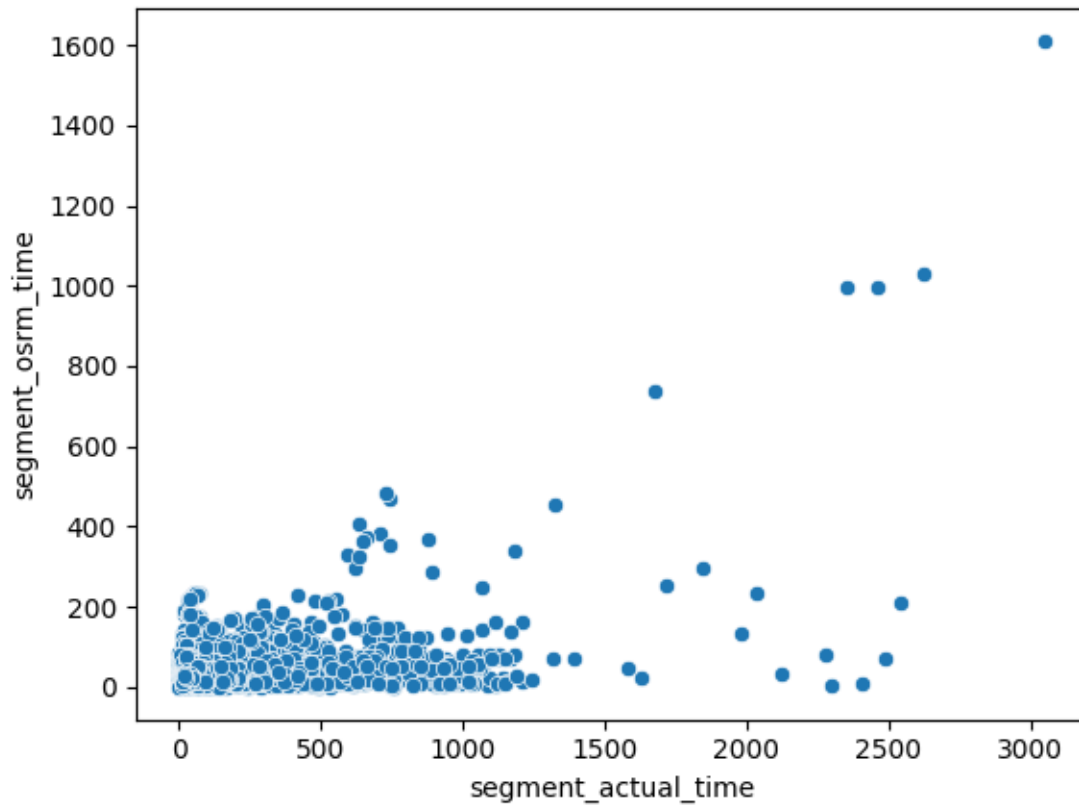
```
[28]: #scatterplot between segment_osrm_distance and segment_osrm_time
sns.scatterplot(data=df, x="segment_osrm_distance", y="segment_osrm_time")
```

```
[28]: <Axes: xlabel='segment_osrm_distance', ylabel='segment_osrm_time'>
```



```
[29]: #scatterplot between segment_actual_time and segment_osrm_time
sns.scatterplot(data=df, x="segment_actual_time",y="segment_osrm_time")
```

```
[29]: <Axes: xlabel='segment_actual_time', ylabel='segment_osrm_time'>
```



```
[35]: #Drop non-numeric columns
df_numeric = df.select_dtypes(include=['float64', 'int64'])
```

```
[34]: df_numeric.corr()
```

```
[34]:
```

	start_scan_to_end_scan	cutoff_factor \
start_scan_to_end_scan	1.000000	0.784656
cutoff_factor	0.784656	1.000000
actual_distance_to_destination	0.784988	0.999986
actual_time	0.785924	0.978719
osrm_time	0.785283	0.995833
osrm_distance	0.784120	0.997116
factor	-0.023192	-0.064559
segment_actual_time	0.093372	0.045063
segment_osrm_time	0.219844	0.157942
segment_osrm_distance	0.306972	0.231109
segment_factor	-0.020225	-0.031439

	actual_distance_to_destination	actual_time \
start_scan_to_end_scan	0.784988	0.785924
cutoff_factor	0.999986	0.978719

actual_distance_to_destination	1.000000	0.978658
actual_time	0.978658	1.000000
osrm_time	0.995872	0.977996
osrm_distance	0.997148	0.979398
factor	-0.064743	0.033498
segment_actual_time	0.045320	0.124483
segment_osrm_time	0.158836	0.171480
segment_osrm_distance	0.232119	0.242296
segment_factor	-0.031588	0.017570

	osrm_time	osrm_distance	factor	\
start_scan_to_end_scan	0.785283	0.784120	-0.023192	
cutoff_factor	0.995833	0.997116	-0.064559	
actual_distance_to_destination	0.995872	0.997148	-0.064743	
actual_time	0.977996	0.979398	0.033498	
osrm_time	1.000000	0.999119	-0.069081	
osrm_distance	0.999119	1.000000	-0.065391	
factor	-0.069081	-0.065391	1.000000	
segment_actual_time	0.049977	0.048787	0.518451	
segment_osrm_time	0.177074	0.169157	-0.053154	
segment_osrm_distance	0.242288	0.239672	-0.036724	
segment_factor	-0.033038	-0.031786	0.540448	

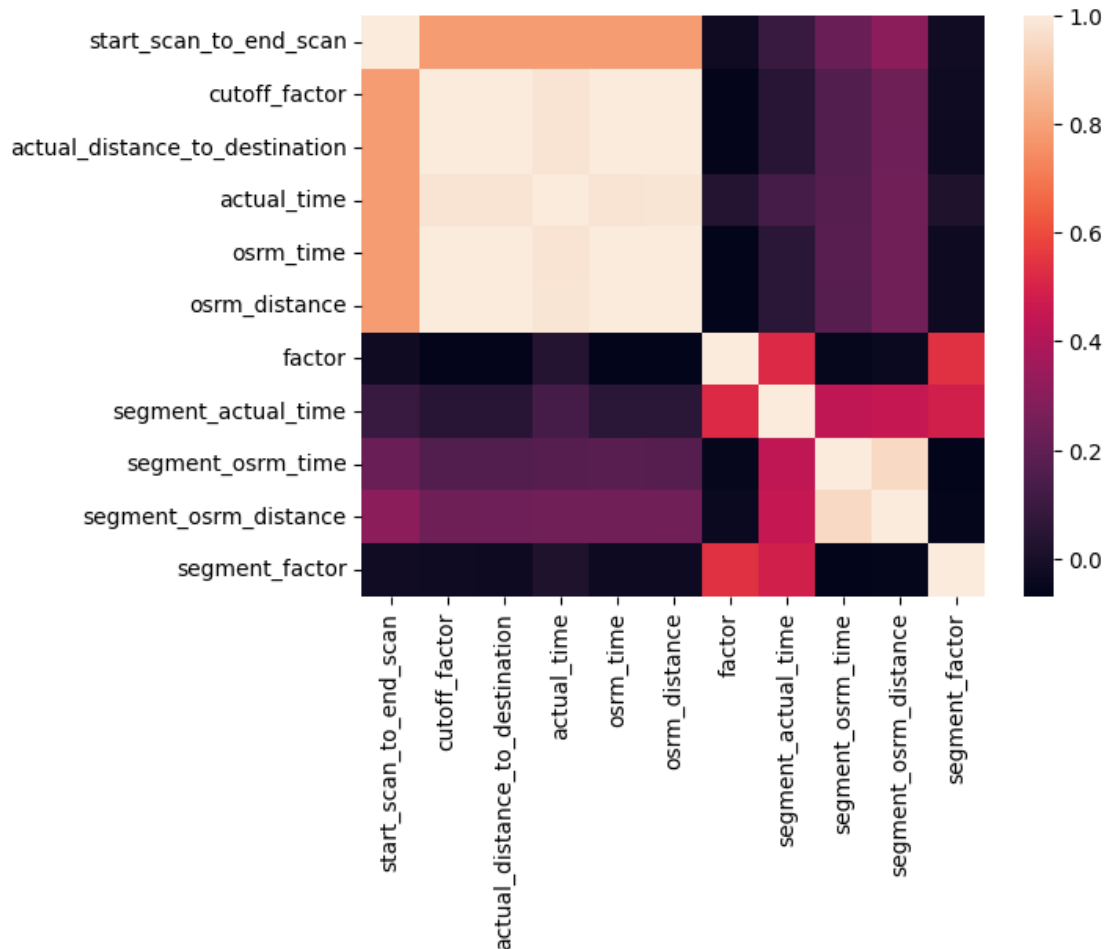
	segment_actual_time	segment_osrm_time	\
start_scan_to_end_scan	0.093372	0.219844	
cutoff_factor	0.045063	0.157942	
actual_distance_to_destination	0.045320	0.158836	
actual_time	0.124483	0.171480	
osrm_time	0.049977	0.177074	
osrm_distance	0.048787	0.169157	
factor	0.518451	-0.053154	
segment_actual_time	1.000000	0.433604	
segment_osrm_time	0.433604	1.000000	
segment_osrm_distance	0.449167	0.948520	
segment_factor	0.483699	-0.068472	

	segment_osrm_distance	segment_factor
start_scan_to_end_scan	0.306972	-0.020225
cutoff_factor	0.231109	-0.031439
actual_distance_to_destination	0.232119	-0.031588
actual_time	0.242296	0.017570
osrm_time	0.242288	-0.033038
osrm_distance	0.239672	-0.031786
factor	-0.036724	0.540448
segment_actual_time	0.449167	0.483699
segment_osrm_time	0.948520	-0.068472
segment_osrm_distance	1.000000	-0.059317

segment_factor -0.059317 1.000000

```
[36]: # Now plotting the heatmap
sns.heatmap(df_numeric.corr())
```

[36]: <Axes: >



5 Data Wrangling

```
[37]: #merging of rows based on trip_id and source and destination details
data=df.
    ↳groupby(["route_type","trip_uuid","trip_creation_time","source_center","source_name","desti
    ↳aggregate({"cutoff_factor":"max","actual_distance_to_destination":
    ↳"max","segment_actual_time":"sum","segment_osrm_time":"sum","actual_time":
    ↳"max","osrm_time":"max","osrm_distance":"max","segment_osrm_distance":
    ↳"sum"}).reset_index()
```

data

```
[37]:
```

	route_type	trip_uid	trip_creation_time	\
0	Carting	trip-153671042288605164	2018-09-12 00:00:22.886430	
1	Carting	trip-153671042288605164	2018-09-12 00:00:22.886430	
2	Carting	trip-153671046011330457	2018-09-12 00:01:00.113710	
3	Carting	trip-153671055416136166	2018-09-12 00:02:34.161600	
4	Carting	trip-153671055416136166	2018-09-12 00:02:34.161600	
...	
26218	FTL	trip-153861014185597051	2018-10-03 23:42:21.856227	
26219	FTL	trip-153861023893369544	2018-10-03 23:43:58.933947	
26220	FTL	trip-153861023893369544	2018-10-03 23:43:58.933947	
26221	FTL	trip-153861118270144424	2018-10-03 23:59:42.701692	
26222	FTL	trip-153861118270144424	2018-10-03 23:59:42.701692	

	source_center	source_name	destination_center	\
0	IND561203AAB	Doddablpur_ChikaDPP_D (Karnataka)	IND562101AAA	
1	IND572101AAA	Tumkur_Veersagr_I (Karnataka)	IND561203AAB	
2	IND400072AAB	Mumbai Hub (Maharashtra)	IND401104AAA	
3	IND600056AAA	Chennai_Poonamallee (Tamil Nadu)	IND602105AAB	
4	IND600116AAB	Chennai_Porur_DPC (Tamil Nadu)	IND600056AAA	
...	
26218	IND462022AAA	Bhopal_Trnsport_H (Madhya Pradesh)	IND209304AAA	
26219	IND382715AAA	Kadi_KaranNGR_D (Gujarat)	IND382430AAB	
26220	IND384205AAA	Mehsana_Panchot_IP (Gujarat)	IND382715AAA	
26221	IND583119AAA	Sandur_WrdN1DPP_D (Karnataka)	IND583101AAA	
26222	IND583201AAA	Hospet (Karnataka)	IND583119AAA	

	destination_name	od_start_time	\
0	Chikblapur_ShntiSgr_D (Karnataka)	2018-09-12 02:03:09.655591	
1	Doddablpur_ChikaDPP_D (Karnataka)	2018-09-12 00:00:22.886430	
2	Mumbai_MiraRd_IP (Maharashtra)	2018-09-12 00:01:00.113710	
3	Chennai_Sriperumbudur_Dc (Tamil Nadu)	2018-09-12 02:12:10.755603	
4	Chennai_Poonamallee (Tamil Nadu)	2018-09-12 00:02:34.161600	
...	
26218	Kanpur_Central_H_6 (Uttar Pradesh)	2018-10-03 23:42:21.856227	
26219	Ahmedabad_East_H_1 (Gujarat)	2018-10-04 01:48:54.382343	
26220	Kadi_KaranNGR_D (Gujarat)	2018-10-03 23:43:58.933947	
26221	Bellary_Dc (Karnataka)	2018-10-04 03:58:40.726547	
26222	Sandur_WrdN1DPP_D (Karnataka)	2018-10-04 02:51:44.712656	

	od_end_time	start_scan_to_end_scan	cutoff_factor	\
0	2018-09-12 03:01:59.598855	58.0	24	
1	2018-09-12 02:03:09.655591	122.0	48	
2	2018-09-12 01:41:29.809822	100.0	17	
3	2018-09-12 03:13:03.432532	60.0	15	
4	2018-09-12 02:12:10.755603	129.0	9	

```

...
26218 2018-10-04 19:57:34.928573 1215.0 442
26219 2018-10-04 04:01:41.425627 132.0 50
26220 2018-10-04 01:48:54.382343 124.0 34
26221 2018-10-04 08:46:09.166940 287.0 40
26222 2018-10-04 03:58:40.726547 66.0 25

actual_distance_to_destination segment_actual_time segment_osrm_time \
0 24.644021 46.0 26.0
1 48.542890 95.0 39.0
2 17.175274 59.0 16.0
3 15.325529 39.0 12.0
4 9.271519 21.0 11.0
...
26218 442.024575 991.0 425.0
26219 50.473578 129.0 55.0
26220 34.270235 57.0 37.0
26221 40.546740 233.0 42.0
26222 25.534793 41.0 25.0

actual_time osrm_time osrm_distance segment_osrm_distance
0 47.0 26.0 28.1994 28.1995
1 96.0 42.0 56.9116 55.9899
2 59.0 15.0 19.6800 19.8766
3 40.0 12.0 16.2225 16.2225
4 21.0 11.0 11.8422 11.8422
...
26218 997.0 395.0 545.1256 573.6479
26219 130.0 54.0 61.9571 67.2659
26220 57.0 38.0 40.4257 40.4256
26221 233.0 42.0 52.5303 52.5303
26222 42.0 26.0 28.0484 28.0484

```

[26223 rows x 18 columns]

```

[38]: #Merging rows based on trip_id
data=data.groupby(["route_type","trip_uuid","trip_creation_time"]).
    ↪aggregate({"source_center":"first","source_name":
    ↪"first","destination_center":"last",
    ↪"destination_name":"last", "od_start_time":
    ↪"first",
    ↪"od_end_time":"last","cutoff_factor":
    ↪"sum","actual_distance_to_destination":"sum","osrm_distance":"sum",
    ↪"start_scan_to_end_scan":"sum",
    ↪"segment_actual_time":"sum",
    ↪"segment_osrm_time":"sum","actual_time":
    ↪"sum",

```



```

                                "osrm_time":"sum","segment_osrm_distance":
↪"sum"}).reset_index()
data

```

```

[38]:
route_type      trip_uuid      trip_creation_time \
0      Carting trip-153671042288605164  2018-09-12 00:00:22.886430
1      Carting trip-153671046011330457  2018-09-12 00:01:00.113710
2      Carting trip-153671055416136166  2018-09-12 00:02:34.161600
3      Carting trip-153671066201138152  2018-09-12 00:04:22.011653
4      Carting trip-153671066826362165  2018-09-12 00:04:28.263977
...
14782      FTL trip-153861004148234782  2018-10-03 23:40:41.482736
14783      FTL trip-153861007249500192  2018-10-03 23:41:12.495257
14784      FTL trip-153861014185597051  2018-10-03 23:42:21.856227
14785      FTL trip-153861023893369544  2018-10-03 23:43:58.933947
14786      FTL trip-153861118270144424  2018-10-03 23:59:42.701692

source_center      source_name destination_center \
0      IND561203AAB  Doddablpur_ChikaDPP_D (Karnataka)  IND561203AAB
1      IND400072AAB      Mumbai Hub (Maharashtra)  IND401104AAA
2      IND600056AAA  Chennai_Poonamallee (Tamil Nadu)  IND600056AAA
3      IND600044AAD  Chennai_Chrompet_DPC (Tamil Nadu)  IND600048AAA
4      IND560043AAC      HBR Layout PC (Karnataka)  IND560043AAC
...
14782      IND814101AAB      Dumka_Dudhani_D (Jharkhand)  IND815351AAA
14783      IND842001AAA      Muzaffrpur_Bbganj_I (Bihar)  IND842001AAA
14784      IND206001AAA  Etawah_MhraChng_D (Uttar Pradesh)  IND209304AAA
14785      IND382715AAA      Kadi_KaranNGR_D (Gujarat)  IND382715AAA
14786      IND583119AAA      Sandur_WrdN1DPP_D (Karnataka)  IND583119AAA

destination_name      od_start_time \
0      Doddablpur_ChikaDPP_D (Karnataka)  2018-09-12 02:03:09.655591
1      Mumbai_MiraRd_IP (Maharashtra)  2018-09-12 00:01:00.113710
2      Chennai_Poonamallee (Tamil Nadu)  2018-09-12 02:12:10.755603
3      Chennai_Vandalur_Dc (Tamil Nadu)  2018-09-12 00:04:22.011653
4      HBR Layout PC (Karnataka)  2018-09-12 00:04:28.263977
...
14782      Jamtara_D (Jharkhand)  2018-10-04 04:22:21.025250
14783      Muzaffrpur_Bbganj_I (Bihar)  2018-10-03 23:41:12.495257
14784      Kanpur_Central_H_6 (Uttar Pradesh)  2018-10-05 02:44:50.858859
14785      Kadi_KaranNGR_D (Gujarat)  2018-10-04 01:48:54.382343
14786      Sandur_WrdN1DPP_D (Karnataka)  2018-10-04 03:58:40.726547

od_end_time      cutoff_factor \
0      2018-09-12 02:03:09.655591  72
1      2018-09-12 01:41:29.809822  17
2      2018-09-12 02:12:10.755603  24

```

3	2018-09-12 01:42:22.349694	9
4	2018-09-12 03:00:55.163423	22
...
14782	2018-10-04 02:24:41.382263	167
14783	2018-10-04 16:40:41.713085	192
14784	2018-10-04 19:57:34.928573	835
14785	2018-10-04 01:48:54.382343	84
14786	2018-10-04 03:58:40.726547	65

	actual_distance_to_destination	osrm_distance	start_scan_to_end_scan \
0	73.186911	85.1110	180.0
1	17.175274	19.6800	100.0
2	24.597048	28.0647	189.0
3	9.100510	12.0184	98.0
4	22.424210	28.9203	146.0
...
14782	168.396341	207.4975	428.0
14783	194.552260	229.2052	1017.0
14784	836.072017	997.7577	2180.0
14785	84.743813	102.3828	256.0
14786	66.081533	80.5787	353.0

	segment_actual_time	segment_osrm_time	actual_time	osrm_time \
0	141.0	65.0	143.0	68.0
1	59.0	16.0	59.0	15.0
2	60.0	23.0	61.0	23.0
3	24.0	13.0	24.0	13.0
4	64.0	34.0	64.0	34.0
...
14782	347.0	220.0	349.0	220.0
14783	845.0	178.0	847.0	178.0
14784	1660.0	891.0	1674.0	724.0
14785	186.0	92.0	187.0	92.0
14786	274.0	67.0	275.0	68.0

	segment_osrm_distance
0	84.1894
1	19.8766
2	28.0647
3	12.0184
4	28.9203
...	...
14782	209.4499
14783	232.5811
14784	1166.3614
14785	107.6915
14786	80.5787

[14787 rows x 18 columns]

```
[39]: data.nunique() # Unique values in the dataset
```

```
[39]: route_type                2
      trip_uuid                14787
      trip_creation_time       14787
      source_center            930
      source_name              930
      destination_center       1035
      destination_name         1035
      od_start_time            14787
      od_end_time              14787
      cutoff_factor            684
      actual_distance_to_destination 14771
      osrm_distance            14706
      start_scan_to_end_scan    2203
      segment_actual_time       1887
      segment_osrm_time         1242
      actual_time               1850
      osrm_time                 827
      segment_osrm_distance     14724
      dtype: int64
```

```
[40]: data.isna().sum() #nullvalues in the data frame
```

```
[40]: route_type                0
      trip_uuid                0
      trip_creation_time       0
      source_center            0
      source_name              0
      destination_center       0
      destination_name         0
      od_start_time            0
      od_end_time              0
      cutoff_factor            0
      actual_distance_to_destination 0
      osrm_distance            0
      start_scan_to_end_scan    0
      segment_actual_time       0
      segment_osrm_time         0
      actual_time               0
      osrm_time                 0
      segment_osrm_distance     0
      dtype: int64
```

```
[41]: data.describe() #statistical summary of dataset
```

```
[41]:
```

	cutoff_factor	actual_distance_to_destination	osrm_distance	\
count	14787.000000	14787.000000	14787.000000	
mean	163.379523	164.290730	204.631953	
std	305.558531	305.678137	370.953239	
min	9.000000	9.002461	9.072900	
25%	22.000000	22.840056	30.875600	
50%	48.000000	48.376934	65.575600	
75%	162.000000	163.685113	207.087600	
max	2185.000000	2187.483994	2840.081000	

	start_scan_to_end_scan	segment_actual_time	segment_osrm_time	\
count	14787.000000	14787.000000	14787.000000	
mean	529.442754	353.118618	180.482924	
std	658.286556	556.439155	314.622727	
min	23.000000	9.000000	6.000000	
25%	149.000000	66.000000	30.000000	
50%	279.000000	147.000000	65.000000	
75%	632.000000	364.000000	184.000000	
max	7898.000000	6230.000000	2564.000000	

	actual_time	osrm_time	segment_osrm_distance
count	14787.000000	14787.000000	14787.000000
mean	356.316224	161.667072	222.66823
std	561.528033	272.406218	416.76499
min	9.000000	6.000000	9.07290
25%	67.000000	29.000000	32.57885
50%	148.000000	60.000000	69.78420
75%	367.000000	168.000000	216.46395
max	6265.000000	2032.000000	3523.63240

```
[42]: data.describe(include=object)
```

```
[42]:
```

	route_type	trip_uuid	trip_creation_time	\
count	14787	14787	14787	
unique	2	14787	14787	
top	Carting	trip-153671042288605164	2018-09-12 00:00:22.886430	
freq	8906	1	1	

	source_center	source_name	destination_center	\
count	14787	14787	14787	
unique	930	930	1035	
top	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	IND000000ACB	
freq	1052	1052	821	

	destination_name	od_start_time	\
--	------------------	---------------	---

count	14787	14787
unique	1035	14787
top	Gurgaon_Bilaspur_HB (Haryana)	2018-09-12 02:03:09.655591
freq	821	1

	od_end_time
count	14787
unique	14787
top	2018-09-12 02:03:09.655591
freq	1

6 Feature Generation

```
[43]: #Feature generation like source_state and destination_state
data["source_state"]=data["source_name"].apply(lambda x: str(x).split("(")[1][:
    ↪-1])
data["destination_state"]=data["destination_name"].apply(lambda x: str(x).
    ↪split("(")[1][::-1])
data
```

```
[43]:
```

	route_type	trip_uuid	trip_creation_time \
0	Carting	trip-153671042288605164	2018-09-12 00:00:22.886430
1	Carting	trip-153671046011330457	2018-09-12 00:01:00.113710
2	Carting	trip-153671055416136166	2018-09-12 00:02:34.161600
3	Carting	trip-153671066201138152	2018-09-12 00:04:22.011653
4	Carting	trip-153671066826362165	2018-09-12 00:04:28.263977
...
14782	FTL	trip-153861004148234782	2018-10-03 23:40:41.482736
14783	FTL	trip-153861007249500192	2018-10-03 23:41:12.495257
14784	FTL	trip-153861014185597051	2018-10-03 23:42:21.856227
14785	FTL	trip-153861023893369544	2018-10-03 23:43:58.933947
14786	FTL	trip-153861118270144424	2018-10-03 23:59:42.701692

	source_center	source_name	destination_center \
0	IND561203AAB	Doddablpur_ChikaDPP_D (Karnataka)	IND561203AAB
1	IND400072AAB	Mumbai Hub (Maharashtra)	IND401104AAA
2	IND600056AAA	Chennai_Poonamallee (Tamil Nadu)	IND600056AAA
3	IND600044AAD	Chennai_Chrompet_DPC (Tamil Nadu)	IND600048AAA
4	IND560043AAC	HBR Layout PC (Karnataka)	IND560043AAC
...
14782	IND814101AAB	Dumka_Dudhani_D (Jharkhand)	IND815351AAA
14783	IND842001AAA	Muzaffrpur_Bbganj_I (Bihar)	IND842001AAA
14784	IND206001AAA	Etawah_MhraChng_D (Uttar Pradesh)	IND209304AAA
14785	IND382715AAA	Kadi_KaranNGR_D (Gujarat)	IND382715AAA
14786	IND583119AAA	Sandur_WrdN1DPP_D (Karnataka)	IND583119AAA

	destination_name	od_start_time \
0	Doddablpur_ChikaDPP_D (Karnataka)	2018-09-12 02:03:09.655591
1	Mumbai_MiraRd_IP (Maharashtra)	2018-09-12 00:01:00.113710
2	Chennai_Poonamallee (Tamil Nadu)	2018-09-12 02:12:10.755603
3	Chennai_Vandalur_Dc (Tamil Nadu)	2018-09-12 00:04:22.011653
4	HBR Layout PC (Karnataka)	2018-09-12 00:04:28.263977
...
14782	Jamtara_D (Jharkhand)	2018-10-04 04:22:21.025250
14783	Muzaffrpur_Bbganj_I (Bihar)	2018-10-03 23:41:12.495257
14784	Kanpur_Central_H_6 (Uttar Pradesh)	2018-10-05 02:44:50.858859
14785	Kadi_KaranNGR_D (Gujarat)	2018-10-04 01:48:54.382343
14786	Sandur_WrdN1DPP_D (Karnataka)	2018-10-04 03:58:40.726547

	od_end_time	cutoff_factor \
0	2018-09-12 02:03:09.655591	72
1	2018-09-12 01:41:29.809822	17
2	2018-09-12 02:12:10.755603	24
3	2018-09-12 01:42:22.349694	9
4	2018-09-12 03:00:55.163423	22
...
14782	2018-10-04 02:24:41.382263	167
14783	2018-10-04 16:40:41.713085	192
14784	2018-10-04 19:57:34.928573	835
14785	2018-10-04 01:48:54.382343	84
14786	2018-10-04 03:58:40.726547	65

	actual_distance_to_destination	osrm_distance	start_scan_to_end_scan \
0	73.186911	85.1110	180.0
1	17.175274	19.6800	100.0
2	24.597048	28.0647	189.0
3	9.100510	12.0184	98.0
4	22.424210	28.9203	146.0
...
14782	168.396341	207.4975	428.0
14783	194.552260	229.2052	1017.0
14784	836.072017	997.7577	2180.0
14785	84.743813	102.3828	256.0
14786	66.081533	80.5787	353.0

	segment_actual_time	segment_osrm_time	actual_time	osrm_time \
0	141.0	65.0	143.0	68.0
1	59.0	16.0	59.0	15.0
2	60.0	23.0	61.0	23.0
3	24.0	13.0	24.0	13.0
4	64.0	34.0	64.0	34.0
...
14782	347.0	220.0	349.0	220.0

14783	845.0	178.0	847.0	178.0
14784	1660.0	891.0	1674.0	724.0
14785	186.0	92.0	187.0	92.0
14786	274.0	67.0	275.0	68.0

	segment_osrm_distance	source_state	destination_state
0	84.1894	Karnataka	Karnataka
1	19.8766	Maharashtra	Maharashtra
2	28.0647	Tamil Nadu	Tamil Nadu
3	12.0184	Tamil Nadu	Tamil Nadu
4	28.9203	Karnataka	Karnataka
...
14782	209.4499	Jharkhand	Jharkhand
14783	232.5811	Bihar	Bihar
14784	1166.3614	Uttar Pradesh	Uttar Pradesh
14785	107.6915	Gujarat	Gujarat
14786	80.5787	Karnataka	Karnataka

[14787 rows x 20 columns]

```
[44]: data.describe(include=object)
```

```
[44]:
```

	route_type	trip_uuid	trip_creation_time	\
count	14787	14787	14787	
unique	2	14787	14787	
top	Carting	trip-153671042288605164	2018-09-12 00:00:22.886430	
freq	8906	1	1	

	source_center	source_name	destination_center	\
count	14787	14787	14787	
unique	930	930	1035	
top	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	IND000000ACB	
freq	1052	1052	821	

	destination_name	od_start_time	\
count	14787	14787	
unique	1035	14787	
top	Gurgaon_Bilaspur_HB (Haryana)	2018-09-12 02:03:09.655591	
freq	821	1	

	od_end_time	source_state	destination_state
count	14787	14787	14787
unique	14787	29	31
top	2018-09-12 02:03:09.655591	Maharashtra	Maharashtra
freq	1	2714	2561

```
[45]: data.nunique() #unique value in dataframe
```

```
[45]: route_type                2
      trip_uuid                14787
      trip_creation_time       14787
      source_center            930
      source_name              930
      destination_center       1035
      destination_name         1035
      od_start_time            14787
      od_end_time              14787
      cutoff_factor            684
      actual_distance_to_destination 14771
      osrm_distance            14706
      start_scan_to_end_scan    2203
      segment_actual_time       1887
      segment_osrm_time         1242
      actual_time               1850
      osrm_time                 827
      segment_osrm_distance     14724
      source_state              29
      destination_state         31
      dtype: int64
```

```
[46]: data["source_state"].value_counts() #source-statewise trip count
```

```
[46]: source_state
      Maharashtra            2714
      Karnataka              2143
      Haryana                1823
      Tamil Nadu             1039
      Telangana               784
      Uttar Pradesh           760
      Gujarat                750
      Delhi                  725
      West Bengal             665
      Punjab                 536
      Rajasthan              514
      Andhra Pradesh          435
      Bihar                  351
      Madhya Pradesh          318
      Kerala                 289
      Assam                  268
      Jharkhand               160
      Uttarakhand            114
      Orissa                 107
      Chandigarh              93
      Goa                    65
      Chhattisgarh           43
```


Himachal Pradesh	34
Jammu & Kashmir	17
Dadra and Nagar Haveli	15
Pondicherry	12
Nagaland	5
Arunachal Pradesh	4
Mizoram	4

Name: count, dtype: int64

```
[47]: data["destination_state"].value_counts() #destination-statewise trip count
```

```
[47]: destination_state
```

Maharashtra	2561
Karnataka	2294
Haryana	1640
Tamil Nadu	1084
Uttar Pradesh	805
Telangana	784
Gujarat	734
West Bengal	697
Delhi	657
Punjab	617
Rajasthan	550
Andhra Pradesh	442
Bihar	367
Madhya Pradesh	350
Kerala	270
Assam	232
Jharkhand	181
Uttarakhand	122
Orissa	119
Chandigarh	65
Goa	52
Chhattisgarh	43
Himachal Pradesh	42
Arunachal Pradesh	25
Jammu & Kashmir	20
Dadra and Nagar Haveli	17
Meghalaya	8
Mizoram	6
Nagaland	1
Daman & Diu	1
Tripura	1

Name: count, dtype: int64

```
[48]: data["source_name"].value_counts().head()
```

```
[48]: source_name
      Gurgaon_Bilaspur_HB (Haryana)      1052
      Bhiwandi_Mankoli_HB (Maharashtra)   697
      Bangalore_Nelmngla_H (Karnataka)    624
      Bengaluru_Bomsndra_HB (Karnataka)    455
      Pune_Tathawde_H (Maharashtra)       396
      Name: count, dtype: int64
```

```
[49]: data["source_name"].value_counts().tail()
```

```
[49]: source_name
      Chikodi_IndraNgr_D (Karnataka)      1
      Atmakur_IndraNgr_D (Andhra Pradesh)  1
      Jetpur_DC (Gujarat)                 1
      Bantwal_Trmltpl_D (Karnataka)        1
      Sandur_WrdN1DPP_D (Karnataka)        1
      Name: count, dtype: int64
```

```
[50]: data["destination_name"].value_counts().head()
```

```
[50]: destination_name
      Gurgaon_Bilaspur_HB (Haryana)      821
      Bangalore_Nelmngla_H (Karnataka)    548
      Bhiwandi_Mankoli_HB (Maharashtra)   403
      Bengaluru_Bomsndra_HB (Karnataka)    342
      Hyderabad_Shamshbd_H (Telangana)    280
      Name: count, dtype: int64
```

```
[51]: data["source-destination"] = data["source_name"] + data["destination_name"]
```

```
[52]: data["source-destination"].value_counts() #Busiest Corridors
```

```
[52]: source-destination
      Bangalore_Nelmngla_H (Karnataka)Bengaluru_KGAirprt_HB (Karnataka)      151
      Gurgaon_Bilaspur_HB (Haryana)Gurgaon_Bilaspur_HB (Haryana)             123
      Bengaluru_Bomsndra_HB (Karnataka)Bengaluru_KGAirprt_HB (Karnataka)      121
      Bengaluru_KGAirprt_HB (Karnataka)Bangalore_Nelmngla_H (Karnataka)        108
      Bhiwandi_Mankoli_HB (Maharashtra)Mumbai Hub (Maharashtra)               105
      ...
      Khammam_NSTRoad_I (Telangana)Nalgonda_HydRoad_DC (Telangana)              1
      Kolkata_Dankuni_HB (West Bengal)Tarkeshwar_Naraynpr_D (West Bengal)        1
      Bamangola_Central_D_1 (West Bengal)Malda_krshnPly_DC (West Bengal)         1
      Nalbari_Bhgtपुरা_D (Assam)Dhubri_Tetultol_D (Assam)                      1
      Sandur_WrdN1DPP_D (Karnataka)Sandur_WrdN1DPP_D (Karnataka)                1
      Name: count, Length: 2165, dtype: int64
```

```
[53]: #Average distance
data[data["source-destination"]=="Bangalore_Nelmn gla_H_
↳(Karnataka)Bengaluru_KGAirprt_HB_
↳(Karnataka)"]["actual_distance_to_destination"].mean()
```

```
[53]: 28.03163476896394
```

```
[54]: #Average time
data[data["source-destination"]=="Bangalore_Nelmn gla_H_
↳(Karnataka)Bengaluru_KGAirprt_HB (Karnataka)"]["actual_time"].mean()
```

```
[54]: 87.87417218543047
```

```
[55]: data.drop("source-destination",axis=1,inplace=True)
```

```
[56]: data["trip_creation_time"]=pd.to_datetime(df["trip_creation_time"]) #_
↳conversion to datetime datatype
```

```
[57]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14787 entries, 0 to 14786
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   route_type                            14787 non-null  object
1   trip_uuid                             14787 non-null  object
2   trip_creation_time                    14783 non-null  datetime64[ns]
3   source_center                         14787 non-null  object
4   source_name                           14787 non-null  object
5   destination_center                    14787 non-null  object
6   destination_name                       14787 non-null  object
7   od_start_time                         14787 non-null  object
8   od_end_time                           14787 non-null  object
9   cutoff_factor                         14787 non-null  int64
10  actual_distance_to_destination         14787 non-null  float64
11  osrm_distance                         14787 non-null  float64
12  start_scan_to_end_scan                 14787 non-null  float64
13  segment_actual_time                    14787 non-null  float64
14  segment_osrm_time                      14787 non-null  float64
15  actual_time                           14787 non-null  float64
16  osrm_time                             14787 non-null  float64
17  segment_osrm_distance                  14787 non-null  float64
18  source_state                           14787 non-null  object
19  destination_state                      14787 non-null  object
dtypes: datetime64[ns](1), float64(8), int64(1), object(10)
memory usage: 2.3+ MB
```

```
[58]: #Feature generation year
data["trip_creation_year"]=data["trip_creation_time"].dt.year
data["trip_creation_year"].value_counts()
```

```
[58]: trip_creation_year
2018.0    14783
Name: count, dtype: int64
```

```
[59]: #Feature generation month
data["trip_creation_month"]=data["trip_creation_time"].dt.month
data["trip_creation_month"].value_counts()
```

```
[59]: trip_creation_month
9.0      13092
10.0     1691
Name: count, dtype: int64
```

```
[60]: #Feature generation day
data["trip_creation_day"]=data["trip_creation_time"].dt.day
data["trip_creation_day"].value_counts()
```

```
[60]: trip_creation_day
25.0     1024
17.0     1000
20.0      854
23.0      820
15.0      809
12.0      779
14.0      762
28.0      731
3.0       695
24.0      674
16.0      657
21.0      657
26.0      642
18.0      580
19.0      571
30.0      552
22.0      544
1.0       539
13.0      516
29.0      463
27.0      457
2.0       457
Name: count, dtype: int64
```

```
[61]: #Feature generation triptime
data["od_start_time"]=pd.to_datetime(data["od_start_time"])
data["od_end_time"]=pd.to_datetime(data["od_end_time"])
data["trip_time"]=data["od_end_time"]-data["od_start_time"]
data
```

```
[61]:
```

	route_type	trip_uuid	trip_creation_time	\
0	Carting	trip-153671042288605164	2018-09-20 02:35:36.476840	
1	Carting	trip-153671046011330457	2018-09-20 02:35:36.476840	
2	Carting	trip-153671055416136166	2018-09-20 02:35:36.476840	
3	Carting	trip-153671066201138152	2018-09-20 02:35:36.476840	
4	Carting	trip-153671066826362165	2018-09-20 02:35:36.476840	
...	
14782	FTL	trip-153861004148234782	2018-09-24 05:06:56.558662	
14783	FTL	trip-153861007249500192	2018-09-24 05:06:56.558662	
14784	FTL	trip-153861014185597051	2018-09-24 05:06:56.558662	
14785	FTL	trip-153861023893369544	2018-09-24 05:06:56.558662	
14786	FTL	trip-153861118270144424	2018-09-24 05:06:56.558662	

	source_center	source_name	destination_center	\
0	IND561203AAB	Doddablpur_ChikaDPP_D (Karnataka)	IND561203AAB	
1	IND400072AAB	Mumbai Hub (Maharashtra)	IND401104AAA	
2	IND600056AAA	Chennai_Poonamallee (Tamil Nadu)	IND600056AAA	
3	IND600044AAD	Chennai_Chrompet_DPC (Tamil Nadu)	IND600048AAA	
4	IND560043AAC	HBR Layout PC (Karnataka)	IND560043AAC	
...	
14782	IND814101AAB	Dumka_Dudhani_D (Jharkhand)	IND815351AAA	
14783	IND842001AAA	Muzaffrpur_Bbganj_I (Bihar)	IND842001AAA	
14784	IND206001AAA	Etawah_MhraChng_D (Uttar Pradesh)	IND209304AAA	
14785	IND382715AAA	Kadi_KaranNGR_D (Gujarat)	IND382715AAA	
14786	IND583119AAA	Sandur_WrdN1DPP_D (Karnataka)	IND583119AAA	

	destination_name	od_start_time	\
0	Doddablpur_ChikaDPP_D (Karnataka)	2018-09-12 02:03:09.655591	
1	Mumbai_MiraRd_IP (Maharashtra)	2018-09-12 00:01:00.113710	
2	Chennai_Poonamallee (Tamil Nadu)	2018-09-12 02:12:10.755603	
3	Chennai_Vandalur_Dc (Tamil Nadu)	2018-09-12 00:04:22.011653	
4	HBR Layout PC (Karnataka)	2018-09-12 00:04:28.263977	
...	
14782	Jamtara_D (Jharkhand)	2018-10-04 04:22:21.025250	
14783	Muzaffrpur_Bbganj_I (Bihar)	2018-10-03 23:41:12.495257	
14784	Kanpur_Central_H_6 (Uttar Pradesh)	2018-10-05 02:44:50.858859	
14785	Kadi_KaranNGR_D (Gujarat)	2018-10-04 01:48:54.382343	
14786	Sandur_WrdN1DPP_D (Karnataka)	2018-10-04 03:58:40.726547	

	od_end_time	cutoff_factor	...	segment_osrm_time	\
0	2018-09-12 02:03:09.655591	72	...	65.0	

1	2018-09-12 01:41:29.809822	17	...	16.0
2	2018-09-12 02:12:10.755603	24	...	23.0
3	2018-09-12 01:42:22.349694	9	...	13.0
4	2018-09-12 03:00:55.163423	22	...	34.0
...
14782	2018-10-04 02:24:41.382263	167	...	220.0
14783	2018-10-04 16:40:41.713085	192	...	178.0
14784	2018-10-04 19:57:34.928573	835	...	891.0
14785	2018-10-04 01:48:54.382343	84	...	92.0
14786	2018-10-04 03:58:40.726547	65	...	67.0

	actual_time	osrm_time	segment_osrm_distance	source_state \
0	143.0	68.0	84.1894	Karnataka
1	59.0	15.0	19.8766	Maharashtra
2	61.0	23.0	28.0647	Tamil Nadu
3	24.0	13.0	12.0184	Tamil Nadu
4	64.0	34.0	28.9203	Karnataka
...
14782	349.0	220.0	209.4499	Jharkhand
14783	847.0	178.0	232.5811	Bihar
14784	1674.0	724.0	1166.3614	Uttar Pradesh
14785	187.0	92.0	107.6915	Gujarat
14786	275.0	68.0	80.5787	Karnataka

	destination_state	trip_creation_year	trip_creation_month \
0	Karnataka	2018.0	9.0
1	Maharashtra	2018.0	9.0
2	Tamil Nadu	2018.0	9.0
3	Tamil Nadu	2018.0	9.0
4	Karnataka	2018.0	9.0
...
14782	Jharkhand	2018.0	9.0
14783	Bihar	2018.0	9.0
14784	Uttar Pradesh	2018.0	9.0
14785	Gujarat	2018.0	9.0
14786	Karnataka	2018.0	9.0

	trip_creation_day	trip_time
0	20.0	0 days 00:00:00
1	20.0	0 days 01:40:29.696112
2	20.0	0 days 00:00:00
3	20.0	0 days 01:38:00.338041
4	20.0	0 days 02:56:26.899446
...
14782	24.0	-1 days +22:02:20.357013
14783	24.0	0 days 16:59:29.217828
14784	24.0	-1 days +17:12:44.069714

14785	24.0	0 days 00:00:00
14786	24.0	0 days 00:00:00

[14787 rows x 24 columns]

```
[62]: data.isnull().sum()
```

```
[62]: route_type           0
      trip_uuid           0
      trip_creation_time   4
      source_center        0
      source_name          0
      destination_center   0
      destination_name     0
      od_start_time        0
      od_end_time          0
      cutoff_factor        0
      actual_distance_to_destination 0
      osrm_distance        0
      start_scan_to_end_scan 0
      segment_actual_time   0
      segment_osrm_time     0
      actual_time           0
      osrm_time             0
      segment_osrm_distance 0
      source_state          0
      destination_state     0
      trip_creation_year    4
      trip_creation_month   4
      trip_creation_day     4
      trip_time             0
      dtype: int64
```

```
[63]: data.dropna(inplace=True)
```

```
[64]: data.isnull().sum()
```

```
[64]: route_type           0
      trip_uuid           0
      trip_creation_time   0
      source_center        0
      source_name          0
      destination_center   0
      destination_name     0
      od_start_time        0
      od_end_time          0
      cutoff_factor        0
```

```

actual_distance_to_destination    0
osrm_distance                    0
start_scan_to_end_scan          0
segment_actual_time              0
segment_osrm_time               0
actual_time                     0
osrm_time                       0
segment_osrm_distance            0
source_state                    0
destination_state               0
trip_creation_year              0
trip_creation_month             0
trip_creation_day               0
trip_time                       0
dtype: int64

```

```
[65]: data.shape
```

```
[65]: (14783, 24)
```

```
[66]: #conversion of triptime to float type data
data["triptime_sec"]=data["trip_time"].dt.total_seconds()
```

```
[67]: data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Index: 14783 entries, 0 to 14786
Data columns (total 25 columns):
 #   Column                                Non-Null Count  Dtype
---  -
 0   route_type                          14783 non-null  object
 1   trip_uuid                          14783 non-null  object
 2   trip_creation_time                 14783 non-null  datetime64[ns]
 3   source_center                     14783 non-null  object
 4   source_name                       14783 non-null  object
 5   destination_center                 14783 non-null  object
 6   destination_name                   14783 non-null  object
 7   od_start_time                     14783 non-null  datetime64[ns]
 8   od_end_time                       14783 non-null  datetime64[ns]
 9   cutoff_factor                     14783 non-null  int64
10   actual_distance_to_destination     14783 non-null  float64
11   osrm_distance                     14783 non-null  float64
12   start_scan_to_end_scan            14783 non-null  float64
13   segment_actual_time               14783 non-null  float64
14   segment_osrm_time                 14783 non-null  float64
15   actual_time                       14783 non-null  float64
16   osrm_time                         14783 non-null  float64

```



```

17 segment_osrm_distance      14783 non-null float64
18 source_state               14783 non-null object
19 destination_state          14783 non-null object
20 trip_creation_year          14783 non-null float64
21 trip_creation_month         14783 non-null float64
22 trip_creation_day           14783 non-null float64
23 trip_time                   14783 non-null timedelta64[ns]
24 triptime_sec                14783 non-null float64
dtypes: datetime64[ns](3), float64(12), int64(1), object(8), timedelta64[ns](1)
memory usage: 2.9+ MB

```

```
[68]: data[data["triptime_sec"]<0]
```

```

[68]:      route_type      trip_uuid      trip_creation_time \
5      Carting  trip-153671074033284934  2018-09-20 02:35:36.476840
14     Carting  trip-153671202698783427  2018-09-23 06:42:06.021680
16     Carting  trip-153671225291120891  2018-09-14 15:42:46.437249
31     Carting  trip-153671440490445199  2018-09-13 20:44:19.424489
35     Carting  trip-153671508851597828  2018-09-29 22:21:45.149226
...
14768    FTL  trip-153860767482259863  2018-09-24 05:06:56.558662
14779    FTL  trip-153860945742225615  2018-09-24 05:06:56.558662
14781    FTL  trip-153860985527721606  2018-09-24 05:06:56.558662
14782    FTL  trip-153861004148234782  2018-09-24 05:06:56.558662
14784    FTL  trip-153861014185597051  2018-09-24 05:06:56.558662

      source_center      source_name destination_center \
5      IND395009AAA      Surat_Central_D_12 (Gujarat)      IND395004AAB
14     IND395001AAA      Surat_Central_D_9 (Gujarat)      IND395006AAA
16     IND712103AAA      Hoogly_Bandel_D (West Bengal)      IND712124AAA
31     IND140501AAA      Lalru_OnkarDPP_D (Punjab)      IND134203AAA
35     IND360530AAB      Jamjodhpur_Court_D (Gujarat)      IND360575AAA
...
14768  IND505122AAA      Jammikunta_ConduDPP_D (Telangana)      IND505467AAA
14779  IND140001AAA      RoopNagar_ChotiHvl_DC (Punjab)      IND140301AAA
14781  IND814133AAB      Godda_Central_D_2 (Jharkhand)      IND815301AAA
14782  IND814101AAB      Dumka_Dudhani_D (Jharkhand)      IND815351AAA
14784  IND206001AAA      Etawah_MhraChng_D (Uttar Pradesh)      IND209304AAA

      destination_name      od_start_time \
5      Surat_Central_D_3 (Gujarat)  2018-09-12 02:31:39.246238
14     Surat_Varachha_DC (Gujarat)  2018-09-12 02:37:19.832796
16     Hooghly_DC (West Bengal)  2018-09-12 03:09:08.473151
31     Naraingarh_Ward2DPP_D (Haryana)  2018-09-12 07:36:00.152620
35     Porbandar_DC (Gujarat)  2018-09-12 06:04:58.698852
...
14768  Husnabad_Greenmkt_D (Telangana)  2018-10-04 03:51:10.928009

```

14779	Chandigarh_Kharar_DC (Chandigarh)	2018-10-04 03:46:12.300247
14781	Giridih_Shivalya_D (Jharkhand)	2018-10-04 08:29:20.440999
14782	Jamtara_D (Jharkhand)	2018-10-04 04:22:21.025250
14784	Kanpur_Central_H_6 (Uttar Pradesh)	2018-10-05 02:44:50.858859

	od_end_time	cutoff_factor	...	actual_time	osrm_time	\
5	2018-09-12 02:01:41.638015	25	...	161.0	29.0	
14	2018-09-12 02:04:22.360575	19	...	170.0	29.0	
16	2018-09-12 02:16:17.710493	51	...	222.0	58.0	
31	2018-09-12 03:55:15.023521	47	...	147.0	64.0	
35	2018-09-12 03:43:56.169739	178	...	553.0	192.0	
...	
14768	2018-10-04 02:25:04.243970	104	...	380.0	119.0	
14779	2018-10-04 02:52:02.434753	183	...	281.0	207.0	
14781	2018-10-04 03:01:57.954149	226	...	511.0	248.0	
14782	2018-10-04 02:24:41.382263	167	...	349.0	220.0	
14784	2018-10-04 19:57:34.928573	835	...	1674.0	724.0	

	segment_osrm_distance	source_state	destination_state	\
5	30.9358	Gujarat	Gujarat	
14	30.5457	Gujarat	Gujarat	
16	71.3328	West Bengal	West Bengal	
31	103.6903	Punjab	Haryana	
35	245.2043	Gujarat	Gujarat	
...	
14768	140.2444	Telangana	Telangana	
14779	216.3882	Punjab	Chandigarh	
14781	378.6774	Jharkhand	Jharkhand	
14782	209.4499	Jharkhand	Jharkhand	
14784	1166.3614	Uttar Pradesh	Uttar Pradesh	

	trip_creation_year	trip_creation_month	trip_creation_day	\
5	2018.0	9.0	20.0	
14	2018.0	9.0	23.0	
16	2018.0	9.0	14.0	
31	2018.0	9.0	13.0	
35	2018.0	9.0	29.0	
...	
14768	2018.0	9.0	24.0	
14779	2018.0	9.0	24.0	
14781	2018.0	9.0	24.0	
14782	2018.0	9.0	24.0	
14784	2018.0	9.0	24.0	

	trip_time	triptime_sec
5	-1 days +23:30:02.391777	-1797.608223
14	-1 days +23:27:02.527779	-1977.472221

```

16      -1 days +23:07:09.237342  -3170.762658
31      -1 days +20:19:14.870901  -13245.129099
35      -1 days +21:38:57.470887  -8462.529113
...
14768 -1 days +22:33:53.315961  -5166.684039
14779 -1 days +23:05:50.134506  -3249.865494
14781 -1 days +18:32:37.513150  -19642.486850
14782 -1 days +22:02:20.357013  -7059.642987
14784 -1 days +17:12:44.069714  -24435.930286

```

[891 rows x 25 columns]

```

[69]: #Here Triptime can not be negative values as travelling time should always be
      ↪positive, so we will drop that rows as its false values
      data.drop(data[data["triptime_sec"]<0].index,inplace=True)

```

```

[70]: data.describe()

```

```

[70]:
      trip_creation_time      od_start_time \
count      13892      13892
mean  2018-09-22 13:07:02.641116416  2018-09-22 13:44:17.722910976
min    2018-09-12 00:25:19.499696    2018-09-12 00:01:00.113710
25%    2018-09-17 03:21:23.289982976  2018-09-17 04:12:26.345313536
50%    2018-09-22 04:51:35.609723904  2018-09-22 04:52:36.690818048
75%    2018-09-27 17:28:45.461110016  2018-09-27 19:53:56.263667968
max     2018-10-03 23:59:42.701692    2018-10-06 04:27:23.392375
std                                NaN                                NaN

      od_end_time  cutoff_factor \
count      13892    13892.000000
mean  2018-09-22 20:29:21.842143744    159.986251
min    2018-09-12 00:50:10.814399      9.000000
25%    2018-09-17 10:16:06.118688512    21.000000
50%    2018-09-22 12:22:51.642557184    46.000000
75%    2018-09-28 02:16:22.134860288   148.000000
max     2018-10-08 03:00:24.353479   2185.000000
std                                NaN    307.520122

      actual_distance_to_destination  osrm_distance  start_scan_to_end_scan \
count      13892.000000    13892.000000      13892.000000
mean          160.852618      200.437664      515.603657
min           9.002461       9.072900       23.000000
25%          22.037144      29.802900      144.000000
50%          46.163919      61.108100      264.000000
75%         149.281573     193.689125      595.000000
max        2187.483994     2840.081000     7898.000000
std         307.627703     373.619022      660.357703

```

	segment_actual_time	segment_osrm_time	actual_time	osrm_time \
count	13892.000000	13892.000000	13892.000000	13892.000000
mean	346.118557	176.627627	349.267492	157.980564
min	9.000000	6.000000	9.000000	6.000000
25%	64.000000	30.000000	65.000000	29.000000
50%	136.000000	62.000000	138.000000	57.500000
75%	349.000000	176.000000	353.000000	163.250000
max	6230.000000	2564.000000	6265.000000	2032.000000
std	561.918712	316.580388	567.051777	274.050917

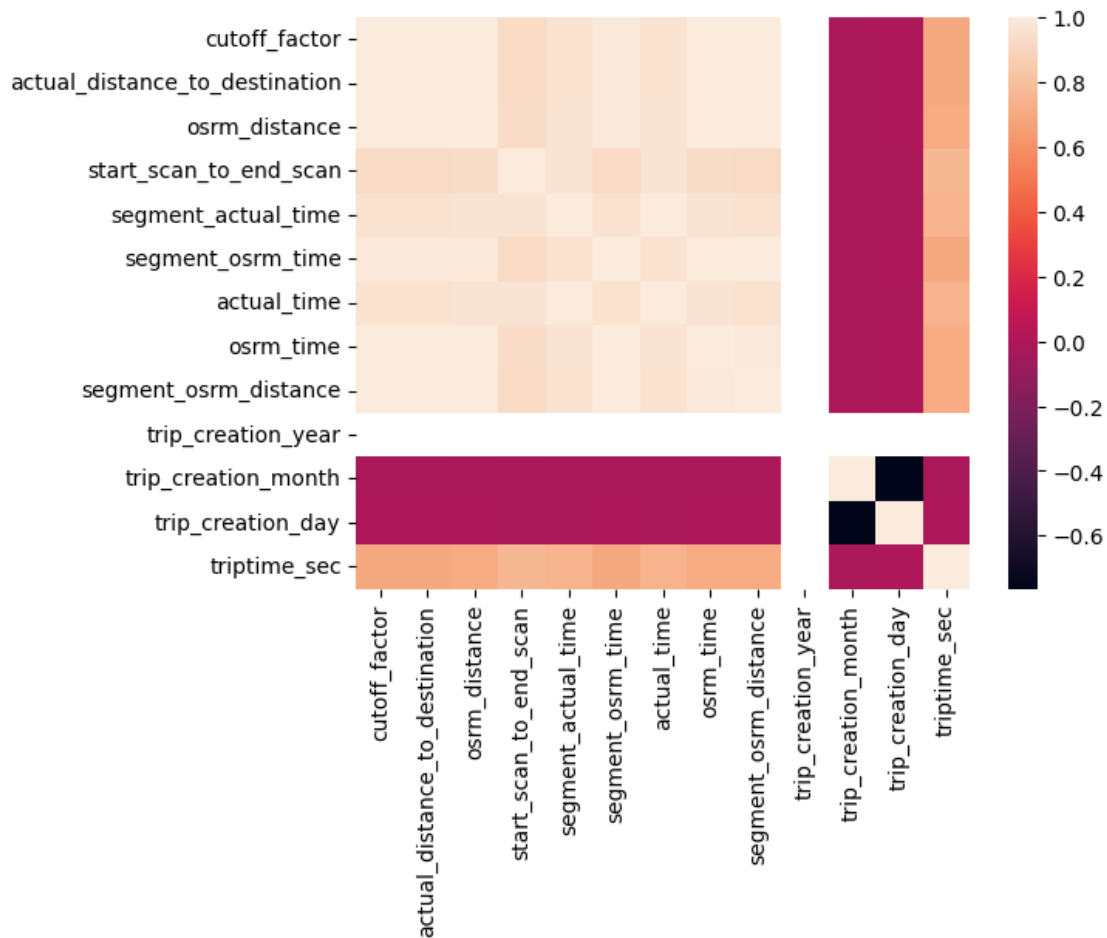
	segment_osrm_distance	trip_creation_year	trip_creation_month \
count	13892.000000	13892.0	13892.000000
mean	218.437771	2018.0	9.114310
min	9.072900	2018.0	9.000000
25%	31.349950	2018.0	9.000000
50%	65.614850	2018.0	9.000000
75%	204.588700	2018.0	9.000000
max	3523.632400	2018.0	10.000000
std	419.933226	0.0	0.318199

	trip_creation_day	trip_time	triptime_sec
count	13892.000000	13892	13892.000000
mean	18.547653	0 days 06:45:04.119232843	24304.119233
min	1.000000	0 days 00:00:00	0.000000
25%	14.000000	0 days 01:51:56.299656750	6716.299657
50%	20.000000	0 days 03:30:54.417636	12654.417636
75%	25.000000	0 days 07:04:21.465968500	25461.465969
max	30.000000	5 days 11:38:33.117274	473913.117274
std	7.753191	0 days 09:33:47.061593031	34427.061593

```
[72]: #Heatmap of dataframe
data_numeric = data.select_dtypes(include=['float64', 'int64'])

sns.heatmap(data_numeric.corr())
```

```
[72]: <Axes: >
```



```
[73]: data_numeric.corr()
```

```
[73]:
```

	cutoff_factor	actual_distance_to_destination \
cutoff_factor	1.000000	0.999997
actual_distance_to_destination	0.999997	1.000000
osrm_distance	0.997444	0.997471
start_scan_to_end_scan	0.921081	0.921345
segment_actual_time	0.954552	0.954682
segment_osrm_time	0.988451	0.988498
actual_time	0.955434	0.955563
osrm_time	0.994283	0.994361
segment_osrm_distance	0.993405	0.993410
trip_creation_year	NaN	NaN
trip_creation_month	-0.019440	-0.019482
trip_creation_day	-0.011132	-0.011104
triptime_sec	0.703598	0.703577

	osrm_distance	start_scan_to_end_scan \
cutoff_factor	0.997444	0.921081
actual_distance_to_destination	0.997471	0.921345
osrm_distance	1.000000	0.927050
start_scan_to_end_scan	0.927050	1.000000
segment_actual_time	0.959676	0.963516
segment_osrm_time	0.992424	0.921375
actual_time	0.960492	0.963525
osrm_time	0.997933	0.929408
segment_osrm_distance	0.995036	0.922120
trip_creation_year	NaN	NaN
trip_creation_month	-0.018993	-0.019450
trip_creation_day	-0.011031	-0.014236
triptime_sec	0.704835	0.765778

	segment_actual_time	segment_osrm_time \
cutoff_factor	0.954552	0.988451
actual_distance_to_destination	0.954682	0.988498
osrm_distance	0.959676	0.992424
start_scan_to_end_scan	0.963516	0.921375
segment_actual_time	1.000000	0.954571
segment_osrm_time	0.954571	1.000000
actual_time	0.999978	0.955367
osrm_time	0.959483	0.993647
segment_osrm_distance	0.957497	0.996487
trip_creation_year	NaN	NaN
trip_creation_month	-0.017506	-0.019008
trip_creation_day	-0.013692	-0.010119
triptime_sec	0.745170	0.701954

	actual_time	osrm_time	segment_osrm_distance \
cutoff_factor	0.955434	0.994283	0.993405
actual_distance_to_destination	0.955563	0.994361	0.993410
osrm_distance	0.960492	0.997933	0.995036
start_scan_to_end_scan	0.963525	0.929408	0.922120
segment_actual_time	0.999978	0.959483	0.957497
segment_osrm_time	0.955367	0.993647	0.996487
actual_time	1.000000	0.960269	0.958320
osrm_time	0.960269	1.000000	0.992408
segment_osrm_distance	0.958320	0.992408	1.000000
trip_creation_year	NaN	NaN	NaN
trip_creation_month	-0.017533	-0.020042	-0.019023
trip_creation_day	-0.013676	-0.010269	-0.010496
triptime_sec	0.745080	0.704238	0.707998

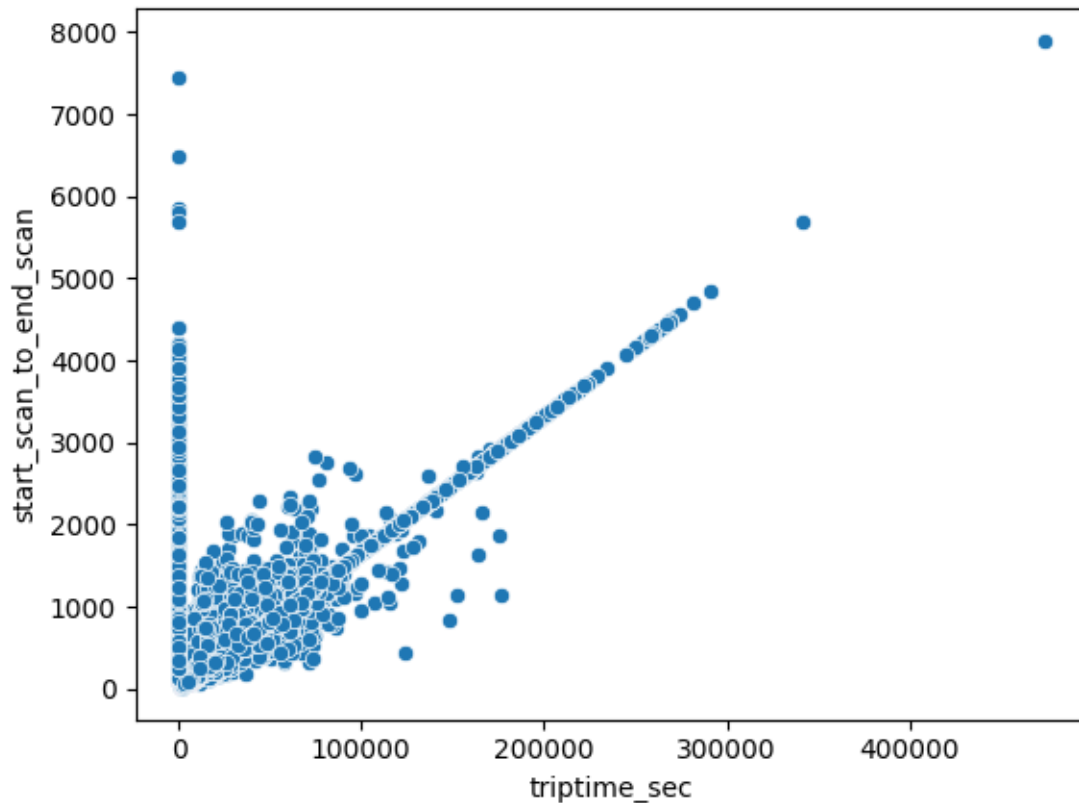
	trip_creation_year	trip_creation_month \
cutoff_factor	NaN	-0.019440

actual_distance_to_destination	NaN	-0.019482
osrm_distance	NaN	-0.018993
start_scan_to_end_scan	NaN	-0.019450
segment_actual_time	NaN	-0.017506
segment_osrm_time	NaN	-0.019008
actual_time	NaN	-0.017533
osrm_time	NaN	-0.020042
segment_osrm_distance	NaN	-0.019023
trip_creation_year	NaN	NaN
trip_creation_month	NaN	1.000000
trip_creation_day	NaN	-0.762671
triptime_sec	NaN	-0.015866

	trip_creation_day	triptime_sec
cutoff_factor	-0.011132	0.703598
actual_distance_to_destination	-0.011104	0.703577
osrm_distance	-0.011031	0.704835
start_scan_to_end_scan	-0.014236	0.765778
segment_actual_time	-0.013692	0.745170
segment_osrm_time	-0.010119	0.701954
actual_time	-0.013676	0.745080
osrm_time	-0.010269	0.704238
segment_osrm_distance	-0.010496	0.707998
trip_creation_year	NaN	NaN
trip_creation_month	-0.762671	-0.015866
trip_creation_day	1.000000	-0.009683
triptime_sec	-0.009683	1.000000

```
[74]: #Visualization of triptime and start_scan_to_end_scan
sns.scatterplot(x=data["triptime_sec"], y=data["start_scan_to_end_scan"])
```

```
[74]: <Axes: xlabel='triptime_sec', ylabel='start_scan_to_end_scan'>
```



7 Hypothesis Testing

- Pearson Test between triptime and start_scan_to_end_scan

H0: Both Variables are not correlated

Ha: Both variables are correlated

```
[75]: #Let us set siginificance level 0.05, confidence level 95%
alpha=0.05
```

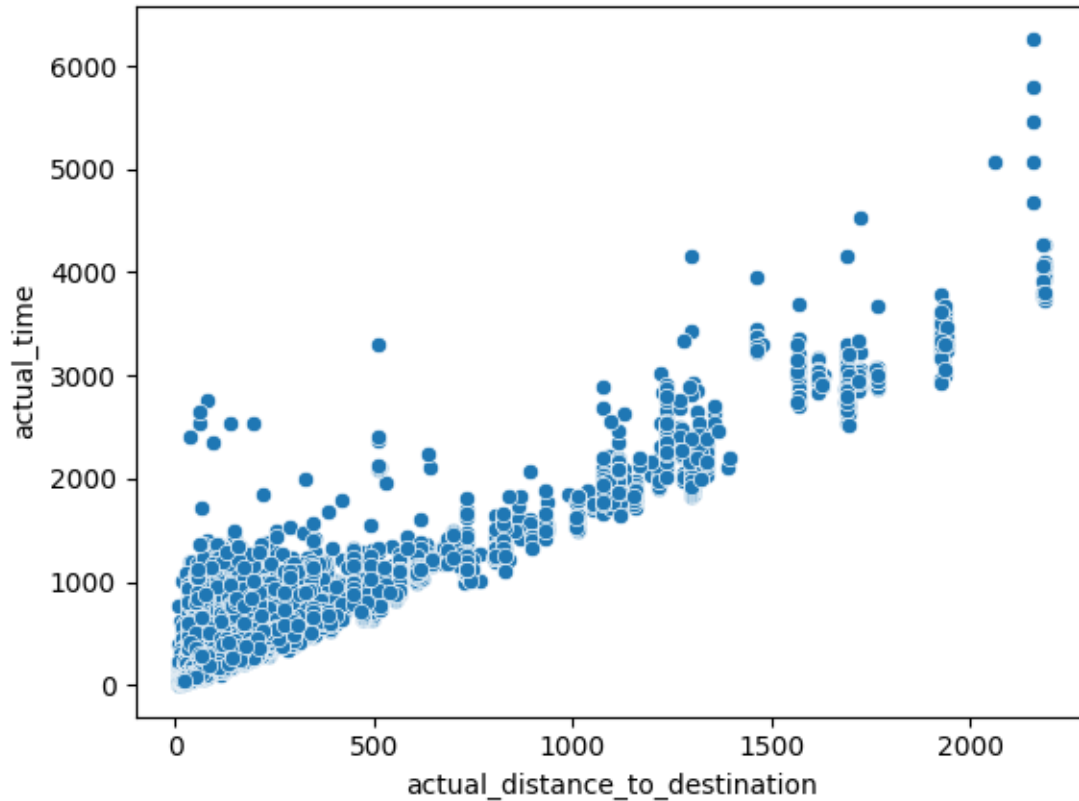
```
[76]: test_statistics,p_value=pearsonr(data["triptime_sec"],
    ↪data["start_scan_to_end_scan"])
print(p_value)
if p_value < alpha:
    print("Reject Null Hypotheis, Both Variables are correlated")
else:
    print("Fail to Reject Null Hypothesis,Both Variables are not correlated")
```

0.0

Reject Null Hypotheis, Both Variables are correlated


```
[77]: #Visualization between distance and time
sns.scatterplot(x=data["actual_distance_to_destination"],y=data["actual_time"])
```

```
[77]: <Axes: xlabel='actual_distance_to_destination', ylabel='actual_time'>
```



- Pearson Test actual_time and actual_distance_to_destination

H0: Both Variables are not correlated

Ha: Both variables are correlated

```
[78]: #Let us set significance level 0.05, confidence level 95%
alpha=0.05
```

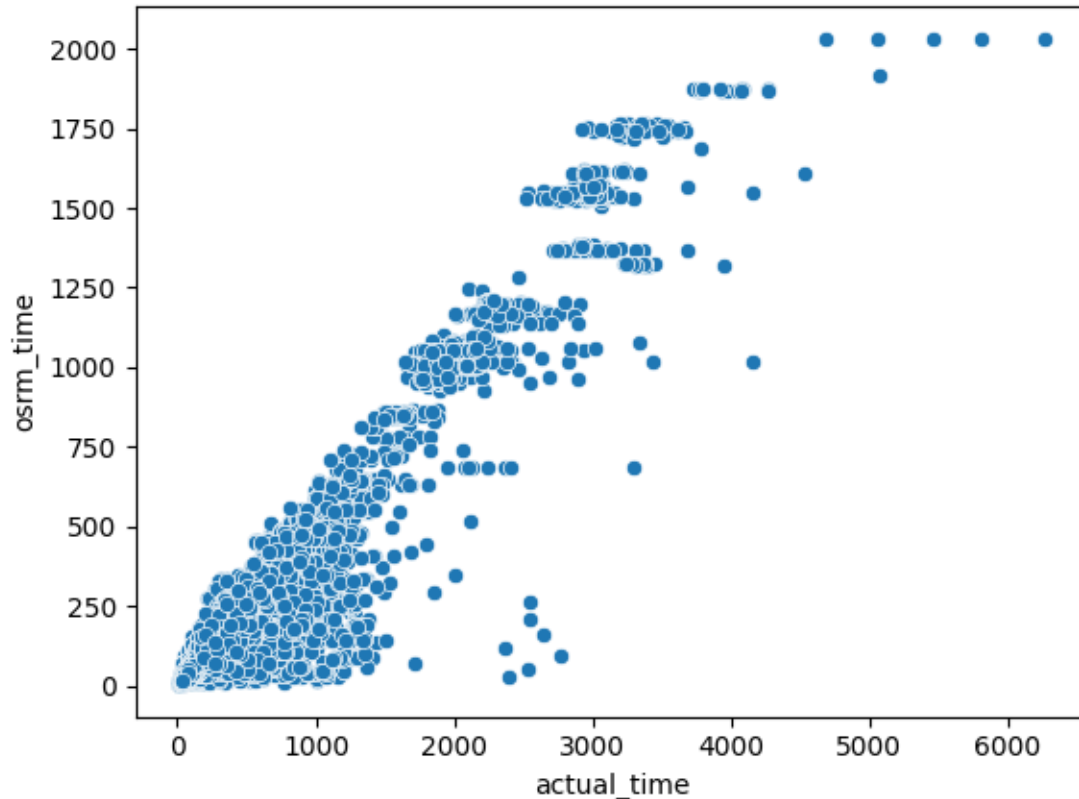
```
[79]: test_statistics,p_value=pearsonr(data["actual_time"],data["actual_distance_to_destination"])
print(p_value)
if p_value < alpha:
    print("Reject Null Hypotheis, Both Variables are correlated")
else:
    print("Fail to Reject Null Hypothesis,Both Variables are not correlated")
```

0.0

Reject Null Hypotheis, Both Variables are correlated

```
[80]: #Visualization between distance and time
sns.scatterplot(x=data["actual_time"],y=data["osrm_time"])
```

```
[80]: <Axes: xlabel='actual_time', ylabel='osrm_time'>
```



- T-Test for actual_time and osrm_time

H0: Mean of actual_time and osrm_time are same ($\mu_1 = \mu_2$)

Ha: Mean of actual_time is higher than osrm_time ($\mu_1 > \mu_2$)

```
[81]: #Let us set significance level 0.05, confidence level 95%
alpha=0.05
```

```
[82]: test_statistics,p_value=ttest_ind(data["actual_time"],data["osrm_time"],
    ↪alternative="greater")
print(p_value)
if p_value < alpha:
    print("Reject Null Hypotheis, Mean of actual_time and osrm_time are same")
else:
    print("Fail to Reject Null Hypothesis,ean of actual_time is higher than
    ↪osrm_time")
```

1.0113592493195362e-274

Reject Null Hypothesis, Mean of actual_time and osrm_time are same

- Pearson Test actual_time and osrm_time

H0: Both Variables are not correlated

Ha: Both variables are correlated

```
[83]: #Let us set significance level 0.05, confidence level 95%
alpha=0.05
```

```
[84]: test_statistics,p_value=pearsonr(data["actual_time"],data["osrm_time"])
print(p_value)
if p_value < alpha:
    print("Reject Null Hypothesis, Both Variables are correlated")
else:
    print("Fail to Reject Null Hypothesis,Both Variables are not correlated")
```

0.0

Reject Null Hypothesis, Both Variables are correlated

- T-Test for actual_time and segment_actual_time

H0: Mean of actual_time and segment_actual_time are same ($\mu_1 = \mu_2$)

Ha: Mean of actual_time and segment_actual_time are not same ($\mu_1 \neq \mu_2$)

```
[85]: #Let us set significance level 0.05, confidence level 95%
alpha=0.05
```

```
[86]: test_statistics,p_value=ttest_ind(data["actual_time"],data["segment_actual_time"])
print(p_value)
if p_value < alpha:
    print("Reject Null Hypothesis, Mean of actual_time and segment_actual_time_
↪are not same")
else:
    print("Fail to Reject Null Hypothesis,Mean of actual_time and_
↪segment_actual_time are same")
```

0.6419956696137739

Fail to Reject Null Hypothesis,Mean of actual_time and segment_actual_time are same

- T-Test for osrm_time and segment_osrm_time

H0: Mean of osrm_time and segment_osrm_time are same ($\mu_1 = \mu_2$)

Ha: Mean of osrm_time and segment_osrm_time are not same ($\mu_1 \neq \mu_2$)

```
[87]: #Let us set significance level 0.05, confidence level 95%
alpha=0.05
```

```
[88]: test_statistics,p_value=ttest_ind(data["osrm_time"],data["segment_osrm_time"])
print(p_value)
if p_value < alpha:
    print("Reject Null Hypotheis, Mean of osrm_time and segment_osrm_time are_
↪not same")
else:
    print("Fail to Reject Null Hypothesis,Mean of osrm_time and_
↪segment_osrm_time are same")
```

1.5413271810594524e-07

Reject Null Hypotheis, Mean of osrm_time and segment_osrm_time are not same

- T-Test for osrm_distance and segment_osrm_distance

H0: Mean of osrm_distance and segment_osrm_distance are same ($\mu_1 = \mu_2$)

Ha: Mean of osrm_distance and segment_osrm_distance are not same ($\mu_1 \neq \mu_2$)

```
[89]: #Let us set siginificance level 0.05, confidence level 95%
alpha=0.05
```

```
[90]: test_statistics,p_value=ttest_ind(data["osrm_distance"],data["segment_osrm_distance"])
print(p_value)
if p_value < alpha:
    print("Reject Null Hypotheis, Mean of osrm_distance and_
↪segment_osrm_distance are not same")
else:
    print("Fail to Reject Null Hypothesis,Mean of osrm_distance and_
↪segment_osrm_distance are same")
```

0.0001606670222265932

Reject Null Hypotheis, Mean of osrm_distance and segment_osrm_distance are not same

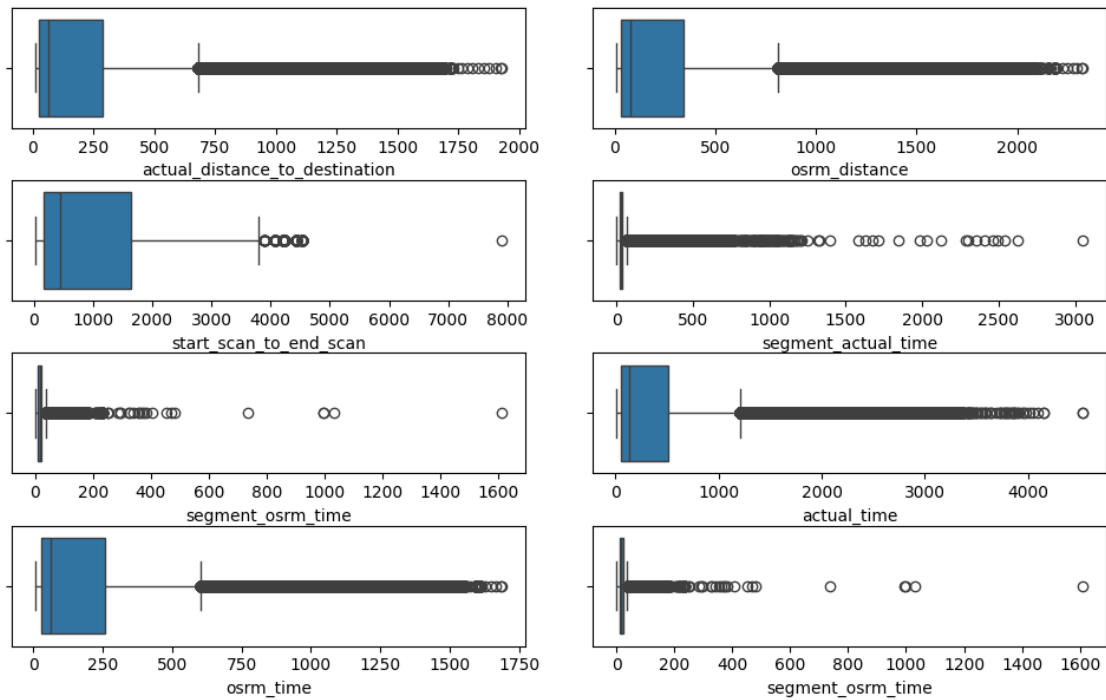
8 Outliers Detection Using IQR Method

```
[96]: num_cols = ['actual_distance_to_destination', 'osrm_distance',_
↪'start_scan_to_end_scan', 'segment_actual_time', 'segment_osrm_time',_
↪'actual_time', 'osrm_time', "segment_osrm_time", "triptime_sec"]
```

```
[110]: fig, axis = plt.subplots(nrows=4, ncols=2, figsize=(10,6))
fig.tight_layout()

index = 0
for row in range(4):
    for col in range(2):
        sns.boxplot(x=df[num_cols[index]], ax=axis[row, col])
        index += 1
```

```
plt.show()
```



```
[111]: data.describe()
```

```
[111]:
```

	trip_creation_time	od_start_time \
count	13892	13892
mean	2018-09-22 13:07:02.641116416	2018-09-22 13:44:17.722910976
min	2018-09-12 00:25:19.499696	2018-09-12 00:01:00.113710
25%	2018-09-17 03:21:23.289982976	2018-09-17 04:12:26.345313536
50%	2018-09-22 04:51:35.609723904	2018-09-22 04:52:36.690818048
75%	2018-09-27 17:28:45.461110016	2018-09-27 19:53:56.263667968
max	2018-10-03 23:59:42.701692	2018-10-06 04:27:23.392375
std	NaN	NaN

	od_end_time	cutoff_factor \
count	13892	13892.000000
mean	2018-09-22 20:29:21.842143744	159.986251
min	2018-09-12 00:50:10.814399	9.000000
25%	2018-09-17 10:16:06.118688512	21.000000
50%	2018-09-22 12:22:51.642557184	46.000000
75%	2018-09-28 02:16:22.134860288	148.000000
max	2018-10-08 03:00:24.353479	2185.000000
std	NaN	307.520122

	actual_distance_to_destination	osrm_distance	start_scan_to_end_scan \
count	13892.000000	13892.000000	13892.000000
mean	160.852618	200.437664	515.603657
min	9.002461	9.072900	23.000000
25%	22.037144	29.802900	144.000000
50%	46.163919	61.108100	264.000000
75%	149.281573	193.689125	595.000000
max	2187.483994	2840.081000	7898.000000
std	307.627703	373.619022	660.357703

	segment_actual_time	segment_osrm_time	actual_time	osrm_time \
count	13892.000000	13892.000000	13892.000000	13892.000000
mean	346.118557	176.627627	349.267492	157.980564
min	9.000000	6.000000	9.000000	6.000000
25%	64.000000	30.000000	65.000000	29.000000
50%	136.000000	62.000000	138.000000	57.500000
75%	349.000000	176.000000	353.000000	163.250000
max	6230.000000	2564.000000	6265.000000	2032.000000
std	561.918712	316.580388	567.051777	274.050917

	segment_osrm_distance	trip_creation_year	trip_creation_month \
count	13892.000000	13892.0	13892.000000
mean	218.437771	2018.0	9.114310
min	9.072900	2018.0	9.000000
25%	31.349950	2018.0	9.000000
50%	65.614850	2018.0	9.000000
75%	204.588700	2018.0	9.000000
max	3523.632400	2018.0	10.000000
std	419.933226	0.0	0.318199

	trip_creation_day	trip_time	triptime_sec
count	13892.000000	13892	13892.000000
mean	18.547653	0 days 06:45:04.119232843	24304.119233
min	1.000000	0 days 00:00:00	0.000000
25%	14.000000	0 days 01:51:56.299656750	6716.299657
50%	20.000000	0 days 03:30:54.417636	12654.417636
75%	25.000000	0 days 07:04:21.465968500	25461.465969
max	30.000000	5 days 11:38:33.117274	473913.117274
std	7.753191	0 days 09:33:47.061593031	34427.061593

9 Outliers Treatment

```
[113]: #Let's remove outliers using IQR Method
```

```
sses25th = 144
sses75th = 595
iqr = sses75th - sses25th
ssesuw = sses75th + 1.5*iqr
ssesuw
```

```
[113]: 1271.5
```

```
[114]: data[data["start_scan_to_end_scan"]> ssesuw]
```

```
[114]:
```

	route_type	trip_uuid	trip_creation_time	\
46	Carting	trip-153671813821616145	2018-09-12 01:33:48.711350	
433	Carting	trip-153679519504536979	2018-09-21 02:54:55.651098	
588	Carting	trip-153681898069638565	2018-10-03 13:07:13.296061	
621	Carting	trip-153682425240623736	2018-09-25 06:14:51.782383	
836	Carting	trip-153687754564273073	2018-09-27 02:48:03.391966	
...	
14748	FTL	trip-153860451596867762	2018-09-24 05:06:56.558662	
14754	FTL	trip-153860570045461434	2018-09-24 05:06:56.558662	
14764	FTL	trip-153860698042160875	2018-09-24 05:06:56.558662	
14773	FTL	trip-153860879439383883	2018-09-24 05:06:56.558662	
14774	FTL	trip-153860880135634048	2018-09-24 05:06:56.558662	

	source_center	source_name	\
46	IND413517AAA	Udgir_NlgaonRd_D (Maharashtra)	
433	IND530012AAA	Visakhapatnam_Gajuwaka_IP (Andhra Pradesh)	
588	IND400072AAI	Mumbai_Chndivli_D (Maharashtra)	
621	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)	
836	IND530012AAA	Visakhapatnam_Gajuwaka_IP (Andhra Pradesh)	
...	
14748	IND712311AAA	Kolkata_Dankuni_HB (West Bengal)	
14754	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	
14764	IND131028AAB	Sonipat_Kundli_H (Haryana)	
14773	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	
14774	IND424006AAA	Dhule_MIDCAvdn_I (Maharashtra)	

	destination_center	destination_name	\
46	IND431603AAA	Nanded_Aswningr_I (Maharashtra)	
433	IND530012AAA	Visakhapatnam_Gajuwaka_IP (Andhra Pradesh)	
588	IND000000AFS	Mumbai_Skynet_INT (Maharashtra)	
621	IND401104AAB	Mumbai_MiraRoad_M (Maharashtra)	
836	IND530012AAA	Visakhapatnam_Gajuwaka_IP (Andhra Pradesh)	
...	
14748	IND712311AAA	Kolkata_Dankuni_HB (West Bengal)	

14754	IND834002AAB	Ranchi_Hub (Jharkhand)
14764	IND131028AAB	Sonipat_Kundli_H (Haryana)
14773	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)
14774	IND425409AAA	Shahada_Nandrbar_D (Maharashtra)

		od_start_time	od_end_time	cutoff_factor	\
46	2018-09-12	09:44:13.402455	2018-09-13 00:46:22.053872	291	
433	2018-09-12	23:33:15.045633	2018-09-13 22:18:58.260294	194	
588	2018-09-13	07:10:45.472473	2018-09-14 06:23:48.451756	9	
621	2018-09-13	07:37:32.406544	2018-09-14 06:17:06.037037	16	
836	2018-09-13	22:25:43.913670	2018-09-14 20:39:35.934714	193	
...		
14748	2018-10-03	22:08:35.968978	2018-10-04 22:26:30.408004	159	
14754	2018-10-03	22:28:20.454881	2018-10-05 08:39:47.996375	1010	
14764	2018-10-05	08:35:15.664489	2018-10-05 08:35:15.664489	1321	
14773	2018-10-06	04:27:23.392375	2018-10-06 04:27:23.392375	1931	
14774	2018-10-03	23:20:01.356596	2018-10-04 03:38:11.949795	211	

	...	actual_time	osrm_time	segment_osrm_distance	source_state	\
46	...	593.0	235.0	337.1176	Maharashtra	
433	...	304.0	168.0	228.2350	Andhra Pradesh	
588	...	113.0	12.0	14.7276	Maharashtra	
621	...	59.0	27.0	29.6053	Maharashtra	
836	...	350.0	167.0	227.4162	Andhra Pradesh	
...	
14748	...	1342.0	145.0	197.2656	West Bengal	
14754	...	1625.0	851.0	1222.2127	Haryana	
14764	...	2003.0	1166.0	1747.4544	Haryana	
14773	...	3307.0	1739.0	2600.9869	Haryana	
14774	...	1293.0	185.0	253.9858	Maharashtra	

	destination_state	trip_creation_year	trip_creation_month	\
46	Maharashtra	2018.0	9.0	
433	Andhra Pradesh	2018.0	9.0	
588	Maharashtra	2018.0	10.0	
621	Maharashtra	2018.0	9.0	
836	Andhra Pradesh	2018.0	9.0	
...	
14748	West Bengal	2018.0	9.0	
14754	Jharkhand	2018.0	9.0	
14764	Haryana	2018.0	9.0	
14773	Haryana	2018.0	9.0	
14774	Maharashtra	2018.0	9.0	

	trip_creation_day	trip_time	triptime_sec
46	12.0 0 days	15:02:08.651417	54128.651417
433	21.0 0 days	22:45:43.214661	81943.214661


```

588          3.0 0 days 23:13:02.979283    83582.979283
621          25.0 0 days 22:39:33.630493    81573.630493
836          27.0 0 days 22:13:52.021044    80032.021044
...
14748        24.0 1 days 00:17:54.439026    87474.439026
14754        24.0 1 days 10:11:27.541494   123087.541494
14764        24.0      0 days 00:00:00         0.000000
14773        24.0      0 days 00:00:00         0.000000
14774        24.0 0 days 04:18:10.593199    15490.593199

```

[1332 rows x 25 columns]

```
[115]: #We will drop outliers which we have found using IQR Method
data.drop(data[data["start_scan_to_end_scan"]> ssesuw].index, inplace=True)
```

```
[116]: data.describe()
```

```
[116]:
```

	trip_creation_time	od_start_time \
count	12560	12560
mean	2018-09-22 15:01:55.974861824	2018-09-22 13:14:45.159265536
min	2018-09-12 00:25:19.499696	2018-09-12 00:01:00.113710
25%	2018-09-17 05:23:36.585742080	2018-09-17 04:06:47.181209856
50%	2018-09-22 09:11:30.250937088	2018-09-22 04:19:12.697853440
75%	2018-09-27 19:50:24.957410560	2018-09-27 19:36:40.066671616
max	2018-10-03 23:59:42.701692	2018-10-04 20:15:07.233819
std	NaN	NaN

	od_end_time	cutoff_factor \
count	12560	12560.000000
mean	2018-09-22 17:59:39.655986432	82.207006
min	2018-09-12 00:50:10.814399	9.000000
25%	2018-09-17 08:22:35.810087936	21.000000
50%	2018-09-22 08:43:56.948343552	39.000000
75%	2018-09-28 00:21:23.487930368	110.000000
max	2018-10-05 02:38:49.857748	768.000000
std	NaN	97.457672

	actual_distance_to_destination	osrm_distance	start_scan_to_end_scan \
count	12560.000000	12560.000000	12560.000000
mean	83.047597	105.864573	340.209634
min	9.002461	9.072900	23.000000
25%	21.371169	28.290200	135.000000
50%	39.126693	49.867150	234.000000
75%	110.677576	140.667500	445.000000
max	769.326535	879.382200	1271.000000
std	97.762589	122.336394	286.650425

	segment_actual_time	segment_osrm_time	actual_time	osrm_time	\
count	12560.000000	12560.000000	12560.000000	12560.000000	
mean	199.925080	97.330016	201.728344	88.656290	
min	9.000000	6.000000	9.000000	6.000000	
25%	60.000000	28.000000	61.000000	27.000000	
50%	114.000000	54.000000	116.000000	51.000000	
75%	264.000000	136.000000	268.000000	118.000000	
max	1212.000000	867.000000	1213.000000	641.000000	
std	208.697465	106.831672	210.071577	95.378159	

	segment_osrm_distance	trip_creation_year	trip_creation_month	\
count	12560.000000	12560.0	12560.000000	
mean	112.928154	2018.0	9.115446	
min	9.072900	2018.0	9.000000	
25%	29.320300	2018.0	9.000000	
50%	54.772450	2018.0	9.000000	
75%	150.009625	2018.0	9.000000	
max	1033.678700	2018.0	10.000000	
std	131.105959	0.0	0.319572	

	trip_creation_day	trip_time	triptime_sec
count	12560.000000	12560	12560.000000
mean	18.591879	0 days 04:44:54.496720952	17094.496721
min	1.000000	0 days 00:00:00	0.000000
25%	14.000000	0 days 01:50:43.119529750	6643.119530
50%	20.000000	0 days 03:18:03.833821	11883.833821
75%	25.000000	0 days 06:02:42.184283250	21762.184283
max	30.000000	2 days 01:09:57.136511	176997.136511
std	7.801503	0 days 04:32:06.822854151	16326.822854

Here We can observe that by removing outliers in start_scan_to_end_scan column, other columns max values has been decreased significantly and further dropping columns will lead to loss of valuable data

10 Data Encoding

```
[117]: data_encoding=data.copy()
```

```
[118]: data_encoding.shape
```

```
[118]: (12560, 25)
```

11 Label Encoding

```
[119]: #Here We will use label encoder for encoding route_type column  
le = LabelEncoder()
```

```
[120]: col="route_type"  
data_encoding[col].value_counts()
```

```
[120]: route_type  
Carting      8465  
FTL          4095  
Name: count, dtype: int64
```

```
[121]: data_encoding[col]=le.fit_transform(data_encoding[col])  
data_encoding[col].value_counts()
```

```
[121]: route_type  
0      8465  
1      4095  
Name: count, dtype: int64
```

12 Target Encoding

```
[137]: !pip install category_encoders  
  
from category_encoders import TargetEncoder
```

```
Requirement already satisfied: category_encoders in  
/usr/local/lib/python3.10/dist-packages (2.6.3)  
Requirement already satisfied: numpy>=1.14.0 in /usr/local/lib/python3.10/dist-  
packages (from category_encoders) (1.26.4)  
Requirement already satisfied: scikit-learn>=0.20.0 in  
/usr/local/lib/python3.10/dist-packages (from category_encoders) (1.3.2)  
Requirement already satisfied: scipy>=1.0.0 in /usr/local/lib/python3.10/dist-  
packages (from category_encoders) (1.13.1)  
Requirement already satisfied: statsmodels>=0.9.0 in  
/usr/local/lib/python3.10/dist-packages (from category_encoders) (0.14.3)  
Requirement already satisfied: pandas>=1.0.5 in /usr/local/lib/python3.10/dist-  
packages (from category_encoders) (2.1.4)  
Requirement already satisfied: patsy>=0.5.1 in /usr/local/lib/python3.10/dist-  
packages (from category_encoders) (0.5.6)  
Requirement already satisfied: python-dateutil>=2.8.2 in  
/usr/local/lib/python3.10/dist-packages (from pandas>=1.0.5->category_encoders)  
(2.8.2)  
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-  
packages (from pandas>=1.0.5->category_encoders) (2024.2)
```

Requirement already satisfied: tzdata>=2022.1 in /usr/local/lib/python3.10/dist-packages (from pandas>=1.0.5->category_encoders) (2024.1)
Requirement already satisfied: six in /usr/local/lib/python3.10/dist-packages (from patsy>=0.5.1->category_encoders) (1.16.0)
Requirement already satisfied: joblib>=1.1.1 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=0.20.0->category_encoders) (1.4.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=0.20.0->category_encoders) (3.5.0)
Requirement already satisfied: packaging>=21.3 in /usr/local/lib/python3.10/dist-packages (from statsmodels>=0.9.0->category_encoders) (24.1)

```
[127]: te=TargetEncoder()
```

```
[128]: #Here we will do target encoding for
↳ "source_center", "source_name", "destination_center", "destination_name", "source_state", "desti
↳ columns
columns=["source_center", "source_name", "destination_center", "destination_name", "source_state",
for col in columns:
    data_encoding[col]=te.fit_transform(data_encoding[col],
↳ data_encoding["route_type"])
```

```
[129]: data_encoding.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 12560 entries, 0 to 14786
Data columns (total 25 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   route_type                            12560 non-null  int64
1   trip_uuid                             12560 non-null  object
2   trip_creation_time                    12560 non-null  datetime64[ns]
3   source_center                         12560 non-null  float64
4   source_name                           12560 non-null  float64
5   destination_center                    12560 non-null  float64
6   destination_name                      12560 non-null  float64
7   od_start_time                        12560 non-null  datetime64[ns]
8   od_end_time                          12560 non-null  datetime64[ns]
9   cutoff_factor                        12560 non-null  int64
10  actual_distance_to_destination         12560 non-null  float64
11  osrm_distance                         12560 non-null  float64
12  start_scan_to_end_scan                 12560 non-null  float64
13  segment_actual_time                   12560 non-null  float64
14  segment_osrm_time                     12560 non-null  float64
15  actual_time                           12560 non-null  float64
16  osrm_time                             12560 non-null  float64
17  segment_osrm_distance                 12560 non-null  float64
```

```

18 source_state          12560 non-null float64
19 destination_state     12560 non-null float64
20 trip_creation_year     12560 non-null float64
21 trip_creation_month    12560 non-null float64
22 trip_creation_day      12560 non-null float64
23 trip_time              12560 non-null timedelta64[ns]
24 triptime_sec           12560 non-null float64
dtypes: datetime64[ns](3), float64(18), int64(2), object(1), timedelta64[ns](1)
memory usage: 2.5+ MB

```

```

[130]: data_encoding.
        ↪drop(["trip_uuid","trip_creation_time","od_start_time","od_end_time","trip_time"],axis=1,
        ↪inplace=True)

```

```

[131]: data_encoding.head()

```

```

[131]:
 route_type  source_center  source_name  destination_center  \
0           0      0.224956    0.224956      2.317944e-01
1           0      0.026667    0.026667      4.952686e-08
2           0      0.256210    0.256210      2.836151e-01
3           0      0.018690    0.018690      2.178526e-01
4           0      0.000891    0.000891      4.033957e-06

 destination_name  cutoff_factor  actual_distance_to_destination  \
0      2.317944e-01             72             73.186911
1      4.952686e-08             17             17.175274
2      2.836151e-01             24             24.597048
3      2.178526e-01              9              9.100510
4      4.033957e-06             22             22.424210

 osrm_distance  start_scan_to_end_scan  segment_actual_time  \
0         85.1110                180.0                141.0
1         19.6800                100.0                 59.0
2         28.0647                189.0                 60.0
3         12.0184                 98.0                 24.0
4         28.9203                146.0                 64.0

 segment_osrm_time  actual_time  osrm_time  segment_osrm_distance  \
0              65.0        143.0        68.0              84.1894
1              16.0         59.0        15.0              19.8766
2              23.0         61.0        23.0              28.0647
3              13.0         24.0        13.0              12.0184
4              34.0         64.0        34.0              28.9203

 source_state  destination_state  trip_creation_year  trip_creation_month  \
0      0.126357          0.156993            2018.0              9.0
1      0.198639          0.177807            2018.0              9.0

```

2	0.259448	0.273547	2018.0	9.0
3	0.259448	0.273547	2018.0	9.0
4	0.126357	0.156993	2018.0	9.0

	trip_creation_day	triptime_sec
0	20.0	0.000000
1	20.0	6029.696112
2	20.0	0.000000
3	20.0	5880.338041
4	20.0	10586.899446

13 Standardization

```
[135]: #Here We will use MinMaxScaler method for standardizing dataframe
scaler=MinMaxScaler()
std_data=scaler.fit_transform(data_encoding)
std_data=pd.DataFrame(std_data, columns=data_encoding.columns)
```

```
[136]: std_data
```

```
[136]:
```

	route_type	source_center	source_name	destination_center	\
0	0.0	0.227004	0.227004	2.455594e-01	
1	0.0	0.026909	0.026909	5.246406e-08	
2	0.0	0.258542	0.258542	3.004574e-01	
3	0.0	0.018860	0.018860	2.307896e-01	
4	0.0	0.000899	0.000899	4.273507e-06	
...	
12555	1.0	0.425476	0.425476	8.941138e-01	
12556	1.0	0.635162	0.635162	1.000000e+00	
12557	1.0	0.893970	0.893970	9.463165e-01	
12558	1.0	0.635162	0.635162	6.668095e-01	
12559	1.0	0.417490	0.417490	5.237077e-01	

	destination_name	cutoff_factor	actual_distance_to_destination	\
0	2.455594e-01	0.083004	0.084417	
1	5.246406e-08	0.010540	0.010749	
2	3.004574e-01	0.019763	0.020510	
3	2.307896e-01	0.000000	0.000129	
4	4.273507e-06	0.017128	0.017653	
...	
12555	8.941138e-01	0.176548	0.178597	
12556	1.000000e+00	0.217391	0.220359	
12557	9.463165e-01	0.241107	0.244040	
12558	6.668095e-01	0.098814	0.099617	
12559	5.237077e-01	0.073781	0.075072	

	osrm_distance	start_scan_to_end_scan	segment_actual_time	\
0	0.087369	0.125801	0.109726	
1	0.012188	0.061699	0.041563	
2	0.021822	0.133013	0.042394	
3	0.003384	0.060096	0.012469	
4	0.022805	0.098558	0.045719	
...	
12555	0.209090	0.664263	0.216958	
12556	0.242704	0.524840	0.306733	
12557	0.252936	0.796474	0.694929	
12558	0.107215	0.186699	0.147132	
12559	0.082161	0.264423	0.220283	

	segment_osrm_time	actual_time	osrm_time	segment_osrm_distance	\
0	0.068525	0.111296	0.097638	0.073313	
1	0.011614	0.041528	0.014173	0.010544	
2	0.019744	0.043189	0.026772	0.018536	
3	0.008130	0.012458	0.011024	0.002875	
4	0.032520	0.045681	0.044094	0.019371	
...	
12555	0.150987	0.218439	0.209449	0.159903	
12556	0.218351	0.306478	0.292913	0.198886	
12557	0.199768	0.696013	0.270866	0.218141	
12558	0.099884	0.147841	0.135433	0.096250	
12559	0.070848	0.220930	0.097638	0.069789	

	source_state	destination_state	trip_creation_year	\
0	0.103203	0.089502	0.0	
1	0.182698	0.112002	0.0	
2	0.249577	0.215500	0.0	
3	0.249577	0.215500	0.0	
4	0.103203	0.089502	0.0	
...	
12555	0.418798	0.342118	0.0	
12556	0.460661	0.442377	0.0	
12557	0.997386	0.935294	0.0	
12558	0.460661	0.442377	0.0	
12559	0.103203	0.089502	0.0	

	trip_creation_month	trip_creation_day	triptime_sec
0	0.0	0.655172	0.000000
1	0.0	0.655172	0.034067
2	0.0	0.655172	0.000000
3	0.0	0.655172	0.033223
4	0.0	0.655172	0.059814
...
12555	0.0	0.793103	0.289181

12556	0.0	0.793103	0.230396
12557	0.0	0.793103	0.345594
12558	0.0	0.793103	0.000000
12559	0.0	0.793103	0.000000

[12560 rows x 20 columns]

14 Business Insights

1. On average, osrm_time is lesser than segment_osrm_time
2. On average, osrm_distance is lesser than segment_osrm_distance
3. On average, There is no difference between actual time and segment actual time
4. On average, Actual time is higher than osrm time. While the maximum osrm_time is 400 mins (6.6 hrs), the actual time goes upto 800 mins (13 hrs) which is almost double
5. (9:00am to 12:00pm) and (5:00pm to 10:00pm) have higher delivery time
6. Wednesday is the busiest day of the week with maximum number of trips
7. (10:00pm to 1:00am) is the busiest time of the day having maximum number of trips- probably because the delivery time is least during these hours - less traffic on the roads
8. Carting route type is used for short-distance (0-100km) and short duration (<500 mins) trips while FTLs are used for long-distance (>100km) and long-duration (>300 mins) trips
9. FTL trips are 50% of carting trips in count
10. It can be observed that average speed in inter-state deliveries is much higher than the avg speed in intra-state deliveries
11. Delhi has the lowest intra-state delivery speed while Punjab has the highest

15 Recommendations

1. Since actual time is higher than OSRM time on an average for all trips, the company needs to either improve their forecasting accuracy or identify root cause of delays in deliveries
2. Identify best practices from Maharashtra and Karnataka (states which have the highest volume of deliveries) to increase business in other states
3. To reduce actual_time, dispatch as many deliveries as possible outside of the busy hours
4. Optimise routes along corridors with maximum average speed to shorten delivery time