

Problem statement:

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. Now I have to help their data science team to give insights and recommendations by Clean, sanitize and manipulate data to get useful features out of raw fields of the given dataset. I have to make sense out of the raw data and help the data science team to build forecasting models on it

Business Insights

- * In most of the cases the predicted time is less compared to actual time taken to deliver the product
- * Median time taken after scanning the product at start and scanning at end is almost 449 mins
- * In most of the cases the predicted distance is more than the actual distance
- * Most of the products delivered under 132 mins
- * Most of the products delivered in metropolitan cities like bengaluru, etc
- * Less no of products were delivered in places like union territories

Recommendations

- * Company should optimize their engine to give more accurate predicted distances and times.
- * Company should focus on reaching to ground level in every place.
- * Company should provide some offers for areas where there are less no of products delivered.
- * Company should provide more faster delivery as much as possible.
- * Company should establish multiple dark stores in every area as possible in metropolitan cities.
- * Company should advertise more in areas where there is more rate of customers coming.

In []:

In [140]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import OneHotEncoder
from scipy import stats
import statsmodels.api as sm
from statsmodels.stats.weightstats import ztest
```

In [2]:

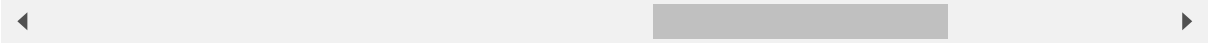
```
data=pd.read_csv('delhivery_data.csv')
```

In [3]:

```
data.head()
```

Out[3]:

imp	actual_distance_to_destination	actual_time	osrm_time	osrm_distance	factor	segment_ac
3-20 7:55	10.435660	14.0	11.0	11.9653	1.272727	
3-20 7:55	18.936842	24.0	20.0	21.7243	1.200000	
3-20 586	27.637279	40.0	28.0	32.5395	1.428571	
3-20 3:57	36.118028	62.0	40.0	45.5620	1.550000	
3-20 3:55	39.386040	68.0	44.0	54.2181	1.545455	



In []:

In [4]:

```
data.shape
```

Out[4]:

(144867, 24)

In [5]:

```
data.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
```

In [6]:

```
data.dtypes
```

Out[6]:

```
data                object
trip_creation_time  object
route_schedule_uuid object
route_type          object
trip_uuid           object
source_center       object
source_name         object
destination_center  object
destination_name    object
od_start_time       object
od_end_time         object
start_scan_to_end_scan float64
is_cutoff           bool
cutoff_factor       int64
cutoff_timestamp    object
actual_distance_to_destination float64
actual_time         float64
osrm_time           float64
osrm_distance       float64
factor              float64
segment_actual_time float64
segment_osrm_time   float64
segment_osrm_distance float64
segment_factor      float64
dtype: object
```

In [7]:

```
data.describe()
```

Out[7]:

actual_time	osrm_time	osrm_distance	factor	segment_actual_time	segment_osrm
144867.000000	144867.000000	144867.000000	144867.000000	144867.000000	144867.
416.927527	213.868272	284.771297	2.120107	36.196111	18.
598.103621	308.011085	421.119294	1.715421	53.571158	14.
9.000000	6.000000	9.008200	0.144000	-244.000000	0.
51.000000	27.000000	29.914700	1.604264	20.000000	11.
132.000000	64.000000	78.525800	1.857143	29.000000	17.
513.000000	257.000000	343.193250	2.213483	40.000000	22.
4532.000000	1686.000000	2326.199100	77.387097	3051.000000	1611.

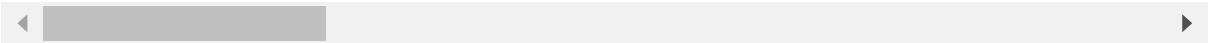
In [8]:

```
data.describe(include='all')
```

Out[8]:

	data	trip_creation_time	route_schedule_uuid	route_type	trip_uuid	sc
count	144867	144867	144867	144867	144867	
unique	2	14817	1504	2	14817	
top	training	2018-09-19 04:07:34.091798	thanos::sroute:4029a8a2- 6c74-4b7e-a6d8- f9e069f...	FTL	153715938946690081	IN
freq	104858	101	1812	99660	101	
mean	NaN	NaN	NaN	NaN	NaN	
std	NaN	NaN	NaN	NaN	NaN	
min	NaN	NaN	NaN	NaN	NaN	
25%	NaN	NaN	NaN	NaN	NaN	
50%	NaN	NaN	NaN	NaN	NaN	
75%	NaN	NaN	NaN	NaN	NaN	
max	NaN	NaN	NaN	NaN	NaN	

11 rows × 24 columns



Statistical Summary

*Here median start_scan_to_end_scan time is less than 449 but there are some orders which took 7898 mins.
* Median actual time taken is under 52 but the median estimated time is 27.

In []:

In [9]:

```
data.isna().sum()
```

```
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
```

There are 293 and 261 missing values in source_name,destination_name respectively.

In [10]:

```
data.nunique()
```

Out[10]:

```
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
```

Here data,route_type,is_cutoff are categorical type

conversion of categorical attributes to 'category'

In [11]:

```
data['data']=data['data'].astype('category')
data['route_type']=data['route_type'].astype('category')
data['is_cutoff']=data['is_cutoff'].astype('category')
```

In [12]:

```
data.dtypes
```

Out[12]:

data	category
trip_creation_time	object
route_schedule_uuid	object
route_type	category
trip_uuid	object
source_center	object
source_name	object
destination_center	object
destination_name	object
od_start_time	object
od_end_time	object
start_scan_to_end_scan	float64
is_cutoff	category
cutoff_factor	int64
cutoff_timestamp	object
actual_distance_to_destination	float64
actual_time	float64
osrm_time	float64
osrm_distance	float64
factor	float64
segment_actual_time	float64
segment_osrm_time	float64
segment_osrm_distance	float64
segment_factor	float64
dtype:	object

- Lets check the valuecounts of categorical variables

In [13]:

```
d=data.select_dtypes(include='category')
a=list(d.columns)
a
```

Out[13]:

```
['data', 'route_type', 'is_cutoff']
```

In [14]:

```
for i in a:  
    print(i)  
    print(data[i].value_counts())  
    print('\n')
```

```
data  
training    104858  
test        40009  
Name: data, dtype: int64
```

```
route_type  
FTL          99660  
Carting      45207  
Name: route_type, dtype: int64
```

```
is_cutoff  
True         118749  
False        26118  
Name: is_cutoff, dtype: int64
```

Visual Analysis

Univariate Analysis of Continuous variables

In [15]:

```
con=list((data.select_dtypes(include=['int64','float64'])))  
con
```

Out[15]:

```
['start_scan_to_end_scan',  
 'cutoff_factor',  
 'actual_distance_to_destination',  
 'actual_time',  
 'osrm_time',  
 'osrm_distance',  
 'factor',  
 'segment_actual_time',  
 'segment_osrm_time',  
 'segment_osrm_distance',  
 'segment_factor']
```


In [16]:

```
#start_scan_to_end_scan
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

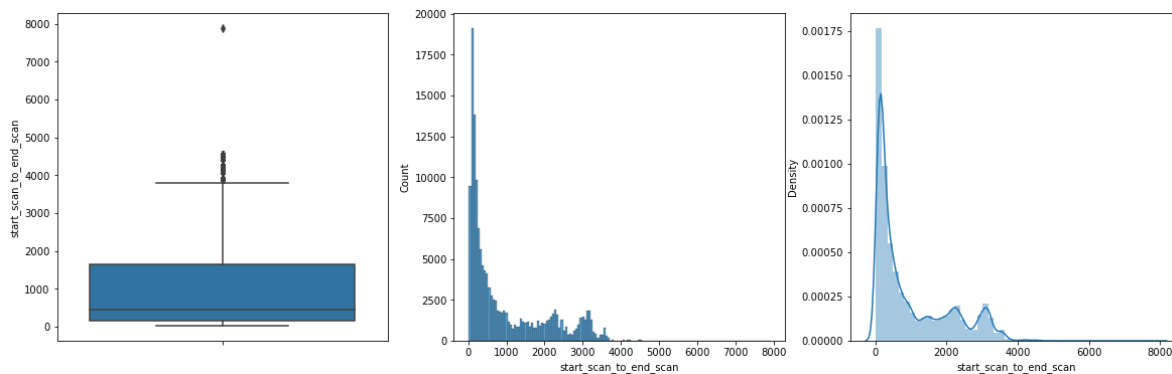
sns.boxplot(data=data,y="start_scan_to_end_scan")
plt.subplot(1,3,2)
sns.histplot(data=data,x="start_scan_to_end_scan")
plt.subplot(1,3,3)
sns.distplot(a=data['start_scan_to_end_scan'],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

```
warnings.warn(msg, FutureWarning)
```

Out[16]:

<AxesSubplot:xlabel='start_scan_to_end_scan', ylabel='Density'>



In [17]:

```
#cutoff_factor
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

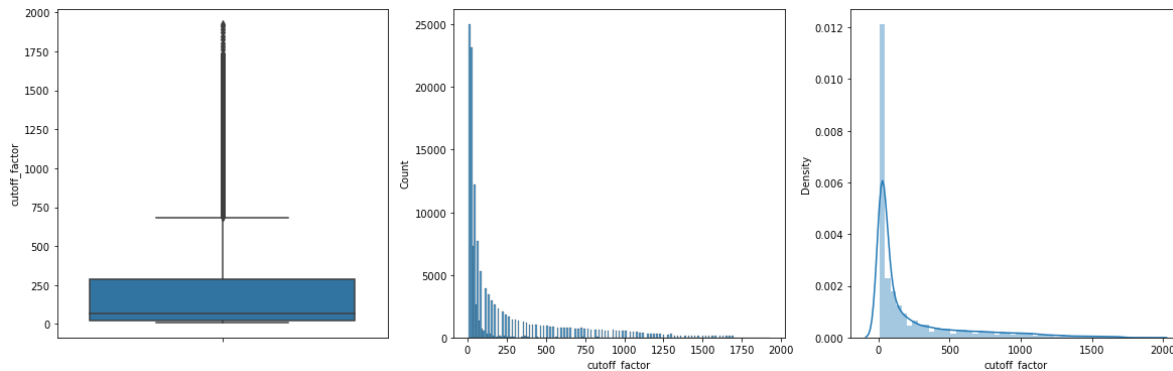
sns.boxplot(data=data,y="cutoff_factor")
plt.subplot(1,3,2)
sns.histplot(data=data,x="cutoff_factor")
plt.subplot(1,3,3)
sns.distplot(a=data['cutoff_factor'],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[17]:

<AxesSubplot:xlabel='cutoff_factor', ylabel='Density'>



In [18]:

```
#actual_distance_to_destination
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

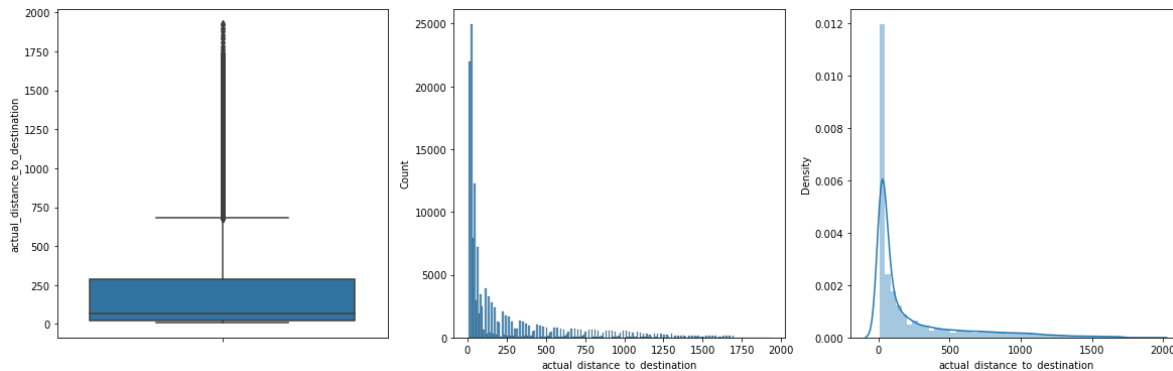
sns.boxplot(data=data,y="actual_distance_to_destination")
plt.subplot(1,3,2)
sns.histplot(data=data,x="actual_distance_to_destination")
plt.subplot(1,3,3)
sns.distplot(a=data['actual_distance_to_destination'],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[18]:

<AxesSubplot:xlabel='actual_distance_to_destination', ylabel='Density'>



In [19]:

```
#actual_time
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

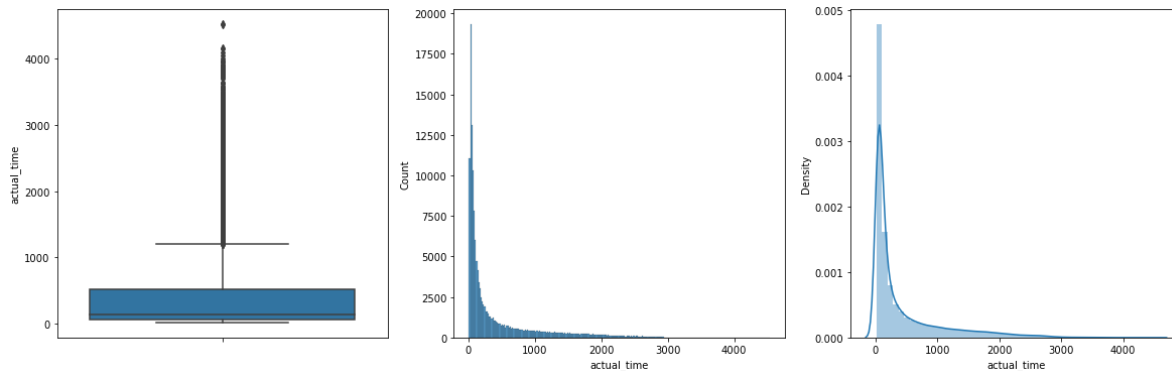
sns.boxplot(data=data,y="actual_time")
plt.subplot(1,3,2)
sns.histplot(data=data,x="actual_time")
plt.subplot(1,3,3)
sns.distplot(a=data['actual_time'],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[19]:

<AxesSubplot:xlabel='actual_time', ylabel='Density'>



In [20]:

```
#osrm_time
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

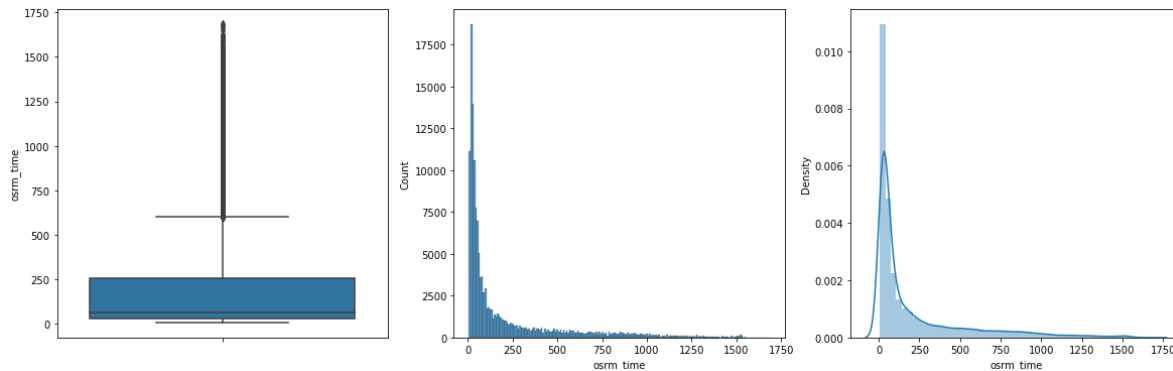
sns.boxplot(data=data,y="osrm_time")
plt.subplot(1,3,2)
sns.histplot(data=data,x="osrm_time")
plt.subplot(1,3,3)
sns.distplot(a=data['osrm_time'],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[20]:

<AxesSubplot:xlabel='osrm_time', ylabel='Density'>



In [21]:

```
#osrm_distance
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

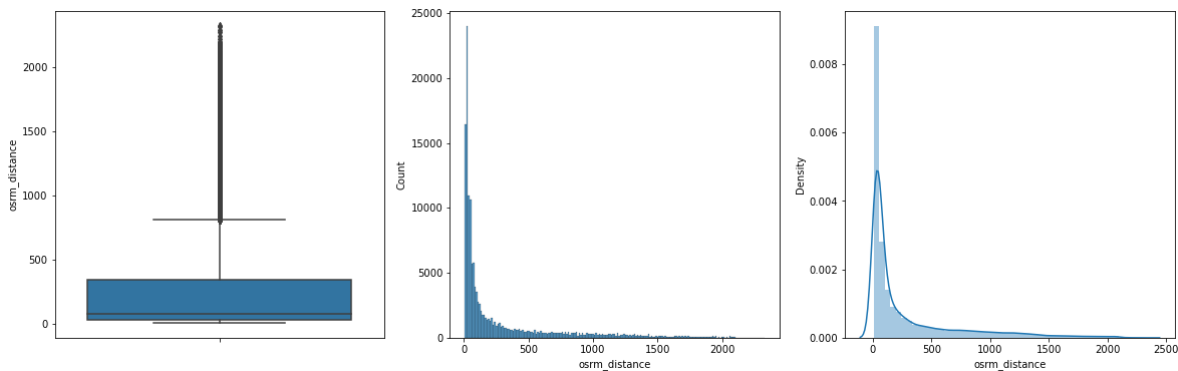
sns.boxplot(data=data,y="osrm_distance")
plt.subplot(1,3,2)
sns.histplot(data=data,x="osrm_distance")
plt.subplot(1,3,3)
sns.distplot(a=data['osrm_distance'],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[21]:

<AxesSubplot:xlabel='osrm_distance', ylabel='Density'>



In [22]:

```
#factor
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

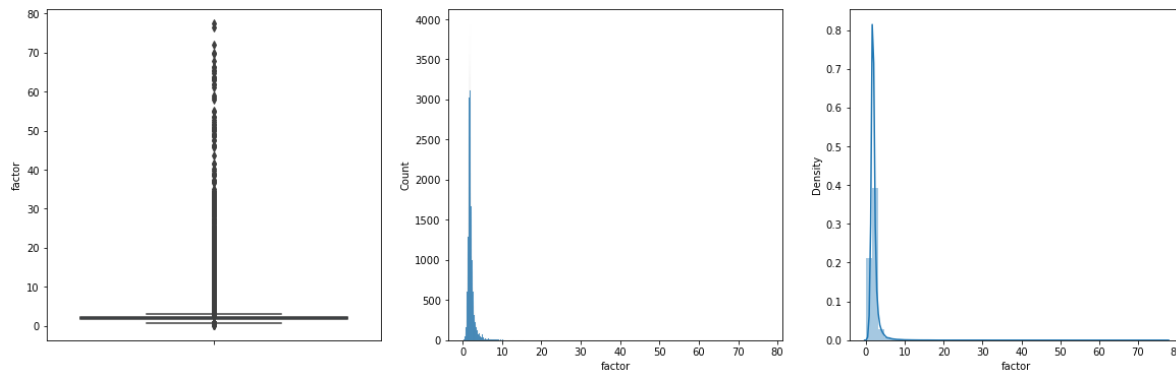
sns.boxplot(data=data,y="factor")
plt.subplot(1,3,2)
sns.histplot(data=data,x="factor")
plt.subplot(1,3,3)
sns.distplot(a=data[ 'factor' ],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[22]:

<AxesSubplot:xlabel='factor', ylabel='Density'>



In [23]:

```

#segment_actual_time
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

sns.boxplot(data=data,y="segment_actual_time")
plt.subplot(1,3,2)
sns.histplot(data=data,x="segment_actual_time")
plt.subplot(1,3,3)
sns.distplot(a=data['segment_actual_time'],kde=True)

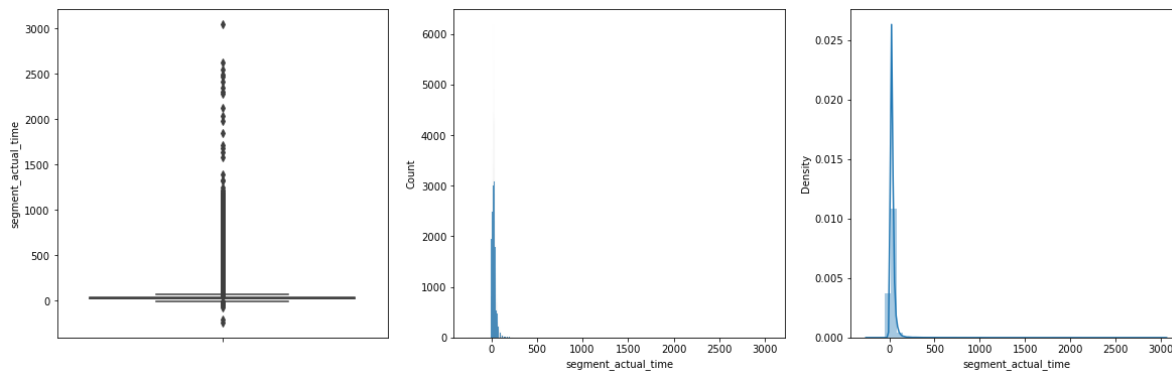
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[23]:

<AxesSubplot:xlabel='segment_actual_time', ylabel='Density'>



In [24]:

```
#segment_osrm_time
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

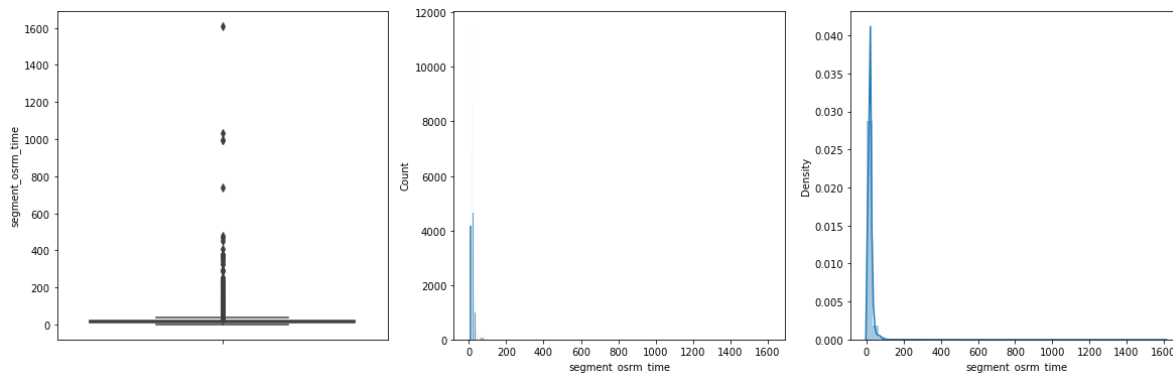
sns.boxplot(data=data,y="segment_osrm_time")
plt.subplot(1,3,2)
sns.histplot(data=data,x="segment_osrm_time")
plt.subplot(1,3,3)
sns.distplot(a=data['segment_osrm_time'],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[24]:

<AxesSubplot:xlabel='segment_osrm_time', ylabel='Density'>



In [25]:

```
#segment_osrm_distance
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

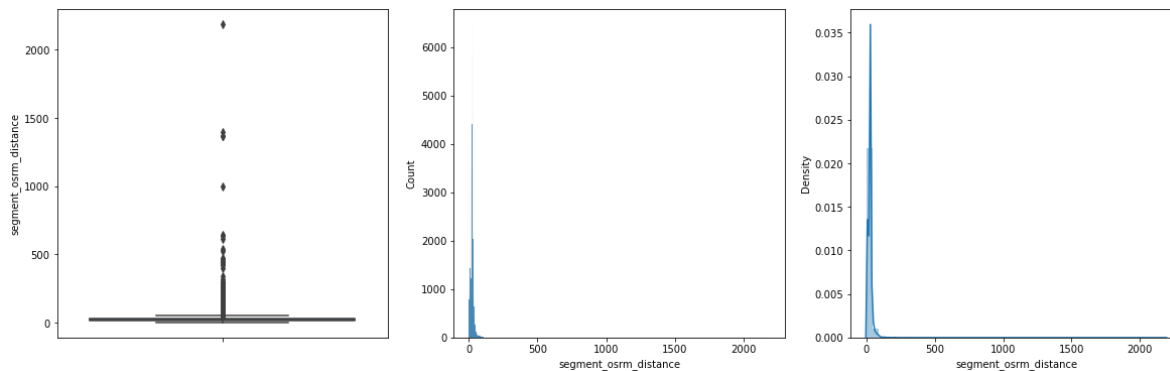
sns.boxplot(data=data,y="segment_osrm_distance")
plt.subplot(1,3,2)
sns.histplot(data=data,x="segment_osrm_distance")
plt.subplot(1,3,3)
sns.distplot(a=data['segment_osrm_distance'],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[25]:

<AxesSubplot:xlabel='segment_osrm_distance', ylabel='Density'>



In [26]:

```
#segment_factor
plt.figure(figsize=(20,6))

plt.subplot(1,3,1)

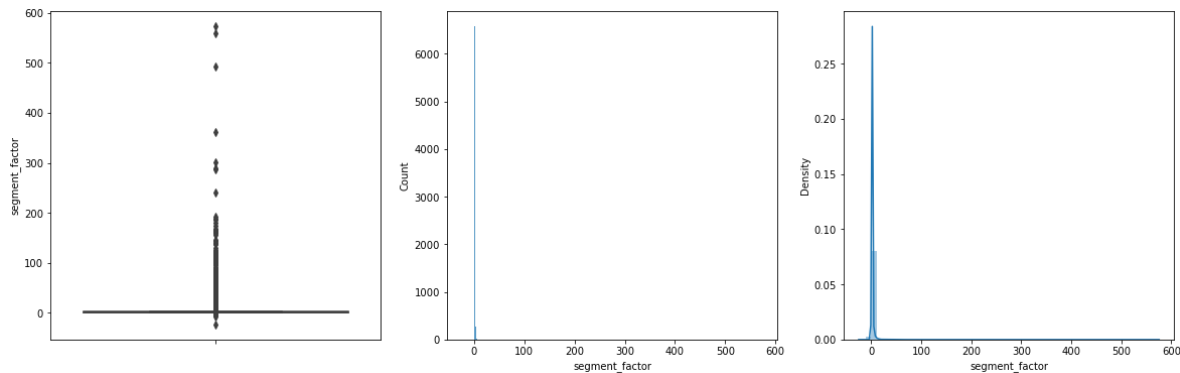
sns.boxplot(data=data,y="segment_factor")
plt.subplot(1,3,2)
sns.histplot(data=data,x="segment_factor")
plt.subplot(1,3,3)
sns.distplot(a=data['segment_factor'],kde=True)
```

C:\Users\Ajith\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[26]:

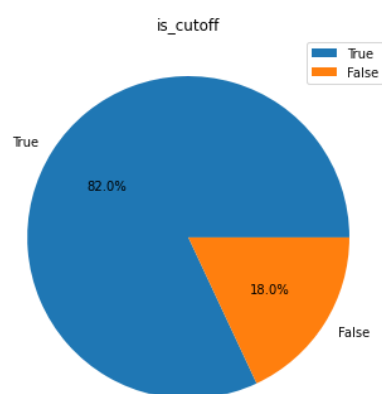
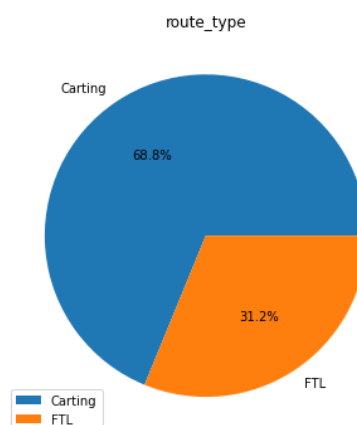
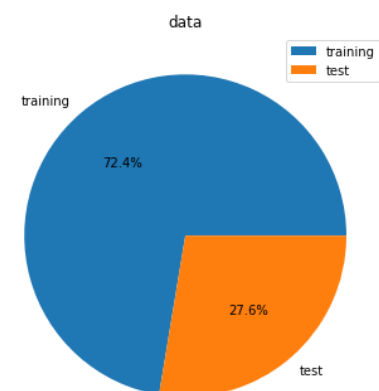
<AxesSubplot:xlabel='segment_factor', ylabel='Density'>



Univariate analysis of categorical variables

In [27]:

```
plt.figure(figsize = (20,20))
cols = ['data', 'route_type', 'is_cutoff']
k = 0
for i in cols:
    plt.subplot(321 + k)
    plt.pie(data[i].value_counts(), autopct = '%1.1f%%',
            labels = data[i].unique())
    plt.title(i)
    plt.legend()
    k += 1
plt.show()
```



Bivariate Analysis

In [28]:

```
con
```

Out[28]:

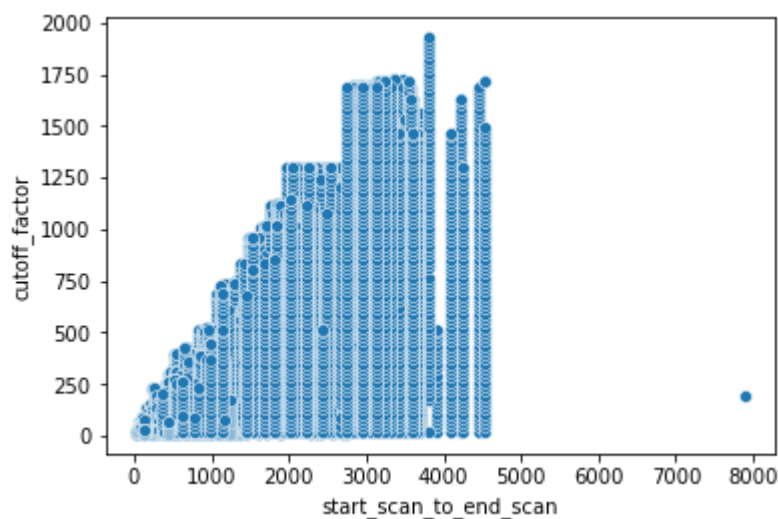
```
['start_scan_to_end_scan',  
'cutoff_factor',  
'actual_distance_to_destination',  
'actual_time',  
'osrm_time',  
'osrm_distance',  
'factor',  
'segment_actual_time',  
'segment_osrm_time',  
'segment_osrm_distance',  
'segment_factor']
```

In [29]:

```
sns.scatterplot(x='start_scan_to_end_scan',y='cutoff_factor',data=data)
```

Out[29]:

<AxesSubplot:xlabel='start_scan_to_end_scan', ylabel='cutoff_factor'>

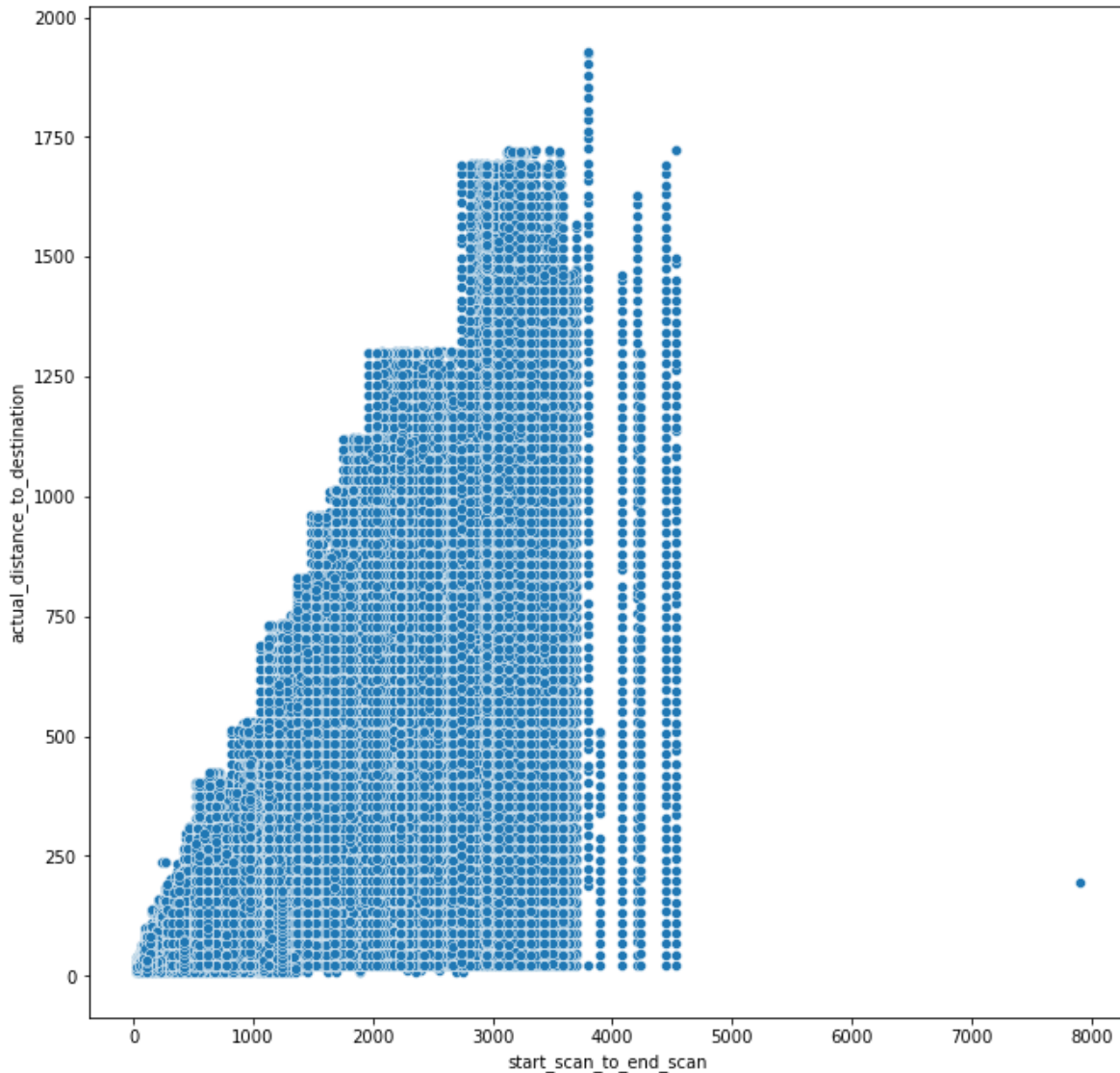


In [30]:

```
plt.figure(figsize=(11,11))  
sns.scatterplot(x='start_scan_to_end_scan',y='actual_distance_to_destination',data=data)
```

Out[30]:

<AxesSubplot:xlabel='start_scan_to_end_scan', ylabel='actual_distance_to_destination'>

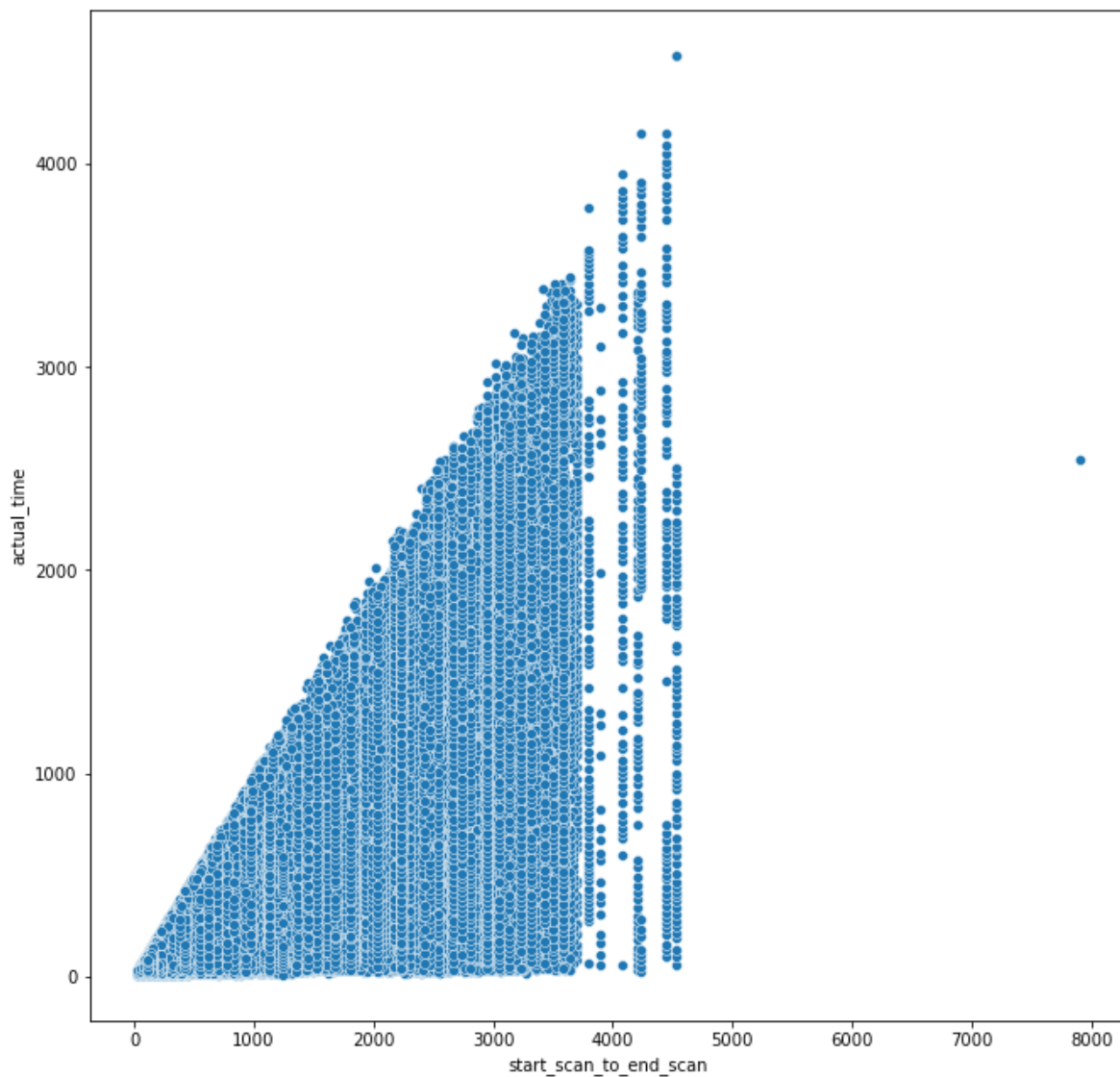


In [31]:

```
plt.figure(figsize=(11,11))  
sns.scatterplot(x='start_scan_to_end_scan',y='actual_time',data=data)
```

Out[31]:

<AxesSubplot:xlabel='start_scan_to_end_scan', ylabel='actual_time'>

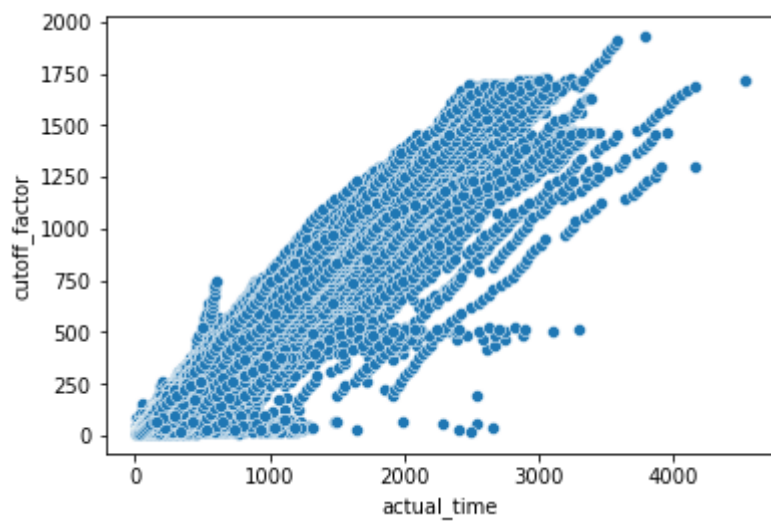


In [32]:

```
sns.scatterplot(x='actual_time',y='cutoff_factor',data=data)
```

Out[32]:

```
<AxesSubplot:xlabel='actual_time', ylabel='cutoff_factor'>
```

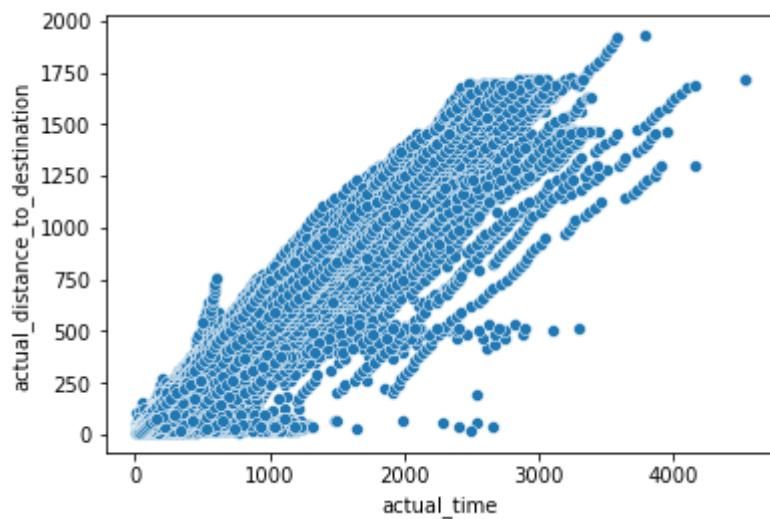


In [33]:

```
sns.scatterplot(x='actual_time',y='actual_distance_to_destination',data=data)
```

Out[33]:

<AxesSubplot:xlabel='actual_time', ylabel='actual_distance_to_destination'>

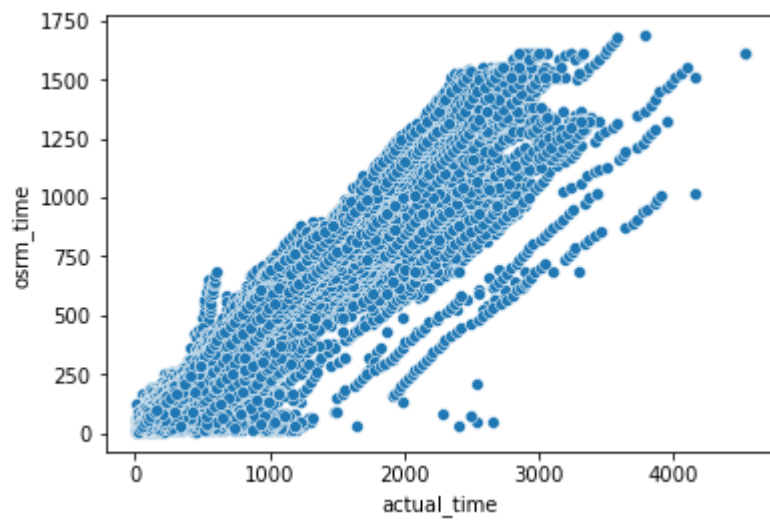


In [34]:

```
sns.scatterplot(x='actual_time',y='osrm_time',data=data)
```

Out[34]:

<AxesSubplot:xlabel='actual_time', ylabel='osrm_time'>

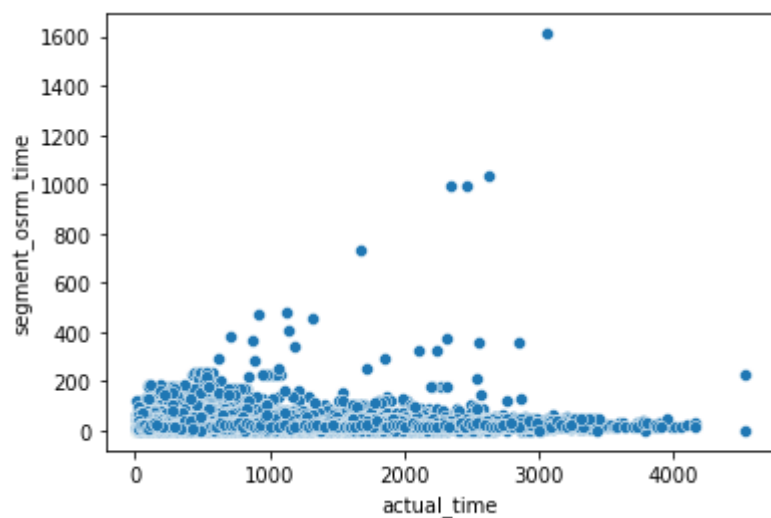


In [35]:

```
sns.scatterplot(x='actual_time',y='segment_osrm_time',data=data)
```

Out[35]:

<AxesSubplot:xlabel='actual_time', ylabel='segment_osrm_time'>

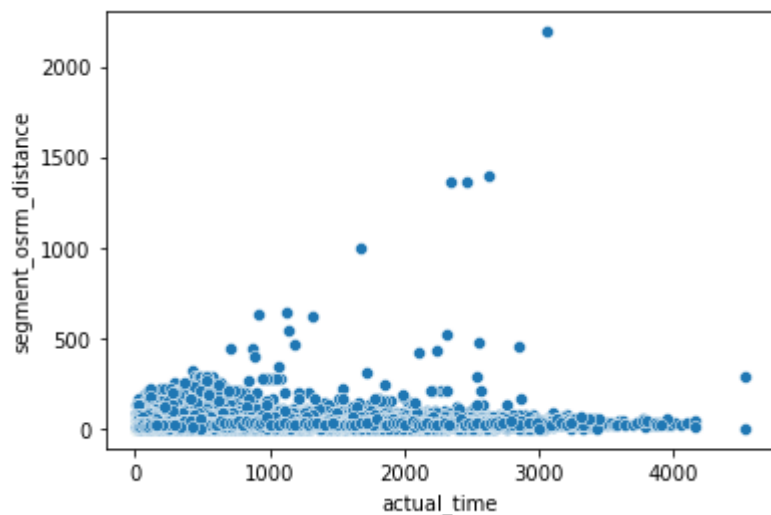


In [36]:

```
sns.scatterplot(x='actual_time',y='segment_osrm_distance',data=data)
```

Out[36]:

<AxesSubplot:xlabel='actual_time', ylabel='segment_osrm_distance'>



In [37]:

```
plt.figure(figsize=(11,11))
sns.heatmap(data.corr(),annot=True)
```

Out[37]:

<AxesSubplot:>



Observations

In [38]:

```
data['od_start_time']
```

Out[38]:

```
0      2018-09-20 03:21:32.418600
1      2018-09-20 03:21:32.418600
2      2018-09-20 03:21:32.418600
3      2018-09-20 03:21:32.418600
4      2018-09-20 03:21:32.418600
...
144862 2018-09-20 16:24:28.436231
144863 2018-09-20 16:24:28.436231
144864 2018-09-20 16:24:28.436231
144865 2018-09-20 16:24:28.436231
144866 2018-09-20 16:24:28.436231
Name: od_start_time, Length: 144867, dtype: object
```

In [39]:

```
data['od_start_time']=pd.to_datetime(data['od_start_time'])
```

In [40]:

```
data['od_start_hour']=data['od_start_time'].dt.hour
data['od_start_min']=data['od_start_time'].dt.minute
```

In [41]:

```
data['od_start_hour']+=data['od_start_min']/60
```

In [42]:

```
data['od_start_hour']
```

Out[42]:

```
0      3.35
1      3.35
2      3.35
3      3.35
4      3.35
...
144862 16.40
144863 16.40
144864 16.40
144865 16.40
144866 16.40
Name: od_start_hour, Length: 144867, dtype: float64
```

In [43]:

```
data.drop(columns='od_start_min',axis=1,inplace=True)
```

In [44]:

```
data['od_end_time']
```

Out[44]:

```
0      2018-09-20 04:47:45.236797
1      2018-09-20 04:47:45.236797
2      2018-09-20 04:47:45.236797
3      2018-09-20 04:47:45.236797
4      2018-09-20 04:47:45.236797
...
144862 2018-09-20 23:32:09.618069
144863 2018-09-20 23:32:09.618069
144864 2018-09-20 23:32:09.618069
144865 2018-09-20 23:32:09.618069
144866 2018-09-20 23:32:09.618069
Name: od_end_time, Length: 144867, dtype: object
```

In [45]:

```
data['od_end_time']=pd.to_datetime(data['od_end_time'])
```

In [46]:

```
data['od_end_hour']=data['od_end_time'].dt.hour
data['od_end_min']=data['od_end_time'].dt.minute
data['od_end_hour']+=data['od_end_min']/60
```

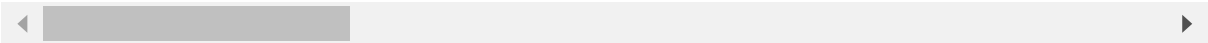
In [47]:

```
data.drop(columns='od_end_min',axis=1,inplace=True)
data
```

Out[47]:

	data	trip_creation_time	route_schedule_uuid	route_type	trip_uuid	so
0	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
1	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
2	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
3	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
4	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
...
144862	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144863	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144864	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144865	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144866	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC

144867 rows × 26 columns



In [48]:

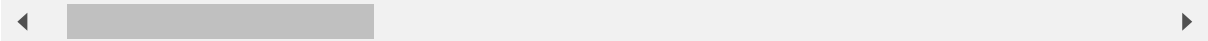
```
data
```

Out[48]:

	data	trip_creation_time	route_schedule_uuid	route_type	trip_uuid	source_cen
1	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	IND388121A
	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	IND388121A
1	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	IND388121A
1	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	IND388121A
1	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	IND388121A

1	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	IND131028A
1	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	IND131028A
1	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	IND131028A
1	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	IND131028A
1	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	IND131028A

rows × 26 columns



In [49]:

```
data['actual_time'].nunique()
```

Out[49]:

3182

In [50]:

```
x=data.groupby(by=['trip_uuid','source_center','destination_center'])
```

In [51]:

```
# Here we dont know what are is_cutoff , cutoff_factor, cutoff_timestamp and segment_factor d
data.drop(columns=['is_cutoff' , 'cutoff_factor', 'cutoff_timestamp', 'segment_factor'],inplace=True)
```

Feature Creation

Merging of rows and aggregation of fields

In [52]:

```
data['agg_actual_time']=data.groupby(by=['trip_uuid','source_center','destination_center'])
data['agg_segment_actual_time']=data.groupby(by=['trip_uuid','source_center','destination_center'])
data['agg_osrm_time']=data.groupby(by=['trip_uuid','source_center','destination_center'])
data['agg_segment_osrm_time']=data.groupby(by=['trip_uuid','source_center','destination_center'])
data['agg_osrm_distance']=data.groupby(by=['trip_uuid','source_center','destination_center'])
data['agg_segment_osrm_distance']=data.groupby(by=['trip_uuid','source_center','destination_center'])
```

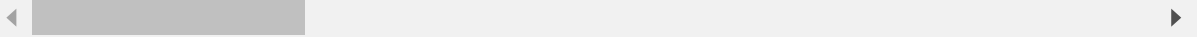

In [53]:

data

Out[53]:

	data	trip_creation_time	route_schedule_uuid	route_type	trip_uuid	so
0	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
1	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
2	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
3	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
4	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
...
144862	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144863	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144864	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144865	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144866	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC

144867 rows × 28 columns



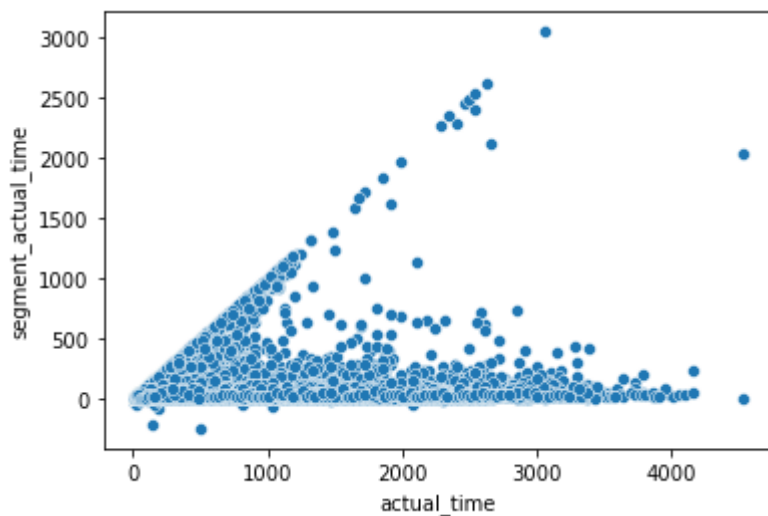
Comparison & Visualization of time and distance fields

In [54]:

```
sns.scatterplot(x='actual_time',y='segment_actual_time',data=data)
```

Out[54]:

<AxesSubplot:xlabel='actual_time', ylabel='segment_actual_time'>

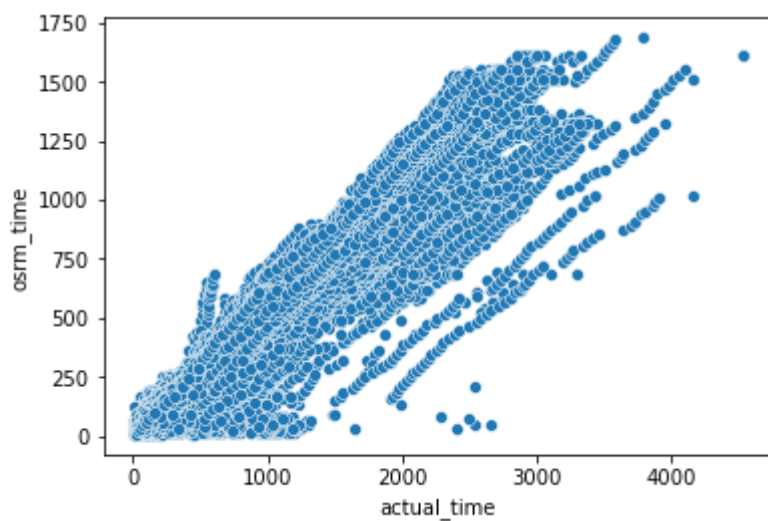


In [55]:

```
sns.scatterplot(x='actual_time',y='osrm_time',data=data)
```

Out[55]:

<AxesSubplot:xlabel='actual_time', ylabel='osrm_time'>

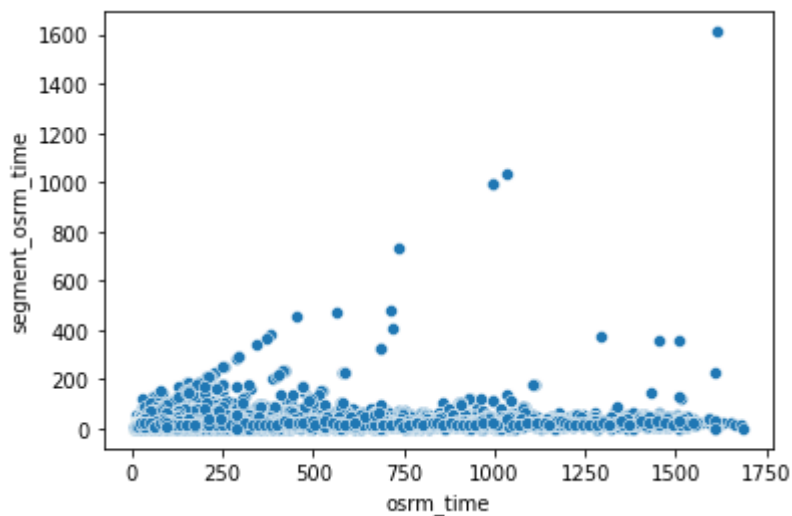


In [56]:

```
sns.scatterplot(x='osrm_time',y='segment_osrm_time',data=data)
```

Out[56]:

<AxesSubplot:xlabel='osrm_time', ylabel='segment_osrm_time'>

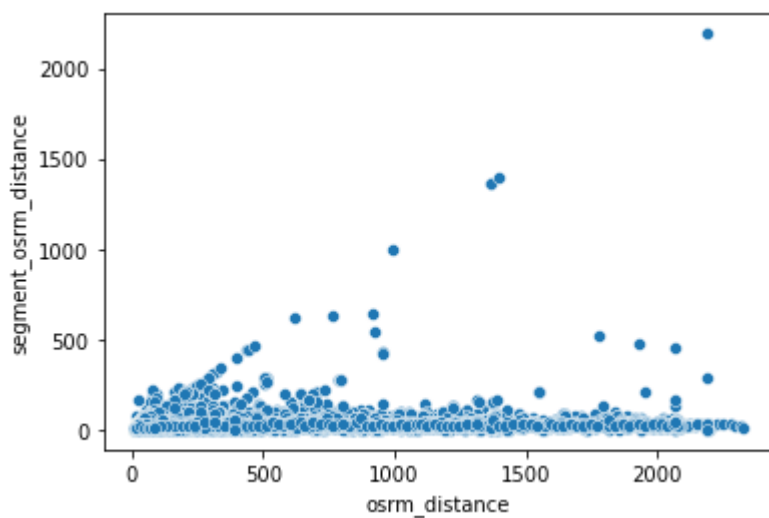


In [57]:

```
sns.scatterplot(x='osrm_distance',y='segment_osrm_distance',data=data)
```

Out[57]:

<AxesSubplot:xlabel='osrm_distance', ylabel='segment_osrm_distance'>

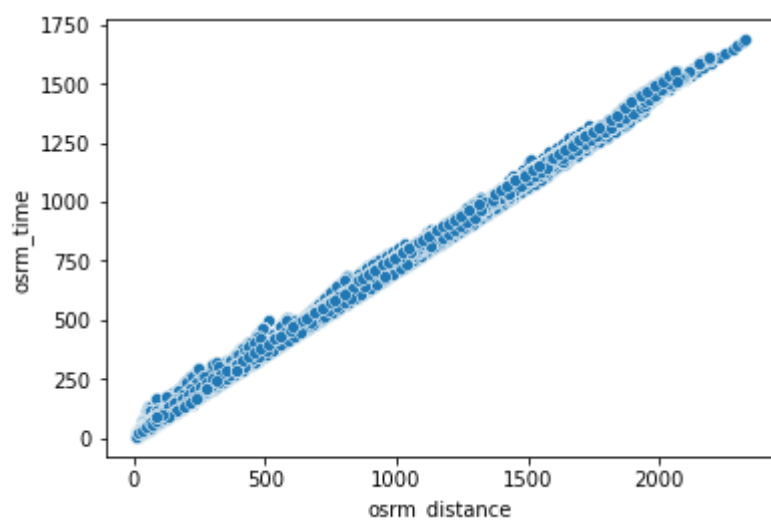


In [58]:

```
sns.scatterplot(x='osrm_distance',y='osrm_time',data=data)
```

Out[58]:

<AxesSubplot:xlabel='osrm_distance', ylabel='osrm_time'>

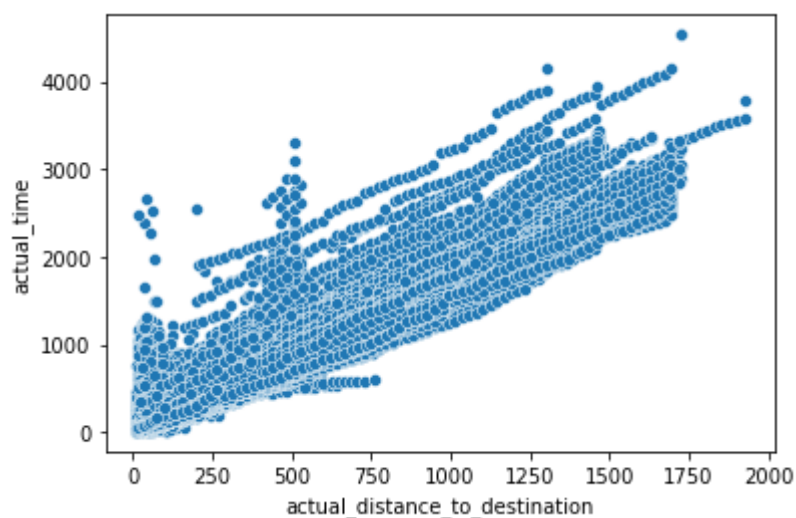


In [59]:

```
sns.scatterplot(x='actual_distance_to_destination',y='actual_time',data=data)
```

Out[59]:

<AxesSubplot:xlabel='actual_distance_to_destination', ylabel='actual_time'>



Observations:

*The actual time and the estimated time are proportional to each other
 * The graphs between actual distance and time is more deviated than the graph between osrm_distance and time.

In [60]:

```
data['destination_name'].unique()
```

Out[60]:

```
array(['Khambhat_MotvdDPP_D (Gujarat)', 'Anand_Vaghasi_IP (Gujarat)',  
      'Pune_Tathawde_H (Maharashtra)', ...,  
      'Chennai_Mylapore (Tamil Nadu)', 'Naraingarh_Ward2DPP_D (Haryana)',  
      'Mumbai_Ghansoli_DC (Maharashtra)'], dtype=object)
```

In [61]:

```
data['source_name'].unique()
```

Out[61]:

```
array(['Anand_VUNagar_DC (Gujarat)', 'Khambhat_MotvdDPP_D (Gujarat)',  
      'Bhiwandi_Mankoli_HB (Maharashtra)', ...,  
      'Dwarka_StnRoad_DC (Gujarat)', 'Bengaluru_Nelmngla_L (Karnataka)',  
      'Kulithalai_AnnaNGR_D (Tamil Nadu)'], dtype=object)
```

In [62]:

```
data['source_name'].str.split('_')
```

Out[62]:

```
0      [Anand, VUNagar, DC (Gujarat)]  
1      [Anand, VUNagar, DC (Gujarat)]  
2      [Anand, VUNagar, DC (Gujarat)]  
3      [Anand, VUNagar, DC (Gujarat)]  
4      [Anand, VUNagar, DC (Gujarat)]  
...  
144862 [Sonipat, Kundli, H (Haryana)]  
144863 [Sonipat, Kundli, H (Haryana)]  
144864 [Sonipat, Kundli, H (Haryana)]  
144865 [Sonipat, Kundli, H (Haryana)]  
144866 [Sonipat, Kundli, H (Haryana)]  
Name: source_name, Length: 144867, dtype: object
```

In [63]:

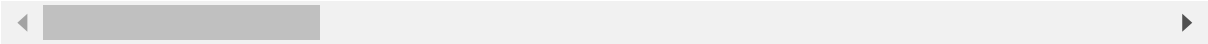
```
data['source_city']=data['source_name'].apply(lambda x:str(x).split('_')[0])

data
```

Out[63]:

	data	trip_creation_time	route_schedule_uuid	route_type	trip_uuid	so
0	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
1	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
2	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
3	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
4	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78- b351-4c0e-a951- fa3d5c3...	Carting	153741093647649320	INC
...
144862	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144863	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144864	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144865	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC
144866	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f- 4e20-4c31-8542- 67b86d5...	Carting	153746066843555182	INC

144867 rows × 29 columns

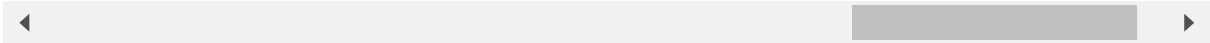


In [85]:

```
data['source_place']=data['source_name'].apply(lambda x:str(x).split('_')[1:2])
data['source_place']=data['source_place'].apply(lambda x:str(x)[1:-1])
data['source_code']=data['source_name'].apply(lambda x:str(x).split('_')[2:3])
data
```

Out[85]:

gment_osrm_time	agg_osrm_distance	agg_segment_osrm_distance	source_city	source_place	soi
11.0	11.9653	11.9653	Anand	'VUNagar'	[DC
20.0	33.6896	21.7243	Anand	'VUNagar'	[DC
27.0	66.2291	32.5395	Anand	'VUNagar'	[DC
39.0	111.7911	45.5619	Anand	'VUNagar'	[DC
44.0	166.0092	49.4772	Anand	'VUNagar'	[DC
...
94.0	228.5453	65.3487	Sonipat	'Kundli'	[H
115.0	314.2282	82.7212	Sonipat	'Kundli'	[H
149.0	411.3215	103.4265	Sonipat	'Kundli'	[H
176.0	522.5924	122.3150	Sonipat	'Kundli'	[H
185.0	611.3243	131.1238	Sonipat	'Kundli'	[H

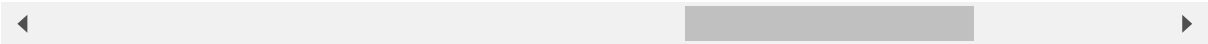


In [86]:

```
data['destination_city']=data['destination_name'].apply(lambda x:str(x).split('_')[0])
data['destination_place']=data['destination_name'].apply(lambda x:str(x).split('_')[1:2])
data['destination_place']=data['destination_place'].apply(lambda x:str(x)[1:-1])
data['destination_code']=data['destination_name'].apply(lambda x:str(x).split('_')[2:3])
data
```

Out[86]:

segment_osrm_time	agg_osrm_distance	agg_segment_osrm_distance	source_city	source_place	sc
11.0	11.9653	11.9653	Anand	'VUNagar'	[D
20.0	33.6896	21.7243	Anand	'VUNagar'	[D
27.0	66.2291	32.5395	Anand	'VUNagar'	[D
39.0	111.7911	45.5619	Anand	'VUNagar'	[D
44.0	166.0092	49.4772	Anand	'VUNagar'	[D
...	
94.0	228.5453	65.3487	Sonipat	'Kundli'	[F
115.0	314.2282	82.7212	Sonipat	'Kundli'	[F
149.0	411.3215	103.4265	Sonipat	'Kundli'	[F
176.0	522.5924	122.3150	Sonipat	'Kundli'	[F
185.0	611.3243	131.1238	Sonipat	'Kundli'	[F



In [87]:

```
data['trip_creation_time']
```

Out[87]:

```
0      2018-09-20 02:35:36.476840
1      2018-09-20 02:35:36.476840
2      2018-09-20 02:35:36.476840
3      2018-09-20 02:35:36.476840
4      2018-09-20 02:35:36.476840
```

```
...
```

```
144862 2018-09-20 16:24:28.436231
144863 2018-09-20 16:24:28.436231
144864 2018-09-20 16:24:28.436231
144865 2018-09-20 16:24:28.436231
144866 2018-09-20 16:24:28.436231
```

Name: trip_creation_time, Length: 144867, dtype: object

In [88]:

```
data['trip_creation_time']=pd.to_datetime(data['trip_creation_time'])
```

In [89]:

```
data['trip_creation_time_month']=data['trip_creation_time'].dt.month_name()
data['trip_creation_time_year']=data['trip_creation_time'].dt.year
data['trip_creation_time_day']=data['trip_creation_time'].dt.day_name()
```

In [90]:

```
data
```

Out[90]:

	data	trip_creation_time	route_schedule_uuid	route_type	trip_uuid	so
0	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	INC
1	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	INC
2	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	INC
3	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	INC
4	training	2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	INC
...
144862	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	INC
144863	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	INC
144864	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	INC
144865	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	INC
144866	training	2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	INC

144867 rows × 7 columns

Calculate the time taken between od_start_time and od_end_time and keep it as a feature. Drop the original columns, if required

In [91]:

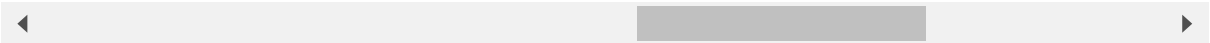
```
data['diff_od_time_mins']=((data['od_end_time']-data['od_start_time']).astype('timedelta64[
```

In [92]:

```
data
```

Out[92]:

city	source_place	source_code	destination_city	destination_place	destination_code	trip_creatio
land	'VUNagar'	[DC (Gujarat)]	Khambhat	'MotvdDPP'	[D (Gujarat)]	
land	'VUNagar'	[DC (Gujarat)]	Khambhat	'MotvdDPP'	[D (Gujarat)]	
land	'VUNagar'	[DC (Gujarat)]	Khambhat	'MotvdDPP'	[D (Gujarat)]	
land	'VUNagar'	[DC (Gujarat)]	Khambhat	'MotvdDPP'	[D (Gujarat)]	
land	'VUNagar'	[DC (Gujarat)]	Khambhat	'MotvdDPP'	[D (Gujarat)]	
...	
ipat	'Kundli'	[H (Haryana)]	Gurgaon	'Bilaspur'	[HB (Haryana)]	
ipat	'Kundli'	[H (Haryana)]	Gurgaon	'Bilaspur'	[HB (Haryana)]	
ipat	'Kundli'	[H (Haryana)]	Gurgaon	'Bilaspur'	[HB (Haryana)]	
ipat	'Kundli'	[H (Haryana)]	Gurgaon	'Bilaspur'	[HB (Haryana)]	
ipat	'Kundli'	[H (Haryana)]	Gurgaon	'Bilaspur'	[HB (Haryana)]	



In []:

In [94]:

hypothesis testing/ visual analysis between actual_time aggregated value and OSRM time aggregated value

In [95]:

```
# H0: The average of actual and predicted times are equal  
# Ha: The average of actual and predicted times are not equal  
ztest(data['agg_actual_time'], data['agg_osrm_time'])
```

Out[95]:

```
(66.07555350813415, 0.0)
```

*Here p value is less than alpha so we reject null hypothesis

Hypothesis testing between actual_time aggregated value and segment actual time aggregated value

In [96]:

```
# H0: The average of actual and segmented actual times are equal  
# Ha: The average of actual and segmented actual times are not equal  
ztest(data['agg_actual_time'], data['agg_segment_actual_time'])
```

Out[96]:

```
(149.45336128899106, 0.0)
```

Do hypothesis testing/ visual analysis between osrm distance aggregated value and segment osrm distance aggregated value

In [97]:

```
# H0: The average of osrm distance and segmented osrm distance are equal  
# Ha: The average of osrm distance and segmented osrm distance are not equal  
ztest(data['agg_osrm_distance'], data['agg_segment_osrm_distance'])
```

Out[97]:

```
(146.87853467796583, 0.0)
```

Do hypothesis testing/ visual analysis between osrm time aggregated value and segment osrm time aggregated value

In [98]:

```
# H0: The average of osrm time and segmented osrm time are equal  
# Ha: The average of osrm time and segmented osrm time are not equal  
ztest(data['agg_osrm_time'], data['agg_segment_osrm_time'])
```

Out[98]:

```
(147.37782612875128, 0.0)
```

Here we are rejecting all our null hypothesis because our features data is not following normal distributions

OUTLIER TREATMENT

In [99]:

```
con=list((data.select_dtypes(include=['int64','float64'])))
con
```

Out[99]:

```
['start_scan_to_end_scan',
 'actual_distance_to_destination',
 'actual_time',
 'osrm_time',
 'osrm_distance',
 'factor',
 'segment_actual_time',
 'segment_osrm_time',
 'segment_osrm_distance',
 'od_start_hour',
 'od_end_hour',
 'agg_actual_time',
 'agg_segment_actual_time',
 'agg_osrm_time',
 'agg_segment_osrm_time',
 'agg_osrm_distance',
 'agg_segment_osrm_distance',
 'trip_creation_time_year',
 'diff_od_time_mins']
```

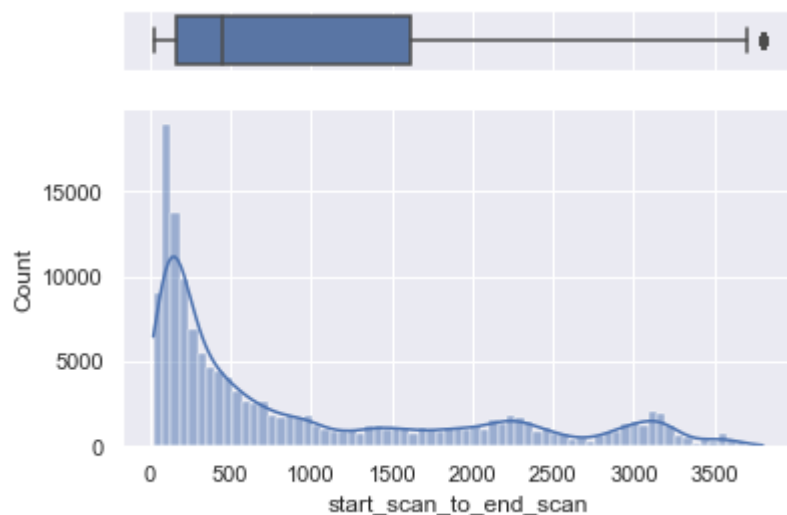
In [100]:

```
#start_scan_to_end_scan
Q3 = data['start_scan_to_end_scan'].quantile(0.75)
Q1 = data['start_scan_to_end_scan'].quantile(0.25)
IQR = Q3-Q1
upper = Q3+(1.5*IQR)
lower = Q1-(1.5*IQR)
print(upper,lower)
data = data[(data['start_scan_to_end_scan']>lower) & (data['start_scan_to_end_scan']<upper)]
```

3843.5 -2048.5

In [101]:

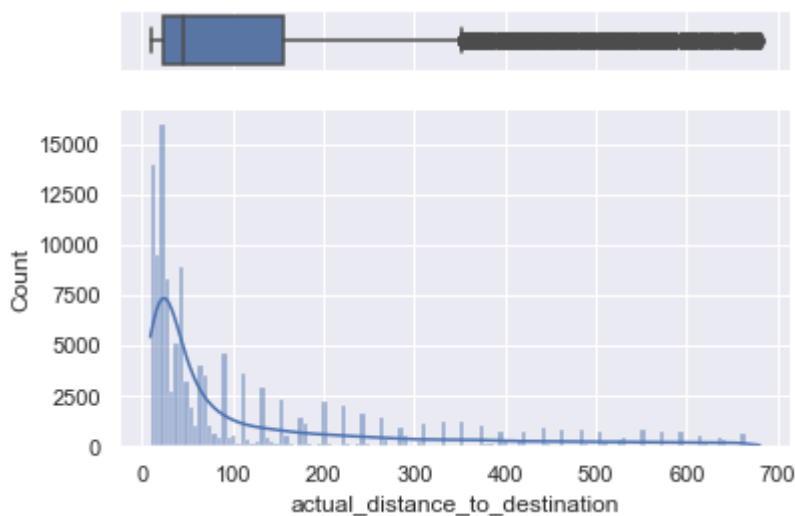
```
sns.set(style="darkgrid")
f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
sns.boxplot(data=data, x='start_scan_to_end_scan', ax=ax_box)
sns.histplot(data=data, x='start_scan_to_end_scan', ax=ax_hist, kde=True)
ax_box.set(xlabel='')
plt.show()
```



In [103]:

```
#actual_distance_to_destination
Q3 = data['actual_distance_to_destination'].quantile(0.75)
Q1 = data['actual_distance_to_destination'].quantile(0.25)
IQR = Q3-Q1
upper = Q3+(1.5*IQR)
lower = Q1-(1.5*IQR)
print(upper,lower)
data = data[(data['actual_distance_to_destination']>lower) & (data['actual_distance_to_destination']<upper)]
sns.set(style="darkgrid")
f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
sns.boxplot(data=data, x='actual_distance_to_destination', ax=ax_box)
sns.histplot(data=data, x='actual_distance_to_destination', ax=ax_hist, kde=True)
ax_box.set(xlabel='')
plt.show()
```

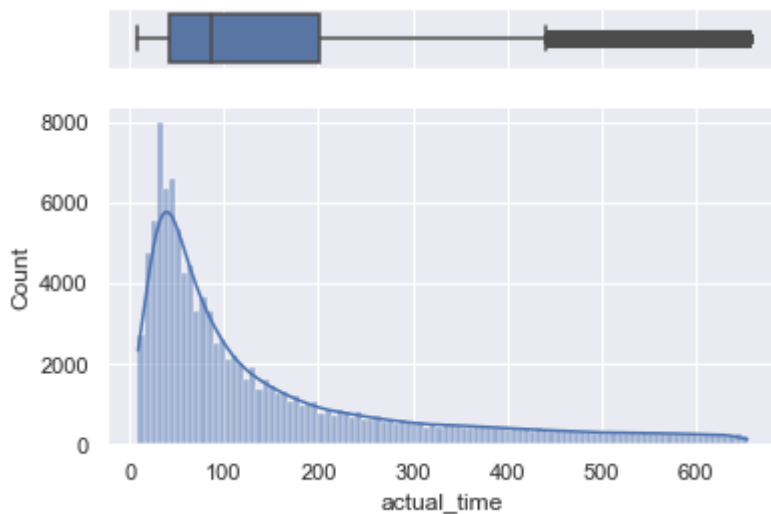
681.0150207887134 -371.27176693600586



In [104]:

```
#actual_time
Q3 = data['actual_time'].quantile(0.75)
Q1 = data['actual_time'].quantile(0.25)
IQR = Q3-Q1
upper = Q3+(1.5*IQR)
lower = Q1-(1.5*IQR)
print(upper,lower)
data = data[(data['actual_time']>lower) & (data['actual_time']<upper)]
sns.set(style="darkgrid")
f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
sns.boxplot(data=data, x='actual_time', ax=ax_box)
sns.histplot(data=data, x='actual_time', ax=ax_hist, kde=True)
ax_box.set(xlabel='')
plt.show()
```

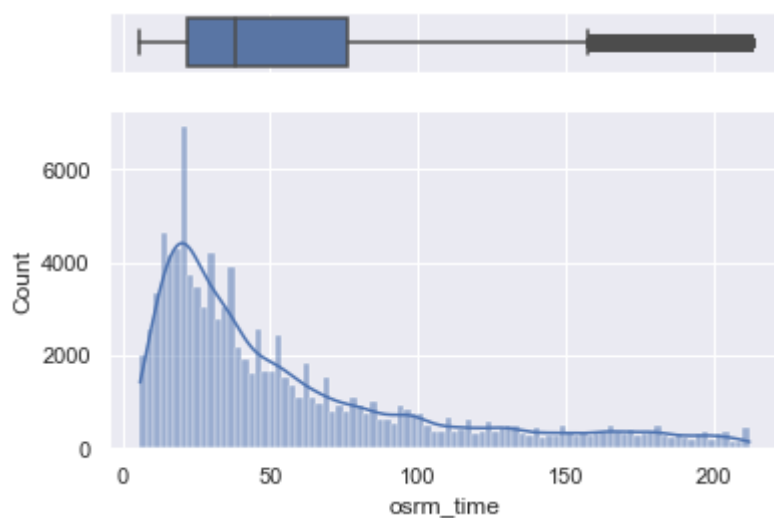
656.0 -320.0



In [106]:

```
#osrm_time
Q3 = data['osrm_time'].quantile(0.75)
Q1 = data['osrm_time'].quantile(0.25)
IQR = Q3-Q1
upper = Q3+(1.5*IQR)
lower = Q1-(1.5*IQR)
print(upper,lower)
data = data[(data['osrm_time']>lower) & (data['osrm_time']<upper)]
sns.set(style="darkgrid")
f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
sns.boxplot(data=data, x='osrm_time', ax=ax_box)
sns.histplot(data=data, x='osrm_time', ax=ax_hist, kde=True)
ax_box.set(xlabel='')
plt.show()
```

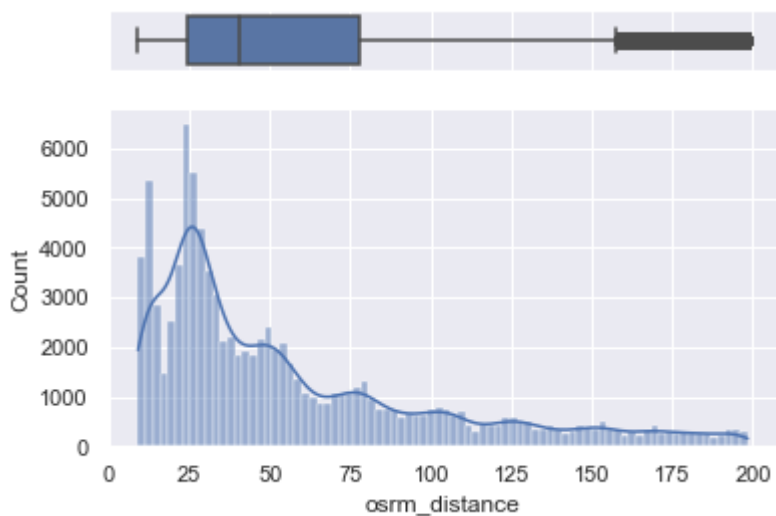
213.0 -91.0



In [107]:

```
#osrm_distance
Q3 = data['osrm_distance'].quantile(0.75)
Q1 = data['osrm_distance'].quantile(0.25)
IQR = Q3-Q1
upper = Q3+(1.5*IQR)
lower = Q1-(1.5*IQR)
print(upper,lower)
data = data[(data['osrm_distance']>lower) & (data['osrm_distance']<upper)]
sns.set(style="darkgrid")
f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
sns.boxplot(data=data, x='osrm_distance', ax=ax_box)
sns.histplot(data=data, x='osrm_distance', ax=ax_hist, kde=True)
ax_box.set(xlabel='')
plt.show()
```

198.37162499999997 -78.85917499999996



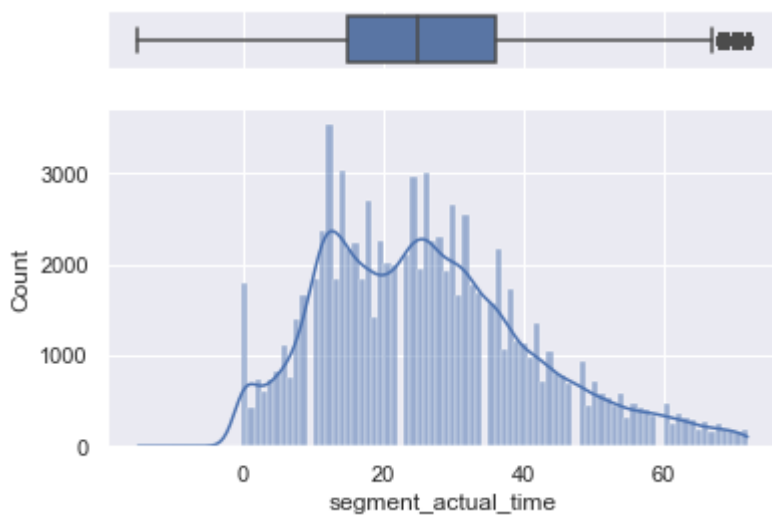
In [108]:

```

#segment_actual_time
Q3 = data['segment_actual_time'].quantile(0.75)
Q1 = data['segment_actual_time'].quantile(0.25)
IQR = Q3-Q1
upper = Q3+(1.5*IQR)
lower = Q1-(1.5*IQR)
print(upper,lower)
data = data[(data['segment_actual_time']>lower) & (data['segment_actual_time']<upper)]
sns.set(style="darkgrid")
f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
sns.boxplot(data=data,x='segment_actual_time', ax=ax_box)
sns.histplot(data=data, x='segment_actual_time', ax=ax_hist,kde=True)
ax_box.set(xlabel='')
plt.show()

```

72.5 -19.5



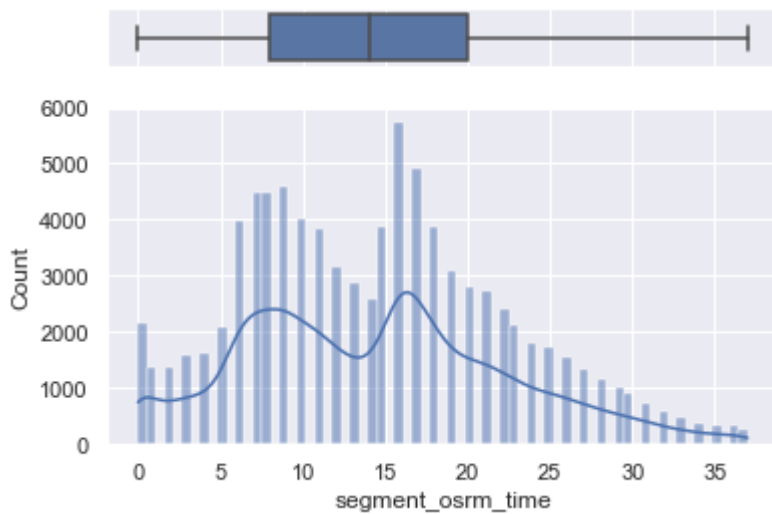
In [109]:

```

#segment_osrm_time
Q3 = data['segment_osrm_time'].quantile(0.75)
Q1 = data['segment_osrm_time'].quantile(0.25)
IQR = Q3-Q1
upper = Q3+(1.5*IQR)
lower = Q1-(1.5*IQR)
print(upper,lower)
data = data[(data['segment_osrm_time']>lower) & (data['segment_osrm_time']<upper)]
sns.set(style="darkgrid")
f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
sns.boxplot(data=data,x='segment_osrm_time', ax=ax_box)
sns.histplot(data=data, x='segment_osrm_time', ax=ax_hist,kde=True)
ax_box.set(xlabel='')
plt.show()

```

38.0 -10.0



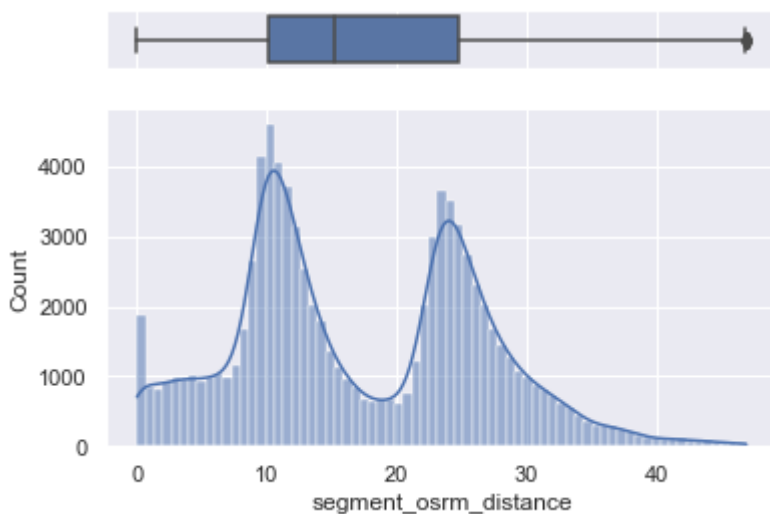
In [110]:

```

#segment_osrm_distance
Q3 = data['segment_osrm_distance'].quantile(0.75)
Q1 = data['segment_osrm_distance'].quantile(0.25)
IQR = Q3-Q1
upper = Q3+(1.5*IQR)
lower = Q1-(1.5*IQR)
print(upper,lower)
data = data[(data['segment_osrm_distance']>lower) & (data['segment_osrm_distance']<upper)]
sns.set(style="darkgrid")
f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
sns.boxplot(data=data,x='segment_osrm_distance', ax=ax_box)
sns.histplot(data=data, x='segment_osrm_distance', ax=ax_hist,kde=True)
ax_box.set(xlabel='')
plt.show()

```

46.97781250000001 -12.066087500000007



SUMMARY

- Above i have done the outlier detection and correction

One hot Encoding

In [113]:

```

from sklearn.preprocessing import OneHotEncoder

```

In [115]:

```
data['route_type']
```

Out[115]:

```
0      Carting
1      Carting
2      Carting
3      Carting
4      Carting
...
144861  Carting
144862  Carting
144863  Carting
144864  Carting
144865  Carting
Name: route_type, Length: 87891, dtype: category
Categories (2, object): ['Carting', 'FTL']
```

In [117]:

```
one=OneHotEncoder()
one_data=pd.DataFrame(one.fit_transform(data[['route_type']]).toarray())
data=data.join(one_data)
```

In [118]:

data

Out[118]:

trip_creation_time	route_schedule_uuid	route_type	trip_uuid	source_center	
2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	IND388121AAA	Anand_
2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	IND388121AAA	Anand_
2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	IND388121AAA	Anand_
2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	IND388121AAA	Anand_
2018-09-20 02:35:36.476840	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting	153741093647649320	IND388121AAA	Anand_
...
2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	IND131028AAB	Son
2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	IND131028AAB	Son
2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	IND131028AAB	Son
2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	IND131028AAB	Son
2018-09-20 16:24:28.436231	thanos::sroute:f0569d2f-4e20-4c31-8542-67b86d5...	Carting	153746066843555182	IND131028AAB	Son

olumns



In [126]:

```
data_con=data[['start_scan_to_end_scan',  
               'actual_distance_to_destination',  
               'actual_time',  
               'osrm_time',  
               'osrm_distance',  
               'segment_actual_time',  
               'segment_osrm_time',  
               'segment_osrm_distance']]  
from sklearn.preprocessing import MinMaxScaler  
  
norm = MinMaxScaler().fit(data_con)  
  
data_norm_con = norm.transform(data_con)  
data_norm_con
```

Out[126]:

```
array([[0.01745108, 0.00818264, 0.00773994, ..., 0.33333333, 0.2972973 ,  
        0.25475484],  
       [0.01745108, 0.05663722, 0.02321981, ..., 0.28735632, 0.24324324,  
        0.20778021],  
       [0.01745108, 0.1062275 , 0.04798762, ..., 0.35632184, 0.18918919,  
        0.23026791],  
       ...,  
       [0.10761502, 0.25701573, 0.17182663, ..., 0.47126437, 0.56756757,  
        0.36988028],  
       [0.10761502, 0.32581771, 0.20278638, ..., 0.40229885, 0.91891892,  
        0.44083938],  
       [0.10761502, 0.36866314, 0.23065015, ..., 0.36781609, 0.72972973,  
        0.40215764]])
```

In []: