

cause-analysis-casestudy-kuntal-k

October 14, 2024

```
[7]: # Import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import norm
import warnings
warnings.filterwarnings('ignore')
```

1 Data Observation

```
[9]: # Loading the dataset
dt = pd.read_csv("https://d2beiqkhq929f0.cloudfront.net/public_assets/assets/
↳000/001/551/original/delhivery_data.csv?1642751181")
dt.head(3)
```

```
[9]:
```

	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	

		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	

	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)	

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

	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	

	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	

	segment_factor
0	1.272727
1	1.111111
2	2.285714

[3 rows x 24 columns]

```
[10]: # getting the counts of rows and columns in the dataset
dt.shape
```

```
[10]: (144867, 24)
```

```
[11]: # getting the information of the dataset
dt.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

2 Column Profiling:

data - tells whether the data is testing or training data

trip_creation_time - Timestamp of trip creation

route_schedule_uuid - Unique Id for a particular route schedule

route_type - Transportation type

FTL - Full Truck Load: FTL shipments get to the destination sooner, as the truck is making no other pickups or drop-offs along the way
Carting: Handling system consisting of small vehicles (carts)

trip_uuid - Unique ID given to a particular trip (A trip may include different source and destination centers)

source_center - Source ID of trip origin

source_name - Source Name of trip origin

destination_cente - Destination ID

destination_name - Destination Name

od_start_time - Trip start time

od_end_time - Trip end time

start_scan_to_end_scan - Time taken to deliver from source to destination

is_cutoff - Unknown field

cutoff_factor - Unknown field

cutoff_timestamp - Unknown field

actual_distance_to_destination - Distance in Kms between source and destination warehouse

actual_time - Actual time taken to complete the delivery (Cumulative)

osrm_time - An open-source routing engine time calculator which computes the shortest path between points in a given map (Includes usual traffic, distance through major and minor roads) and gives the time (Cumulative)

osrm_distance - An open-source routing engine which computes the shortest path between points in a given map (Includes usual traffic, distance through major and minor roads) (Cumulative)

factor - Unknown field

segment_actual_time - This is a segment time. Time taken by the subset of the package delivery

segment_osrm_time - This is the OSRM segment time. Time taken by the subset of the pack-

age delivery

segment_osrm_distance – This is the OSRM distance. Distance covered by subset of the package delivery

segment_factor – Unknown field

```
[12]: # checking for null values
      dt.isnull().sum()
```

```
[12]: data
      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
```

We can see that null values are present in source_name and destination_name column

```
[13]: # Removing rows if any null values are present
      data = dt.dropna(how = 'any')
```

```
[14]: # getting the rows and column no. of original table
      dt.shape
```

```
[14]: (144867, 24)
```

```
[15]: # getting the rows and column no. of new table after removing null values rows
      data.shape
```

```
[15]: (144316, 24)
```

```
[16]: # chcking the null values of nww tables
data.isnull().sum()
```

```
[16]: data
trip_creation_time      0
route_schedule_uuid     0
route_type              0
trip_uuid               0
source_center           0
source_name             0
destination_center      0
destination_name        0
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
```

```
[17]: # no of unique values in each column of new table
for i in data.columns:
    print(i,data[i].nunique())
```

```
data 2
trip_creation_time 14787
route_schedule_uuid 1497
route_type 2
trip_uuid 14787
source_center 1496
source_name 1496
destination_center 1466
destination_name 1466
od_start_time 26223
od_end_time 26223
start_scan_to_end_scan 1914
```

```

is_cutoff 2
cutoff_factor 501
cutoff_timestamp 92894
actual_distance_to_destination 143965
actual_time 3182
osrm_time 1531
osrm_distance 137544
factor 45588
segment_actual_time 746
segment_osrm_time 214
segment_osrm_distance 113497
segment_factor 5663

```

```

[18]: # converting the object dtype time columns into pandas datetime dtype
data['od_start_time'] = pd.to_datetime(data['od_start_time'])
data['od_end_time'] = pd.to_datetime(data['od_end_time'])
data['trip_creation_time'] = pd.to_datetime(data['trip_creation_time'])

```

```

[19]: # checking datatypes of new table
data.dtypes

```

```

[19]: data                                object
trip_creation_time                       datetime64[ns]
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                          datetime64[ns]
od_end_time                            datetime64[ns]
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

```

```
[20]: # Get the unique values of the 'data' column
unique_data = data['data'].unique()
unique_data
```

```
[20]: array(['training', 'test'], dtype=object)
```

```
[21]: # Get the unique values of the 'route_type' column
unique_route_type = data['route_type'].unique()
unique_route_type
```

```
[21]: array(['Carting', 'FTL'], dtype=object)
```

```
[22]: # creating a new new col by merging 3 columns
data['segment_key'] = data['trip_uuid'] + data['source_center'] +
↳ data['destination_center']
data['segment_key'].head(3)
```

```
[22]: 0    trip-153741093647649320IND388121AAAIND388620AAB
1    trip-153741093647649320IND388121AAAIND388620AAB
2    trip-153741093647649320IND388121AAAIND388620AAB
Name: segment_key, dtype: object
```

```
[23]: # creating a df segment_col of 3 columns
segment_col =
↳ ['segment_actual_time', 'segment_osrm_time', 'segment_osrm_distance']
data[segment_col].head()
```

```
[23]:    segment_actual_time  segment_osrm_time  segment_osrm_distance
0                14.0             11.0             11.9653
1                10.0              9.0              9.7590
2                16.0              7.0             10.8152
3                21.0             12.0             13.0224
4                 6.0              5.0              3.9153
```

```
[24]: # Grouping by the segment_key of sub-journey in the trip
for i in segment_col:
    data[i+'_sum'] = data.groupby('segment_key')[i].cumsum()

segment_col_sum = [i+'_sum' for i in segment_col]
data[segment_col_sum].head()
```

```
[24]:    segment_actual_time_sum  segment_osrm_time_sum  segment_osrm_distance_sum
0                14.0             11.0             11.9653
1                24.0             20.0             21.7243
2                40.0             27.0             32.5395
3                61.0             39.0             45.5619
4                67.0             44.0             49.4772
```

```
[25]: # creating a dictionary of 19 columns for aggregating at sub journey level
agg_dict = {'data': 'first', 'trip_creation_time': 'first',
            'route_schedule_uuid': 'first', 'route_type': 'first',
            'trip_uuid': 'first', 'source_center': 'first', 'source_name':
            'first', 'destination_center': 'last',
            'destination_name': 'last', 'od_start_time': 'first',
            'od_end_time': 'first', 'start_scan_to_end_scan': 'first',
            'actual_distance_to_destination': 'last', 'actual_time': 'last',
            'osrm_time': 'last', 'osrm_distance': 'last',
            'segment_actual_time_sum': 'last', 'segment_osrm_distance_sum':
            'last', 'segment_osrm_time_sum': 'last',}
agg_dict
```

```
[25]: {'data': 'first',
      'trip_creation_time': 'first',
      'route_schedule_uuid': 'first',
      'route_type': 'first',
      'trip_uuid': 'first',
      'source_center': 'first',
      'source_name': 'first',
      'destination_center': 'last',
      'destination_name': 'last',
      'od_start_time': 'first',
      'od_end_time': 'first',
      'start_scan_to_end_scan': 'first',
      'actual_distance_to_destination': 'last',
      'actual_time': 'last',
      'osrm_time': 'last',
      'osrm_distance': 'last',
      'segment_actual_time_sum': 'last',
      'segment_osrm_distance_sum': 'last',
      'segment_osrm_time_sum': 'last'}
```

```
[26]: segment = data.groupby('segment_key').agg(agg_dict).reset_index()
segment.head()
```

```
[26]:
```

	segment_key	data \
0	trip-153671041653548748IND209304AAAIND000000ACB	training
1	trip-153671041653548748IND462022AAAIND209304AAA	training
2	trip-153671042288605164IND561203AABIND562101AAA	training
3	trip-153671042288605164IND572101AAAIND561203AAB	training
4	trip-153671043369099517IND000000ACBIND160002AAC	training


```

      trip_creation_time \
0 2018-09-12 00:00:16.535741
1 2018-09-12 00:00:16.535741
2 2018-09-12 00:00:22.886430
```


3 2018-09-12 00:00:22.886430
 4 2018-09-12 00:00:33.691250

	route_schedule_uuid	route_type	\
0	thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...	FTL	
1	thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...	FTL	
2	thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...	Carting	
3	thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...	Carting	
4	thanos::sroute:de5e208e-7641-45e6-8100-4d9fb1e...	FTL	

	trip_uuid	source_center	source_name	\
0	trip-153671041653548748	IND209304AAA	Kanpur_Central_H_6 (Uttar Pradesh)	
1	trip-153671041653548748	IND462022AAA	Bhopal_Trnsport_H (Madhya Pradesh)	
2	trip-153671042288605164	IND561203AAB	Doddablpur_ChikaDPP_D (Karnataka)	
3	trip-153671042288605164	IND572101AAA	Tumkur_Veersagr_I (Karnataka)	
4	trip-153671043369099517	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	

	destination_center	destination_name	\
0	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	
1	IND209304AAA	Kanpur_Central_H_6 (Uttar Pradesh)	
2	IND562101AAA	Chikblapur_ShntiSgr_D (Karnataka)	
3	IND561203AAB	Doddablpur_ChikaDPP_D (Karnataka)	
4	IND160002AAC	Chandigarh_Mehmdpur_H (Punjab)	

	od_start_time	od_end_time	\
0	2018-09-12 16:39:46.858469	2018-09-13 13:40:23.123744	
1	2018-09-12 00:00:16.535741	2018-09-12 16:39:46.858469	
2	2018-09-12 02:03:09.655591	2018-09-12 03:01:59.598855	
3	2018-09-12 00:00:22.886430	2018-09-12 02:03:09.655591	
4	2018-09-14 03:40:17.106733	2018-09-14 17:34:55.442454	

	start_scan_to_end_scan	actual_distance_to_destination	actual_time	\
0	1260.0	383.759164	732.0	
1	999.0	440.973689	830.0	
2	58.0	24.644021	47.0	
3	122.0	48.542890	96.0	
4	834.0	237.439610	611.0	

	osrm_time	osrm_distance	segment_actual_time_sum	\
0	329.0	446.5496	728.0	
1	388.0	544.8027	820.0	
2	26.0	28.1994	46.0	
3	42.0	56.9116	95.0	
4	212.0	281.2109	608.0	

	segment_osrm_distance_sum	segment_osrm_time_sum
0	670.6205	534.0

1	649.8528	474.0
2	28.1995	26.0
3	55.9899	39.0
4	317.7408	231.0

```
[27]: #Groupby mini-trips, sorting by time
segment = segment.sort_values(by=['segment_key', 'od_end_time'],
    ↪ascending=True).reset_index()
segment.head(3)
```

```
[27]:      index      segment_key      data \
0      0  trip-153671041653548748IND209304AAAAIND000000ACB  training
1      1  trip-153671041653548748IND462022AAAAIND209304AAA  training
2      2  trip-153671042288605164IND561203AABIND562101AAA  training

      trip_creation_time \
0 2018-09-12 00:00:16.535741
1 2018-09-12 00:00:16.535741
2 2018-09-12 00:00:22.886430

      route_schedule_uuid route_type \
0  thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...      FTL
1  thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...      FTL
2  thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...  Carting

      trip_uuid source_center      source_name \
0  trip-153671041653548748  IND209304AAA  Kanpur_Central_H_6 (Uttar Pradesh)
1  trip-153671041653548748  IND462022AAA  Bhopal_Trnsport_H (Madhya Pradesh)
2  trip-153671042288605164  IND561203AAB  Doddablpur_ChikaDPP_D (Karnataka)

      destination_center ...      od_start_time \
0      IND000000ACB ... 2018-09-12 16:39:46.858469
1      IND209304AAA ... 2018-09-12 00:00:16.535741
2      IND562101AAA ... 2018-09-12 02:03:09.655591

      od_end_time start_scan_to_end_scan \
0 2018-09-13 13:40:23.123744      1260.0
1 2018-09-12 16:39:46.858469      999.0
2 2018-09-12 03:01:59.598855      58.0

      actual_distance_to_destination actual_time osrm_time osrm_distance \
0      383.759164      732.0      329.0      446.5496
1      440.973689      830.0      388.0      544.8027
2      24.644021      47.0      26.0      28.1994

      segment_actual_time_sum segment_osrm_distance_sum segment_osrm_time_sum
0      728.0      670.6205      534.0
```

1	820.0	649.8528	474.0
2	46.0	28.1995	26.0

[3 rows x 21 columns]

```
[28]: # getting info of one trip id from segment table
segment[segment['trip_uuid'] == 'trip-153671041653548748']
```

```
[28]:      index      segment_key      data \
0      0  trip-153671041653548748IND209304AAAAIND000000ACB  training
1      1  trip-153671041653548748IND462022AAAAIND209304AAA  training

      trip_creation_time \
0 2018-09-12 00:00:16.535741
1 2018-09-12 00:00:16.535741

      route_schedule_uuid route_type \
0  thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...      FTL
1  thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...      FTL

      trip_uuid source_center      source_name \
0  trip-153671041653548748  IND209304AAA  Kanpur_Central_H_6 (Uttar Pradesh)
1  trip-153671041653548748  IND462022AAA  Bhopal_Trnsport_H (Madhya Pradesh)

      destination_center ...      od_start_time \
0      IND000000ACB ... 2018-09-12 16:39:46.858469
1      IND209304AAA ... 2018-09-12 00:00:16.535741

      od_end_time start_scan_to_end_scan \
0 2018-09-13 13:40:23.123744      1260.0
1 2018-09-12 16:39:46.858469      999.0

      actual_distance_to_destination  actual_time  osrm_time  osrm_distance \
0      383.759164      732.0      329.0      446.5496
1      440.973689      830.0      388.0      544.8027

      segment_actual_time_sum  segment_osrm_distance_sum  segment_osrm_time_sum
0      728.0      670.6205      534.0
1      820.0      649.8528      474.0
```

[2 rows x 21 columns]

```
[29]: # getting the info of segment table
segment.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 26222 entries, 0 to 26221
```

Data columns (total 21 columns):

#	Column	Non-Null Count	Dtype
0	index	26222 non-null	int64
1	segment_key	26222 non-null	object
2	data	26222 non-null	object
3	trip_creation_time	26222 non-null	datetime64[ns]
4	route_schedule_uuid	26222 non-null	object
5	route_type	26222 non-null	object
6	trip_uuid	26222 non-null	object
7	source_center	26222 non-null	object
8	source_name	26222 non-null	object
9	destination_center	26222 non-null	object
10	destination_name	26222 non-null	object
11	od_start_time	26222 non-null	datetime64[ns]
12	od_end_time	26222 non-null	datetime64[ns]
13	start_scan_to_end_scan	26222 non-null	float64
14	actual_distance_to_destination	26222 non-null	float64
15	actual_time	26222 non-null	float64
16	osrm_time	26222 non-null	float64
17	osrm_distance	26222 non-null	float64
18	segment_actual_time_sum	26222 non-null	float64
19	segment_osrm_distance_sum	26222 non-null	float64
20	segment_osrm_time_sum	26222 non-null	float64

dtypes: datetime64[ns](3), float64(8), int64(1), object(9)
memory usage: 4.2+ MB

2.1 Feature Creation

2.1.1 Calculate time taken between od_start_time and od_end_time

```
[30]: #getting the time difference between od_start_time and od_end_time in hour in
      ↪an separate column in segment table
segment['hour_taken'] = (segment['od_end_time'] - segment['od_start_time']).dt.
      ↪total_seconds()/60
segment['hour_taken'].head(3)
```

```
[30]: 0    1260.604421
      1     999.505379
      2     58.832388
      Name: hour_taken, dtype: float64
```

```
[31]: segment.head()
```

```
[31]:   index          segment_key  data \
0      0  trip-153671041653548748IND209304AAAINDD000000ACB  training
1      1  trip-153671041653548748IND462022AAAINDD209304AAA  training
2      2  trip-153671042288605164IND561203AABIND562101AAA  training
```

3	3	trip-153671042288605164IND572101AAAIND561203AAB	training
4	4	trip-153671043369099517IND000000ACBIND160002AAC	training

	trip_creation_time	\
0	2018-09-12 00:00:16.535741	
1	2018-09-12 00:00:16.535741	
2	2018-09-12 00:00:22.886430	
3	2018-09-12 00:00:22.886430	
4	2018-09-12 00:00:33.691250	

	route_schedule_uuid	route_type	\
0	thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...	FTL	
1	thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...	FTL	
2	thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...	Carting	
3	thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...	Carting	
4	thanos::sroute:de5e208e-7641-45e6-8100-4d9fb1e...	FTL	

	trip_uuid	source_center	source_name	\
0	trip-153671041653548748	IND209304AAA	Kanpur_Central_H_6 (Uttar Pradesh)	
1	trip-153671041653548748	IND462022AAA	Bhopal_Trnsport_H (Madhya Pradesh)	
2	trip-153671042288605164	IND561203AAB	Doddablpur_ChikaDPP_D (Karnataka)	
3	trip-153671042288605164	IND572101AAA	Tumkur_Veersagr_I (Karnataka)	
4	trip-153671043369099517	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	

	destination_center	...	od_end_time	start_scan_to_end_scan	\
0	IND000000ACB	...	2018-09-13 13:40:23.123744	1260.0	
1	IND209304AAA	...	2018-09-12 16:39:46.858469	999.0	
2	IND562101AAA	...	2018-09-12 03:01:59.598855	58.0	
3	IND561203AAB	...	2018-09-12 02:03:09.655591	122.0	
4	IND160002AAC	...	2018-09-14 17:34:55.442454	834.0	

	actual_distance_to_destination	actual_time	osrm_time	osrm_distance	\
0	383.759164	732.0	329.0	446.5496	
1	440.973689	830.0	388.0	544.8027	
2	24.644021	47.0	26.0	28.1994	
3	48.542890	96.0	42.0	56.9116	
4	237.439610	611.0	212.0	281.2109	

	segment_actual_time_sum	segment_osrm_distance_sum	segment_osrm_time_sum	\
0	728.0	670.6205	534.0	
1	820.0	649.8528	474.0	
2	46.0	28.1995	26.0	
3	95.0	55.9899	39.0	
4	608.0	317.7408	231.0	

	hour_taken
0	1260.604421

```

1    999.505379
2     58.832388
3    122.779486
4    834.638929

```

[5 rows x 22 columns]

```

[32]: # creating another dictionary for aggregation in gesment table
trip_dict = {'data': 'first', 'trip_creation_time': 'first',
             ↪ 'route_schedule_uuid': 'first', 'route_type': 'first', 'trip_uuid':
             ↪ 'first', 'source_center': 'first', 'source_name': 'first',
               'destination_center': 'last', 'destination_name': 'last',
             ↪ 'start_scan_to_end_scan': 'sum', 'hour_taken': 'sum',
             ↪ 'actual_distance_to_destination': 'sum',
               'actual_time': 'sum', 'osrm_time': 'sum', 'osrm_distance':
             ↪ 'sum', 'segment_actual_time_sum': 'sum', 'segment_osrm_distance_sum':
             ↪ 'sum', 'segment_osrm_time_sum': 'sum',}

trip_dict

```

```

[32]: {'data': 'first',
       'trip_creation_time': 'first',
       'route_schedule_uuid': 'first',
       'route_type': 'first',
       'trip_uuid': 'first',
       'source_center': 'first',
       'source_name': 'first',
       'destination_center': 'last',
       'destination_name': 'last',
       'start_scan_to_end_scan': 'sum',
       'hour_taken': 'sum',
       'actual_distance_to_destination': 'sum',
       'actual_time': 'sum',
       'osrm_time': 'sum',
       'osrm_distance': 'sum',
       'segment_actual_time_sum': 'sum',
       'segment_osrm_distance_sum': 'sum',
       'segment_osrm_time_sum': 'sum'}

```

```

[88]: # creating another table, grouppping by trip id in segment table
trip = segment.groupby('trip_uuid').agg(trip_dict)
trip = trip.reset_index(drop = True)
trip.head()

```

```

[88]:      data      trip_creation_time \
0  training 2018-09-12 00:00:16.535741
1  training 2018-09-12 00:00:22.886430

```

```

2 training 2018-09-12 00:00:33.691250
3 training 2018-09-12 00:01:00.113710
4 training 2018-09-12 00:02:09.740725

```

```

                                route_schedule_uuid route_type \
0 thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...      FTL
1 thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...      Carting
2 thanos::sroute:de5e208e-7641-45e6-8100-4d9fb1e...      FTL
3 thanos::sroute:f0176492-a679-4597-8332-bbd1c7f...      Carting
4 thanos::sroute:d9f07b12-65e0-4f3b-bec8-df06134...      FTL

```

```

                                trip_uuid source_center source_name \
0 trip-153671041653548748 IND209304AAA Kanpur_Central_H_6 (Uttar Pradesh)
1 trip-153671042288605164 IND561203AAB Doddablpur_ChikaDPP_D (Karnataka)
2 trip-153671043369099517 IND000000ACB Gurgaon_Bilaspur_HB (Haryana)
3 trip-153671046011330457 IND400072AAB Mumbai_Hub (Maharashtra)
4 trip-153671052974046625 IND583101AAA Bellary_Dc (Karnataka)

```

```

destination_center destination_name \
0 IND209304AAA Kanpur_Central_H_6 (Uttar Pradesh)
1 IND561203AAB Doddablpur_ChikaDPP_D (Karnataka)
2 IND000000ACB Gurgaon_Bilaspur_HB (Haryana)
3 IND401104AAA Mumbai_MiraRd_IP (Maharashtra)
4 IND583119AAA Sandur_WrdN1DPP_D (Karnataka)

```

```

start_scan_to_end_scan hour_taken actual_distance_to_destination \
0 2259.0 2260.109800 824.732854
1 180.0 181.611874 73.186911
2 3933.0 3934.362520 1927.404273
3 100.0 100.494935 17.175274
4 717.0 718.349042 127.448500

```

```

actual_time osrm_time osrm_distance segment_actual_time_sum \
0 1562.0 717.0 991.3523 1548.0
1 143.0 68.0 85.1110 141.0
2 3347.0 1740.0 2354.0665 3308.0
3 59.0 15.0 19.6800 59.0
4 341.0 117.0 146.7918 340.0

```

```

segment_osrm_distance_sum segment_osrm_time_sum
0 1320.4733 1008.0
1 84.1894 65.0
2 2545.2678 1941.0
3 19.8766 16.0
4 146.7919 115.0

```

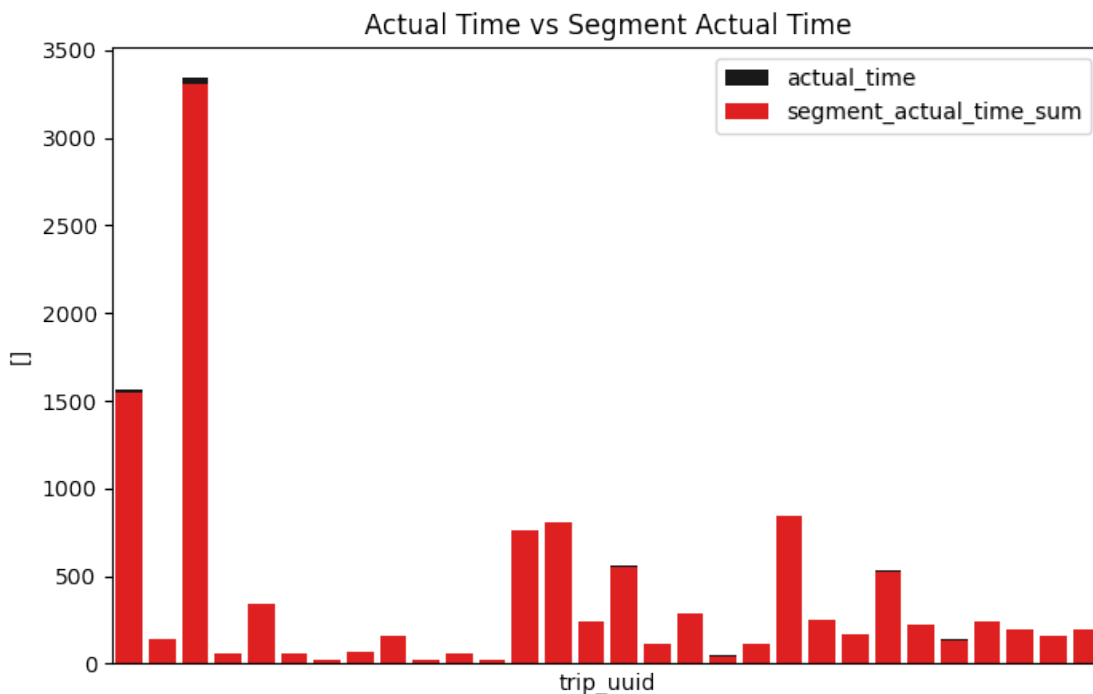
```
[34]: trip1 = trip[['actual_time', 'segment_actual_time_sum', 'trip_uuid']].head(30)
trip1.head(3)
```

```
[34]:    actual_time  segment_actual_time_sum  trip_uuid
0      1562.0          1548.0  trip-153671041653548748
1       143.0          141.0  trip-153671042288605164
2     3347.0          3308.0  trip-153671043369099517
```

```
[35]: # Create the figure with a transparent background
plt.figure(figsize=(8, 5), facecolor='none')

sns.barplot(data=trip1, y = 'actual_time', x = 'trip_uuid', label = 'actual_time', color = 'black', alpha = 0.9)
sns.barplot(data=trip1, y = 'segment_actual_time_sum', x = 'trip_uuid', color = 'red', label = 'segment_actual_time_sum')
plt.title('Actual Time vs Segment Actual Time')
# Remove x-axis labels
plt.xticks([])
plt.ylabel('')

# Show the plot
plt.show()
```



As we can see from the trend, 'actual_time' and 'segment_actual_time_sum' of every 'trip_uuid'

are not that different

```
[36]: trip.head(3)
```

```
[36]:      data      trip_creation_time \
0  training 2018-09-12 00:00:16.535741
1  training 2018-09-12 00:00:22.886430
2  training 2018-09-12 00:00:33.691250

      route_schedule_uuid route_type \
0  thanos::sroute:d7c989ba-a29b-4a0b-b2f4-288cdc6...      FTL
1  thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...      Carting
2  thanos::sroute:de5e208e-7641-45e6-8100-4d9fb1e...      FTL

      trip_uuid source_center      source_name \
0  trip-153671041653548748  IND209304AAA  Kanpur_Central_H_6 (Uttar Pradesh)
1  trip-153671042288605164  IND561203AAB  Doddablpur_ChikaDPP_D (Karnataka)
2  trip-153671043369099517  IND000000ACB  Gurgaon_Bilaspur_HB (Haryana)

      destination_center      destination_name \
0      IND209304AAA  Kanpur_Central_H_6 (Uttar Pradesh)
1      IND561203AAB  Doddablpur_ChikaDPP_D (Karnataka)
2      IND000000ACB  Gurgaon_Bilaspur_HB (Haryana)

      start_scan_to_end_scan  hour_taken  actual_distance_to_destination \
0              2259.0      2260.109800              824.732854
1              180.0       181.611874              73.186911
2              3933.0      3934.362520             1927.404273

      actual_time  osrm_time  osrm_distance  segment_actual_time_sum \
0       1562.0       717.0       991.3523       1548.0
1        143.0        68.0        85.1110        141.0
2       3347.0      1740.0      2354.0665      3308.0

      segment_osrm_distance_sum  segment_osrm_time_sum
0              1320.4733              1008.0
1               84.1894               65.0
2             2545.2678             1941.0
```

```
[37]: # checking the one trip id if from trip table
trip[trip['trip_uuid']=='trip-153671042288605164']
```

```
[37]:      data      trip_creation_time \
1  training 2018-09-12 00:00:22.886430

      route_schedule_uuid route_type \
1  thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...      Carting
```

	trip_uuid	source_center	source_name	\
1	trip-153671042288605164	IND561203AAB	Doddablpur_ChikaDPP_D	(Karnataka)

	destination_center	destination_name	\
1	IND561203AAB	Doddablpur_ChikaDPP_D	(Karnataka)

	start_scan_to_end_scan	hour_taken	actual_distance_to_destination	\
1	180.0	181.611874	73.186911	

	actual_time	osrm_time	osrm_distance	segment_actual_time_sum	\
1	143.0	68.0	85.111	141.0	

	segment_osrm_distance_sum	segment_osrm_time_sum
1	84.1894	65.0

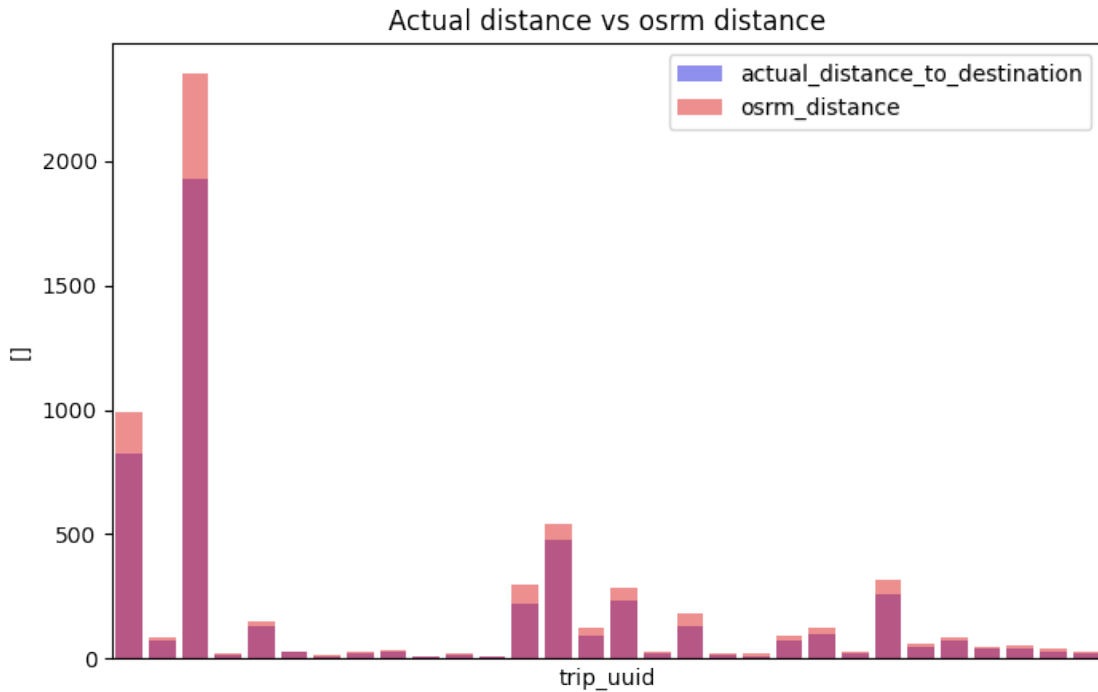
```
[38]: trip2 = trip[['actual_distance_to_destination', 'osrm_distance', 'trip_uuid']].
      ↪head(30)
      trip2.head(3)
```

```
[38]:      actual_distance_to_destination  osrm_distance      trip_uuid
0                824.732854          991.3523  trip-153671041653548748
1                 73.186911           85.1110  trip-153671042288605164
2            1927.404273        2354.0665  trip-153671043369099517
```

```
[39]: # Create the figure with a transparent background
plt.figure(figsize=(8, 5), facecolor='none')

sns.barplot(data=trip2, y = 'actual_distance_to_destination', x = 'trip_uuid',
      ↪label = 'actual_distance_to_destination', color = 'blue', alpha = 0.5)
sns.barplot(data=trip2, y = 'osrm_distance', x = 'trip_uuid', color = 'red',
      ↪label = 'osrm_distance', alpha = 0.5)
plt.title('Actual distance vs osrm distance')
# Remove x-axis labels
plt.xticks([])
plt.ylabel([])

# Show the plot
plt.show()
```



2.2 ### as we can observe that there are slight differences between ‘actual_distance_to_destination’ and ‘osrm_distance’

```
[40]: # converting all the values of destination_name and source_name column in lower
      ↪ case
```

```
trip['destination_name'] = trip['destination_name'].str.lower()
trip['source_name'] = trip['source_name'].str.lower()
trip['source_name'].head(3)
```

```
[40]: 0    kanpur_central_h_6 (uttar pradesh)
      1    doddablpur_chikadpp_d (karnataka)
      2    gurgaon_bilaspur_hb (haryana)
      Name: source_name, dtype: object
```

2.2.1 creating function to get the proper names from the columns

```
[41]: def get_state(x):
      state = x.split('(')[1] # getting 'uttar pradesh)' from 'kanpur_central_h_6
      ↪(uttar pradesh)'

      return state[:-1] # returning 'uttar pradesh' from 'uttar pradesh)'
```

```
[42]: def get_city(x):
    city = x.split('(')[0] # getting 'kanpur_central_h_6' from
    ↪ 'kanpur_central_h_6 (uttar pradesh)'
    city = city.split('_')[0] # getting 'kanpur'
    #Now dealing with edge cases
    if city == 'pnq vadgaon sheri dpc':
        return 'vadgaonsheri'

    # ['PNQ Pashan DPC', 'Bhopal MP Nagar', 'HBR Layout PC',
    #  'PNQ Rahatani DPC', 'Pune Balaji Nagar', 'Mumbai Antop Hill']

    if city in ['pnq pashan dpc', 'pnq rahatani dpc', 'pune balaji nagar']:
        return 'pune'

    if city == 'hbr layout pc' : return 'bengaluru'
    if city == 'bhopal mp nagar' : return 'bhopal'
    if city == 'mumbai antop hill' : return 'mumbai'

    return city
```

```
[43]: def place2city_place(x):
    # We will remove state
    x = x.split(' ')[0]
    len_ = len(x.split('_'))
    if len_ >= 3:
        return x.split('_')[1]

    # Small cities have same city and place name
    if len_ == 2:
        return x.split('_')[0]
    return x.split(' ')[0]
```

```
[44]: def get_code(x):
    # We will remove state
    x = x.split(' ')[0]

    if len(x.split('_')) >= 3 :
        return x.split('_')[-1]

    return 'none'
```

```
[45]: # getting the separate column of place , city, state , code from the
    ↪ destination name
trip['destination_state'] = trip['destination_name'].apply(lambda x:
    ↪ get_state(x))
trip['destination_city'] = trip['destination_name'].apply(lambda x:
    ↪ get_city(x))
```

```

trip['destination_place'] = trip['destination_name'].apply(lambda x:
    ↪place2city_place(x))
trip['destination_code'] = trip['destination_name'].apply(lambda x:
    ↪get_code(x))

trip[['destination_state', 'destination_city', 'destination_place',
    ↪'destination_code']]

```

```

[45]:      destination_state destination_city destination_place destination_code
0          uttar pradesh          kanpur          central          6
1            karnataka      doddablpur      chikadpp          d
2            haryana        gurgaon      bilaspur          hb
3      maharashtra        mumbai        mirard          ip
4            karnataka        sandur      wrdn1dpp          d
...
14782         punjab      chandigarh      mehmdpur          h
14783         haryana      faridabad      blbgarh          dc
14784      uttar pradesh          kanpur      govndngr          dc
14785         tamil nadu      tirschndr      shnmgprm          d
14786         karnataka        sandur      wrdn1dpp          d

```

[14787 rows x 4 columns]

```

[46]: # getting the separate column of place , city, state , code from the source name
trip['source_state'] = trip['source_name'].apply(lambda x: get_state(x))
trip['source_city'] = trip['source_name'].apply(lambda x: get_city(x))
trip['source_place'] = trip['source_name'].apply(lambda x: place2city_place(x))
trip['source_code'] = trip['source_name'].apply(lambda x: get_code(x))

trip[['source_state', 'source_city', 'source_place', 'source_code']]

```

```

[46]:      source_state source_city source_place source_code
0          uttar pradesh          kanpur          central          6
1            karnataka      doddablpur      chikadpp          d
2            haryana        gurgaon      bilaspur          hb
3      maharashtra      mumbai hub          mumbai          none
4            karnataka        bellary        bellary          none
...
14782         punjab      chandigarh      mehmdpur          h
14783         haryana          fbd      balabhgarh          dpc
14784      uttar pradesh          kanpur      govndngr          dc
14785         tamil nadu      tirunelveli      vdckusrt          i
14786         karnataka        sandur      wrdn1dpp          d

```

[14787 rows x 4 columns]

Univariate - Categorical Data

```
[77]: trip_states = trip[['source_state']].value_counts().reset_index()
trip_states
```

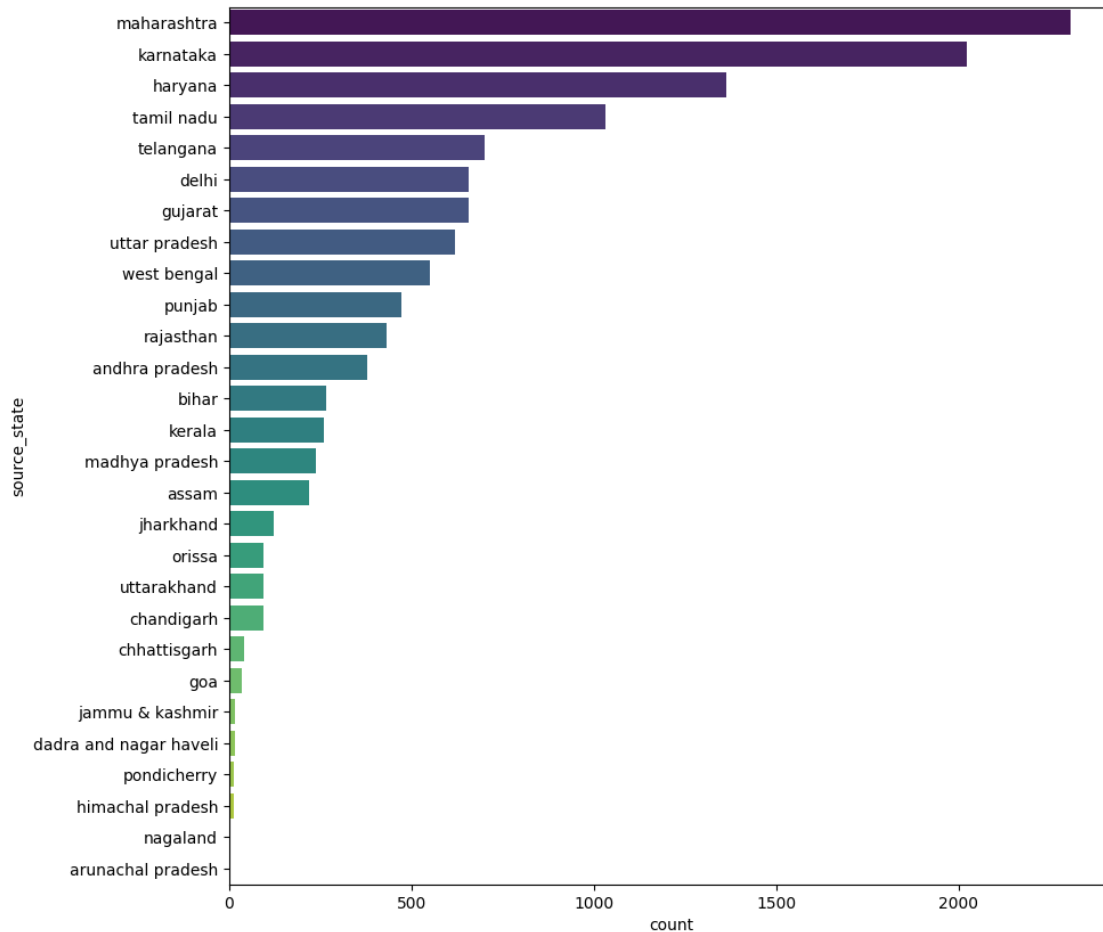
```
[77]:
```

	source_state	count
0	maharashtra	2308
1	karnataka	2025
2	haryana	1365
3	tamil nadu	1032
4	telangana	701
5	delhi	658
6	gujarat	656
7	uttar pradesh	619
8	west bengal	551
9	punjab	472
10	rajasthan	431
11	andhra pradesh	378
12	bihar	267
13	kerala	261
14	madhya pradesh	238
15	assam	220
16	jharkhand	123
17	orissa	94
18	uttarakhand	93
19	chandigarh	93
20	chhattisgarh	42
21	goa	34
22	jammu & kashmir	16
23	dadra and nagar haveli	15
24	pondicherry	12
25	himachal pradesh	12
26	nagaland	4
27	arunachal pradesh	3

Visualization of above chart

```
[81]: plt.figure(figsize=(10,10))
sns.barplot(data = trip_states, y = 'source_state', x = 'count',
palette='viridis')
```

```
[81]: <Axes: xlabel='count', ylabel='source_state'>
```



From the above chart, we get to know that from Maharashtra, maximum trip starts. North, South and West Zones corridors have significant traffic of orders. But, have a smaller presence in Central, Eastern and North-Eastern zone

```
[47]: # getting the separate columns of year, month, hour, day , week, day of week,
      ↪from trip_creation_time
trip['trip_year'] = trip['trip_creation_time'].dt.year
trip['trip_month'] = trip['trip_creation_time'].dt.month
trip['trip_hour'] = trip['trip_creation_time'].dt.hour
trip['trip_day'] = trip['trip_creation_time'].dt.day
trip['trip_week'] = trip['trip_creation_time'].dt.isocalendar().week
trip['trip_dayofweek'] = trip['trip_creation_time'].dt.dayofweek

trip[['trip_year', 'trip_month', 'trip_hour', 'trip_day', 'trip_week',
      ↪'trip_dayofweek']]
```

```
[47]:      trip_year  trip_month  trip_hour  trip_day  trip_week  trip_dayofweek
0          2018           9           0         12          37           2
```

1	2018	9	0	12	37	2
2	2018	9	0	12	37	2
3	2018	9	0	12	37	2
4	2018	9	0	12	37	2
...
14782	2018	10	23	3	40	2
14783	2018	10	23	3	40	2
14784	2018	10	23	3	40	2
14785	2018	10	23	3	40	2
14786	2018	10	23	3	40	2

[14787 rows x 6 columns]

```
[48]: # numerical columns
num_col = ['start_scan_to_end_scan', 'actual_distance_to_destination',
↳ 'actual_time', 'osrm_time',
↳ 'osrm_distance', 'segment_actual_time_sum',
↳ 'segment_osrm_distance_sum',
↳ 'segment_osrm_time_sum', 'hour_taken']

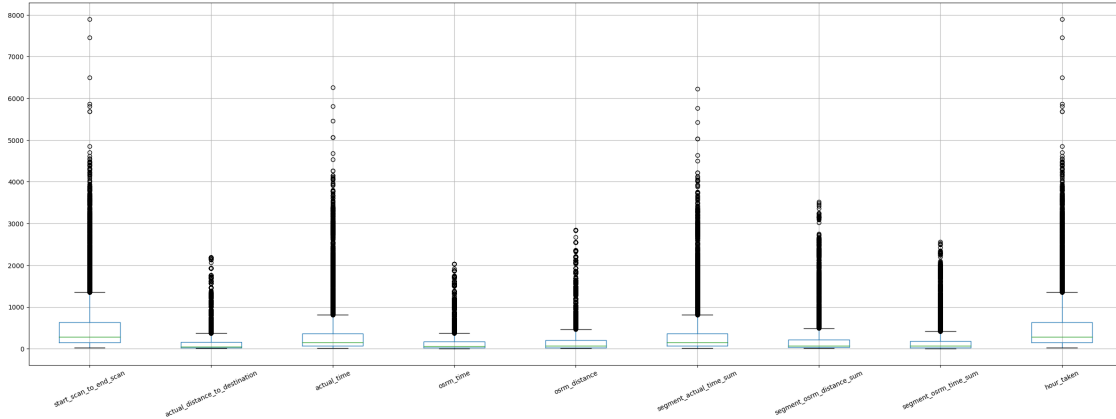
num_col
```

```
[48]: ['start_scan_to_end_scan',
'actual_distance_to_destination',
'actual_time',
'osrm_time',
'osrm_distance',
'segment_actual_time_sum',
'segment_osrm_distance_sum',
'segment_osrm_time_sum',
'hour_taken']
```

2.3 Find outliers in numerical variable

```
[49]: trip[num_col].boxplot(rot=25, figsize=(30,10))
```

```
[49]: <Axes: >
```

as we can observe, every numerical column has outliers

2.4 Handle the outliers using IQR method

```
[50]: Q1 = trip[num_col].quantile(0.25) #Calculates the first quartile (25th
      ↪percentile)
      Q3 = trip[num_col].quantile(0.75) # Calculates the third quartile (75th
      ↪percentile)
      IQR = Q3 - Q1 # Computes the Interquartile Range

      # Filtering out the Outliers, keeping the rows that are not outliers from trip
      ↪dataframe

      trip = trip[~((trip[num_col] < (Q1 - 1.5 * IQR)) | (trip[num_col] > (Q3 + 1.5 *
      ↪IQR))).any(axis=1)]
      trip = trip.reset_index(drop=True)

      trip.head( 3)
```

```
[50]:      data      trip_creation_time \
0  training 2018-09-12 00:00:22.886430
1  training 2018-09-12 00:01:00.113710
2  training 2018-09-12 00:02:09.740725

      route_schedule_uuid route_type \
0  thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...   Carting
1  thanos::sroute:f0176492-a679-4597-8332-bbd1c7f...   Carting
2  thanos::sroute:d9f07b12-65e0-4f3b-bec8-df06134...     FTL

      trip_uuid source_center      source_name \
0  trip-153671042288605164  IND561203AAB  doddablpur_chikadpp_d (karnataka)
1  trip-153671046011330457  IND400072AAB      mumbai hub (maharashtra)
```

```
2 trip-153671052974046625 IND583101AAA bellary_dc (karnataka)
```

```

destination_center      destination_name \
0      IND561203AAB  doddablpur_chikadpp_d (karnataka)
1      IND401104AAA   mumbai_mirard_ip (maharashtra)
2      IND583119AAA   sandur_wrdn1dpp_d (karnataka)

start_scan_to_end_scan ... source_state source_city source_place \
0      180.0 ...      karnataka  doddablpur      chikadpp
1      100.0 ...      maharashtra mumbai hub      mumbai
2      717.0 ...      karnataka   bellary      bellary

source_code  trip_year  trip_month  trip_hour  trip_day  trip_week \
0      d      2018      9      0      12      37
1      none   2018      9      0      12      37
2      none   2018      9      0      12      37

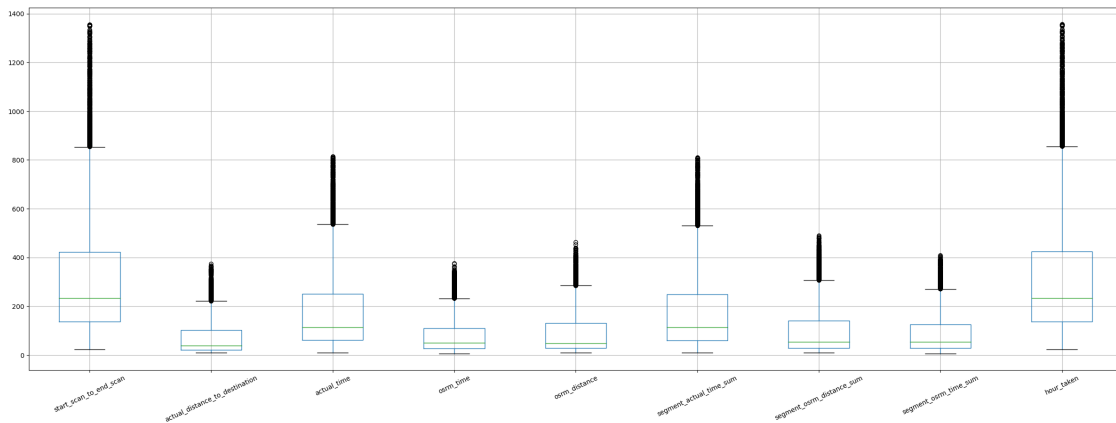
trip_dayofweek
0      2
1      2
2      2

```

```
[3 rows x 32 columns]
```

```
[51]: # getting the chart after filtering the trip dataframe
trip[num_col].boxplot(rot=25, figsize=(30,10))
```

```
[51]: <Axes: >
```



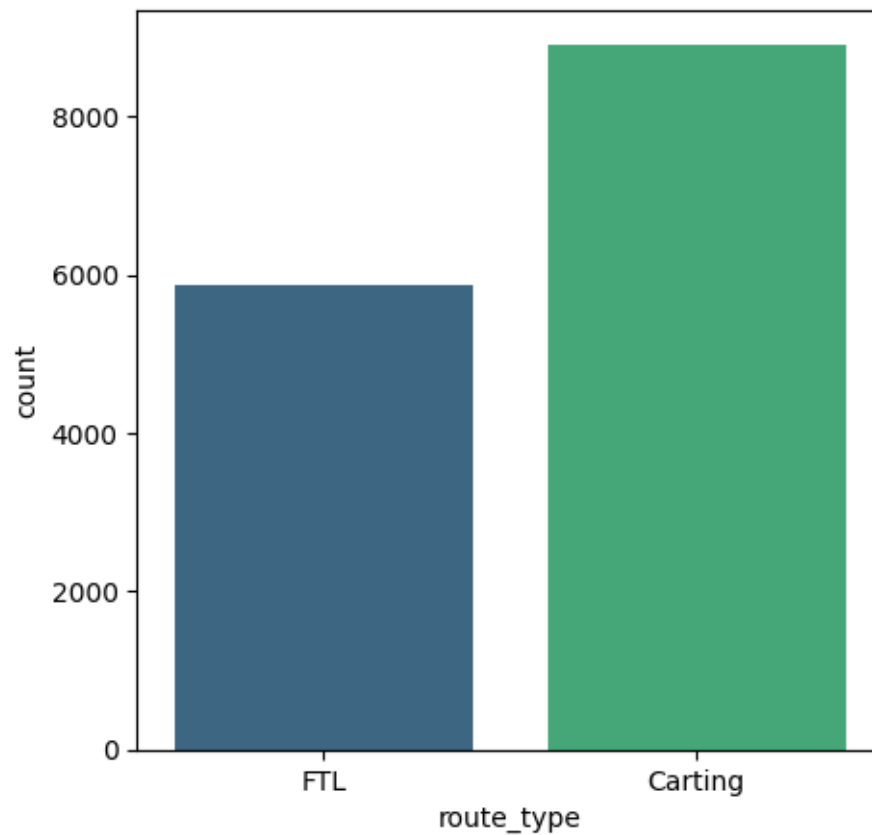
2.5 Categorical Variables

```
[89]: # getting the count of two types of route types  
trip['route_type'].value_counts()
```

```
[89]: route_type  
Carting    8906  
FTL        5881  
Name: count, dtype: int64
```

```
[91]: # visualization of above chart  
plt.figure(figsize=(5,5))  
sns.countplot(data = trip, x = 'route_type',palette='viridis')
```

```
[91]: <Axes: xlabel='route_type', ylabel='count'>
```



```
[56]: #mapping the 2 trpes of routes  
trip['route_type'] = trip['route_type'].map({'FTL':0, 'Carting':1})
```

```
[58]: trip
```

```

[58]:      data      trip_creation_time  \
0      training 2018-09-12 00:00:22.886430
1      training 2018-09-12 00:01:00.113710
2      training 2018-09-12 00:02:09.740725
3      training 2018-09-12 00:02:34.161600
4      training 2018-09-12 00:04:22.011653
...
12718      test 2018-10-03 23:55:56.258533
12719      test 2018-10-03 23:57:23.863155
12720      test 2018-10-03 23:57:44.429324
12721      test 2018-10-03 23:59:14.390954
12722      test 2018-10-03 23:59:42.701692

      route_schedule_uuid route_type  \
0      thanos::sroute:3a1b0ab2-bb0b-4c53-8c59-eb2a2c0...      NaN
1      thanos::sroute:f0176492-a679-4597-8332-bbd1c7f...      NaN
2      thanos::sroute:d9f07b12-65e0-4f3b-bec8-df06134...      NaN
3      thanos::sroute:9bf03170-d0a2-4a3f-aa4d-9aaab3d...      NaN
4      thanos::sroute:a97698cc-846e-41a7-916b-88b1741...      NaN
...
12718      thanos::sroute:8a120994-f577-4491-9e4b-b7e4a14...      NaN
12719      thanos::sroute:b30e1ec3-3bfa-4bd2-a7fb-3b75769...      NaN
12720      thanos::sroute:5609c268-e436-4e0a-8180-3db4a74...      NaN
12721      thanos::sroute:c5f2ba2c-8486-4940-8af6-d1d2a6a...      NaN
12722      thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...      NaN

      trip_uuid source_center  \
0      trip-153671042288605164      IND561203AAB
1      trip-153671046011330457      IND400072AAB
2      trip-153671052974046625      IND583101AAA
3      trip-153671055416136166      IND600056AAA
4      trip-153671066201138152      IND600044AAD
...
12718      trip-153861095625827784      IND160002AAC
12719      trip-153861104386292051      IND121004AAB
12720      trip-153861106442901555      IND208006AAA
12721      trip-153861115439069069      IND627005AAA
12722      trip-153861118270144424      IND583119AAA

      source_name destination_center  \
0      doddablpur_chikadpp_d (karnataka)      IND561203AAB
1      mumbai hub (maharashtra)      IND401104AAA
2      bellary_dc (karnataka)      IND583119AAA
3      chennai_poonamallee (tamil nadu)      IND600056AAA
4      chennai_chrompet_dpc (tamil nadu)      IND600048AAA
...
12718      chandigarh_mehmdpur_h (punjab)      IND160002AAC

```

12719	fbd_balabhgarh_dpc (haryana)	IND121004AAA
12720	kanpur_govndngr_dc (uttar pradesh)	IND208006AAA
12721	tirunelveli_vdkkusrt_i (tamil nadu)	IND628204AAA
12722	sandur_wrdn1dpp_d (karnataka)	IND583119AAA

	destination_name	start_scan_to_end_scan	...	\
0	doddablpur_chikadpp_d (karnataka)	180.0	...	
1	mumbai_mirard_ip (maharashtra)	100.0	...	
2	sandur_wrdn1dpp_d (karnataka)	717.0	...	
3	chennai_poonamallee (tamil nadu)	189.0	...	
4	chennai_vandalur_dc (tamil nadu)	98.0	...	
...	
12718	chandigarh_mehmdpur_h (punjab)	257.0	...	
12719	faridabad_blbgarh_dc (haryana)	60.0	...	
12720	kanpur_govndngr_dc (uttar pradesh)	421.0	...	
12721	tirchchndr_shnmgprm_d (tamil nadu)	347.0	...	
12722	sandur_wrdn1dpp_d (karnataka)	353.0	...	

	source_state	source_city	source_place	source_code	trip_year	\
0	karnataka	doddablpur	chikadpp	d	2018	
1	maharashtra	mumbai hub	mumbai	none	2018	
2	karnataka	bellary	bellary	none	2018	
3	tamil nadu	chennai	chennai	none	2018	
4	tamil nadu	chennai	chrompet	dpc	2018	
...	
12718	punjab	chandigarh	mehmdpur	h	2018	
12719	haryana	fbd	balabhgarh	dpc	2018	
12720	uttar pradesh	kanpur	govndngr	dc	2018	
12721	tamil nadu	tirunelveli	vdkkusrt	i	2018	
12722	karnataka	sandur	wrdn1dpp	d	2018	

	trip_month	trip_hour	trip_day	trip_week	trip_dayofweek
0	9	0	12	37	2
1	9	0	12	37	2
2	9	0	12	37	2
3	9	0	12	37	2
4	9	0	12	37	2
...
12718	10	23	3	40	2
12719	10	23	3	40	2
12720	10	23	3	40	2
12721	10	23	3	40	2
12722	10	23	3	40	2

[12723 rows x 32 columns]

2.5.1 Standardize the numerical features using StandardScaler

```
[61]: from sklearn.preprocessing import StandardScaler

# standardize features and calculates the mean and standard deviation for each
# of these columns.
scaler = StandardScaler()
scaler.fit(trip[num_col])
```

```
[61]: StandardScaler()
```

```
[63]: #transform the specified numerical columns, making them have a mean of 0 and a
# standard deviation of 1
trip[num_col] = scaler.transform(trip[num_col])
trip[num_col]
```

```
[63]:
```

	start_scan_to_end_scan	actual_distance_to_destination	actual_time	\
0	-1.255068	-1.003307	-1.123469	
1	-1.256293	-1.014092	-1.126828	
2	-1.246845	-0.992860	-1.115552	
3	-1.254930	-1.012663	-1.126748	
4	-1.256323	-1.015647	-1.128227	
...	
12718	-1.253889	-1.006277	-1.125868	
12719	-1.256905	-1.014412	-1.128347	
12720	-1.251377	-1.009950	-1.117911	
12721	-1.252510	-0.991459	-1.118631	
12722	-1.252419	-1.004675	-1.118191	

	osrm_time	osrm_distance	segment_actual_time_sum	\
0	-1.086461	-1.025065	-1.120464	
1	-1.096592	-1.033222	-1.123787	
2	-1.077095	-1.017376	-1.112401	
3	-1.095063	-1.032177	-1.123747	
4	-1.096974	-1.034177	-1.125205	
...	
12718	-1.087608	-1.026517	-1.122855	
12719	-1.097165	-1.033670	-1.125327	
12720	-1.090284	-1.028332	-1.114791	
12721	-1.065245	-1.014344	-1.115723	
12722	-1.086461	-1.025630	-1.115075	

	segment_osrm_distance_sum	segment_osrm_time_sum	hour_taken
0	-1.034821	-1.082525	-1.256735
1	-1.041975	-1.090258	-1.257974
2	-1.027858	-1.074634	-1.248537
3	-1.041064	-1.089154	-1.256599

4	-1.042849	-1.090732	-1.258012
...
12718	-1.036972	-1.082999	-1.255568
12719	-1.042397	-1.091047	-1.258583
12720	-1.032519	-1.078896	-1.253062
12721	-1.019322	-1.057906	-1.254186
12722	-1.035223	-1.082210	-1.254096

[12723 rows x 9 columns]

```
[65]: #get a statistical summary of a specific numerical columns in the DataFrame trip
trip[num_col].describe()
```

```
[65]:
```

	start_scan_to_end_scan	actual_distance_to_destination	actual_time \
count	12723.000000	12723.000000	12723.000000
mean	-1.252921	-1.003475	-1.122091
std	0.003913	0.013876	0.006324
min	-1.257472	-1.015665	-1.128827
25%	-1.255742	-1.013279	-1.126748
50%	-1.254256	-1.009981	-1.124629
75%	-1.251347	-0.997823	-1.119151
max	-1.237075	-0.945496	-1.096599

	osrm_time	osrm_distance	segment_actual_time_sum \
count	12723.000000	12723.000000	12723.000000
mean	-1.084466	-1.024240	-1.119054
std	0.013826	0.011166	0.006366
min	-1.098312	-1.034545	-1.125813
25%	-1.094298	-1.032142	-1.123747
50%	-1.089902	-1.029640	-1.121599
75%	-1.078624	-1.019305	-1.116129
max	-1.027590	-0.977896	-1.093356

	segment_osrm_distance_sum	segment_osrm_time_sum	hour_taken
count	12723.000000	12723.000000	12723.000000
mean	-1.033288	-1.079226	-1.254606
std	0.010547	0.012563	0.003908
min	-1.043177	-1.091837	-1.259151
25%	-1.040912	-1.088365	-1.257424
50%	-1.038169	-1.084419	-1.255942
75%	-1.028554	-1.073056	-1.253034
max	-0.989729	-1.028394	-1.238776

There is a significant difference between OSRM and actual parameters.

3 Recommendation

1. There is a significant difference between OSRM and actual parameters.
2. We need to check information fed to routing engine for trip planning.
3. North, South and West Zones have significant numbers of orders.
4. we need to increasing our presence in Central, Eastern and North-Eastern zone. As we have small presence in these area.
5. we have maximum number of orders in Maharashtra followed by Karnataka.
6. we need to prepare for resources on ground level in these states on festivals

[]: