

Delhivery__Business__CASE__

September 24, 2024

```
[6]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from IPython.core.display import display, HTML
display(HTML("<style>.container { width:100% !important; }</style>"))
from IPython.core.display import display, HTML
display(HTML("<style>.container { width:100% !important; }</style>"))
```

<IPython.core.display.HTML object>

<IPython.core.display.HTML object>

```
[7]: df=pd.read_csv('/content/sample_data/delhivery_data.csv')
df.head(20)
```

```
[7]:      data      trip_creation_time \
0  training  2018-09-20 02:35:36.476840
1  training  2018-09-20 02:35:36.476840
2  training  2018-09-20 02:35:36.476840
3  training  2018-09-20 02:35:36.476840
4  training  2018-09-20 02:35:36.476840
5  training  2018-09-20 02:35:36.476840
6  training  2018-09-20 02:35:36.476840
7  training  2018-09-20 02:35:36.476840
8  training  2018-09-20 02:35:36.476840
9  training  2018-09-20 02:35:36.476840
10 training  2018-09-23 06:42:06.021680
11 training  2018-09-23 06:42:06.021680
12 training  2018-09-23 06:42:06.021680
13 training  2018-09-23 06:42:06.021680
14 training  2018-09-23 06:42:06.021680
15 training  2018-09-14 15:42:46.437249
16 training  2018-09-14 15:42:46.437249
17 training  2018-09-13 20:44:19.424489
18 training  2018-09-13 20:44:19.424489
19 training  2018-09-13 20:44:19.424489
```

	route_schedule_uuid	route_type \
0	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
1	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
2	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
3	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
4	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
5	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
6	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
7	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
8	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
9	thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3...	Carting
10	thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172...	FTL
11	thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172...	FTL
12	thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172...	FTL
13	thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172...	FTL
14	thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172...	FTL
15	thanos::sroute:a16bfa03-3462-4bce-9c82-5784c7d...	Carting
16	thanos::sroute:a16bfa03-3462-4bce-9c82-5784c7d...	Carting
17	thanos::sroute:76951383-1608-44e4-a284-46d92e8...	FTL
18	thanos::sroute:76951383-1608-44e4-a284-46d92e8...	FTL
19	thanos::sroute:76951383-1608-44e4-a284-46d92e8...	FTL

	trip_uuid	source_center	source_name \
0	trip-153741093647649320	IND388121AAA	Anand_VUNagar_DC (Gujarat)
1	trip-153741093647649320	IND388121AAA	Anand_VUNagar_DC (Gujarat)
2	trip-153741093647649320	IND388121AAA	Anand_VUNagar_DC (Gujarat)
3	trip-153741093647649320	IND388121AAA	Anand_VUNagar_DC (Gujarat)
4	trip-153741093647649320	IND388121AAA	Anand_VUNagar_DC (Gujarat)
5	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
6	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
7	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
8	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
9	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
10	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
11	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
12	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
13	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
14	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
15	trip-153693976643699843	IND400011AAA	LowerParel_CP (Maharashtra)
16	trip-153693976643699843	IND400011AAA	LowerParel_CP (Maharashtra)
17	trip-153687145942424248	IND562132AAA	Bangalore_Nelmngla_H (Karnataka)
18	trip-153687145942424248	IND562132AAA	Bangalore_Nelmngla_H (Karnataka)
19	trip-153687145942424248	IND560099AAB	Bengaluru_Bomsndra_HB (Karnataka)

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

2	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
3	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
4	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
5	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
6	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
7	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
8	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
9	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
10	IND411033AAA	Pune_Tathawde_H (Maharashtra)
11	IND411033AAA	Pune_Tathawde_H (Maharashtra)
12	IND411033AAA	Pune_Tathawde_H (Maharashtra)
13	IND411033AAA	Pune_Tathawde_H (Maharashtra)
14	IND411033AAA	Pune_Tathawde_H (Maharashtra)
15	IND400072AAD	Mumbai_Chndivli_PC (Maharashtra)
16	IND400072AAD	Mumbai_Chndivli_PC (Maharashtra)
17	IND560099AAB	Bengaluru_Bomsndra_HB (Karnataka)
18	IND560099AAB	Bengaluru_Bomsndra_HB (Karnataka)
19	IND683511AAA	Aluva_Peedika_H (Kerala)

	od_start_time	...	cutoff_timestamp	\
0	2018-09-20 03:21:32.418600	...	2018-09-20 04:27:55	
1	2018-09-20 03:21:32.418600	...	2018-09-20 04:17:55	
2	2018-09-20 03:21:32.418600	...	2018-09-20 04:01:19.505586	
3	2018-09-20 03:21:32.418600	...	2018-09-20 03:39:57	
4	2018-09-20 03:21:32.418600	...	2018-09-20 03:33:55	
5	2018-09-20 04:47:45.236797	...	2018-09-20 06:15:58	
6	2018-09-20 04:47:45.236797	...	2018-09-20 05:47:29	
7	2018-09-20 04:47:45.236797	...	2018-09-20 05:25:58	
8	2018-09-20 04:47:45.236797	...	2018-09-20 05:15:56	
9	2018-09-20 04:47:45.236797	...	2018-09-20 04:49:20	
10	2018-09-23 06:42:06.021680	...	2018-09-23 11:05:19	
11	2018-09-23 06:42:06.021680	...	2018-09-23 10:27:22	
12	2018-09-23 06:42:06.021680	...	2018-09-23 09:45:25	
13	2018-09-23 06:42:06.021680	...	2018-09-23 09:21:27	
14	2018-09-23 06:42:06.021680	...	2018-09-23 08:39:31	
15	2018-09-14 15:42:46.437249	...	2018-09-14 16:29:54	
16	2018-09-14 15:42:46.437249	...	2018-09-14 16:15:53	
17	2018-09-13 20:44:19.424489	...	2018-09-13 23:25:20	
18	2018-09-13 20:44:19.424489	...	2018-09-13 22:47:26	
19	2018-09-13 23:59:56.061158	...	2018-09-14 12:45:25	

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

5	10.403038	15.0	11.0	12.1171
6	18.045481	44.0	17.0	21.2890
7	28.061896	65.0	29.0	35.8252
8	38.939167	76.0	39.0	47.1900
9	43.595802	102.0	45.0	53.2334
10	23.194334	38.0	24.0	26.8622
11	44.045659	76.0	41.0	54.4326
12	72.849327	117.0	68.0	89.6680
13	88.076599	141.0	80.0	108.3939
14	100.708423	183.0	95.0	129.3519
15	9.355852	46.0	11.0	11.4344
16	16.431273	60.0	16.0	18.7941
17	23.635811	30.0	30.0	28.9765
18	39.806036	67.0	53.0	52.1256
19	24.319864	50.0	24.0	29.7046

	factor	segment_actual_time	segment_osrm_time	segment_osrm_distance	\
0	1.272727	14.0	11.0	11.9653	
1	1.200000	10.0	9.0	9.7590	
2	1.428571	16.0	7.0	10.8152	
3	1.550000	21.0	12.0	13.0224	
4	1.545455	6.0	5.0	3.9153	
5	1.363636	15.0	11.0	12.1171	
6	2.588235	28.0	6.0	9.1719	
7	2.241379	21.0	11.0	14.5362	
8	1.948718	10.0	10.0	11.3648	
9	2.266667	26.0	6.0	6.0434	
10	1.583333	38.0	24.0	26.8622	
11	1.853659	37.0	27.0	30.1058	
12	1.720588	41.0	26.0	35.2353	
13	1.762500	23.0	14.0	17.2476	
14	1.926316	41.0	15.0	20.9580	
15	4.181818	46.0	11.0	11.4344	
16	3.750000	14.0	5.0	7.3597	
17	1.000000	30.0	30.0	28.9765	
18	1.264151	37.0	26.0	24.9545	
19	2.083333	50.0	24.0	29.7046	

	segment_factor
0	1.272727
1	1.111111
2	2.285714
3	1.750000
4	1.200000
5	1.363636
6	4.666667
7	1.909091

```

8      1.000000
9      4.333333
10     1.583333
11     1.370370
12     1.576923
13     1.642857
14     2.733333
15     4.181818
16     2.800000
17     1.000000
18     1.423077
19     2.083333

```

[20 rows x 24 columns]

```

[8]: df[['route_schedule_uuid', 'route_type', 'trip_uuid', 'source_center', 'source_name',
        'destination_center', 'destination_name', 'od_start_time', 'od_end_time',
        'start_scan_to_end_scan', 'is_cutoff', 'actual_distance_to_destination', 'actual_time', 'segment_id']]
      head(20)

```

```

[8]:
   route_schedule_uuid route_type \
0  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
1  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
2  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
3  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
4  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
5  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
6  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
7  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
8  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
9  thanos::sroute:eb7bfc78-b351-4c0e-a951-fa3d5c3... Carting
10 thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172... FTL
11 thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172... FTL
12 thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172... FTL
13 thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172... FTL
14 thanos::sroute:ff52ef7a-4d0d-4063-9bfe-cc21172... FTL
15 thanos::sroute:a16bfa03-3462-4bce-9c82-5784c7d... Carting
16 thanos::sroute:a16bfa03-3462-4bce-9c82-5784c7d... Carting
17 thanos::sroute:76951383-1608-44e4-a284-46d92e8... FTL
18 thanos::sroute:76951383-1608-44e4-a284-46d92e8... FTL
19 thanos::sroute:76951383-1608-44e4-a284-46d92e8... FTL

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

```

3	trip-153741093647649320	IND388121AAA	Anand_VUNagar_DC (Gujarat)
4	trip-153741093647649320	IND388121AAA	Anand_VUNagar_DC (Gujarat)
5	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
6	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
7	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
8	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
9	trip-153741093647649320	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
10	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
11	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
12	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
13	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
14	trip-153768492602129387	IND421302AAG	Bhiwandi_Mankoli_HB (Maharashtra)
15	trip-153693976643699843	IND400011AAA	LowerParel_CP (Maharashtra)
16	trip-153693976643699843	IND400011AAA	LowerParel_CP (Maharashtra)
17	trip-153687145942424248	IND562132AAA	Bangalore_Nelmngla_H (Karnataka)
18	trip-153687145942424248	IND562132AAA	Bangalore_Nelmngla_H (Karnataka)
19	trip-153687145942424248	IND560099AAB	Bengaluru_Bomsndra_HB (Karnataka)

	destination_center	destination_name \
0	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
1	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
2	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
3	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
4	IND388620AAB	Khambhat_MotvdDPP_D (Gujarat)
5	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
6	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
7	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
8	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
9	IND388320AAA	Anand_Vaghasi_IP (Gujarat)
10	IND411033AAA	Pune_Tathawde_H (Maharashtra)
11	IND411033AAA	Pune_Tathawde_H (Maharashtra)
12	IND411033AAA	Pune_Tathawde_H (Maharashtra)
13	IND411033AAA	Pune_Tathawde_H (Maharashtra)
14	IND411033AAA	Pune_Tathawde_H (Maharashtra)
15	IND400072AAD	Mumbai_Chndivli_PC (Maharashtra)
16	IND400072AAD	Mumbai_Chndivli_PC (Maharashtra)
17	IND560099AAB	Bengaluru_Bomsndra_HB (Karnataka)
18	IND560099AAB	Bengaluru_Bomsndra_HB (Karnataka)
19	IND683511AAA	Aluva_Peedika_H (Kerala)

	od_start_time	od_end_time \
0	2018-09-20 03:21:32.418600	2018-09-20 04:47:45.236797
1	2018-09-20 03:21:32.418600	2018-09-20 04:47:45.236797
2	2018-09-20 03:21:32.418600	2018-09-20 04:47:45.236797
3	2018-09-20 03:21:32.418600	2018-09-20 04:47:45.236797
4	2018-09-20 03:21:32.418600	2018-09-20 04:47:45.236797
5	2018-09-20 04:47:45.236797	2018-09-20 06:36:55.627764

6	2018-09-20 04:47:45.236797	2018-09-20 06:36:55.627764
7	2018-09-20 04:47:45.236797	2018-09-20 06:36:55.627764
8	2018-09-20 04:47:45.236797	2018-09-20 06:36:55.627764
9	2018-09-20 04:47:45.236797	2018-09-20 06:36:55.627764
10	2018-09-23 06:42:06.021680	2018-09-23 11:44:28.365845
11	2018-09-23 06:42:06.021680	2018-09-23 11:44:28.365845
12	2018-09-23 06:42:06.021680	2018-09-23 11:44:28.365845
13	2018-09-23 06:42:06.021680	2018-09-23 11:44:28.365845
14	2018-09-23 06:42:06.021680	2018-09-23 11:44:28.365845
15	2018-09-14 15:42:46.437249	2018-09-14 17:31:45.368791
16	2018-09-14 15:42:46.437249	2018-09-14 17:31:45.368791
17	2018-09-13 20:44:19.424489	2018-09-13 23:59:56.061158
18	2018-09-13 20:44:19.424489	2018-09-13 23:59:56.061158
19	2018-09-13 23:59:56.061158	2018-09-14 13:55:58.765334

	start_scan_to_end_scan	is_cutoff	actual_distance_to_destination \
0	86.0	True	10.435660
1	86.0	True	18.936842
2	86.0	True	27.637279
3	86.0	True	36.118028
4	86.0	False	39.386040
5	109.0	True	10.403038
6	109.0	True	18.045481
7	109.0	True	28.061896
8	109.0	True	38.939167
9	109.0	False	43.595802
10	302.0	True	23.194334
11	302.0	True	44.045659
12	302.0	True	72.849327
13	302.0	True	88.076599
14	302.0	False	100.708423
15	108.0	True	9.355852
16	108.0	False	16.431273
17	195.0	True	23.635811
18	195.0	False	39.806036
19	836.0	True	24.319864

	actual_time	segment_osrm_time	segment_osrm_distance
0	14.0	11.0	11.9653
1	24.0	9.0	9.7590
2	40.0	7.0	10.8152
3	62.0	12.0	13.0224
4	68.0	5.0	3.9153
5	15.0	11.0	12.1171
6	44.0	6.0	9.1719
7	65.0	11.0	14.5362
8	76.0	10.0	11.3648

9	102.0	6.0	6.0434
10	38.0	24.0	26.8622
11	76.0	27.0	30.1058
12	117.0	26.0	35.2353
13	141.0	14.0	17.2476
14	183.0	15.0	20.9580
15	46.0	11.0	11.4344
16	60.0	5.0	7.3597
17	30.0	30.0	28.9765
18	67.0	26.0	24.9545
19	50.0	24.0	29.7046

```
[9]: sample=df[(df['route_schedule_uuid']=='thanos::sroute:
↳eb7bfc78-b351-4c0e-a951-fa3d5c3297ef') &
↳(df['trip_uuid']=='trip-153741093647649320')]
sample[['source_name','destination_name','od_start_time','od_start_time','od_end_time','start_
```

```
[9]:
```

	source_name	destination_name \
0	Anand_VUNagar_DC (Gujarat)	Khambhat_MotvdDPP_D (Gujarat)
1	Anand_VUNagar_DC (Gujarat)	Khambhat_MotvdDPP_D (Gujarat)
2	Anand_VUNagar_DC (Gujarat)	Khambhat_MotvdDPP_D (Gujarat)
3	Anand_VUNagar_DC (Gujarat)	Khambhat_MotvdDPP_D (Gujarat)
4	Anand_VUNagar_DC (Gujarat)	Khambhat_MotvdDPP_D (Gujarat)
5	Khambhat_MotvdDPP_D (Gujarat)	Anand_Vaghasi_IP (Gujarat)
6	Khambhat_MotvdDPP_D (Gujarat)	Anand_Vaghasi_IP (Gujarat)
7	Khambhat_MotvdDPP_D (Gujarat)	Anand_Vaghasi_IP (Gujarat)
8	Khambhat_MotvdDPP_D (Gujarat)	Anand_Vaghasi_IP (Gujarat)
9	Khambhat_MotvdDPP_D (Gujarat)	Anand_Vaghasi_IP (Gujarat)

	od_start_time	od_start_time \
0	2018-09-20 03:21:32.418600	2018-09-20 03:21:32.418600
1	2018-09-20 03:21:32.418600	2018-09-20 03:21:32.418600
2	2018-09-20 03:21:32.418600	2018-09-20 03:21:32.418600
3	2018-09-20 03:21:32.418600	2018-09-20 03:21:32.418600
4	2018-09-20 03:21:32.418600	2018-09-20 03:21:32.418600
5	2018-09-20 04:47:45.236797	2018-09-20 04:47:45.236797
6	2018-09-20 04:47:45.236797	2018-09-20 04:47:45.236797
7	2018-09-20 04:47:45.236797	2018-09-20 04:47:45.236797
8	2018-09-20 04:47:45.236797	2018-09-20 04:47:45.236797
9	2018-09-20 04:47:45.236797	2018-09-20 04:47:45.236797

	od_end_time	start_scan_to_end_scan	is_cutoff \
0	2018-09-20 04:47:45.236797	86.0	True
1	2018-09-20 04:47:45.236797	86.0	True
2	2018-09-20 04:47:45.236797	86.0	True
3	2018-09-20 04:47:45.236797	86.0	True
4	2018-09-20 04:47:45.236797	86.0	False

5	2018-09-20 06:36:55.627764	109.0	True
6	2018-09-20 06:36:55.627764	109.0	True
7	2018-09-20 06:36:55.627764	109.0	True
8	2018-09-20 06:36:55.627764	109.0	True
9	2018-09-20 06:36:55.627764	109.0	False

	actual_distance_to_destination	actual_time	segment_osrm_time \
0	10.435660	14.0	11.0
1	18.936842	24.0	9.0
2	27.637279	40.0	7.0
3	36.118028	62.0	12.0
4	39.386040	68.0	5.0
5	10.403038	15.0	11.0
6	18.045481	44.0	6.0
7	28.061896	65.0	11.0
8	38.939167	76.0	10.0
9	43.595802	102.0	6.0

	segment_osrm_distance
0	11.9653
1	9.7590
2	10.8152
3	13.0224
4	3.9153
5	12.1171
6	9.1719
7	14.5362
8	11.3648
9	6.0434

```
[10]: df.shape
```

```
[10]: (27323, 24)
```

```
[11]: df.dtypes
```

```
[11]: 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                   object
cutoff_factor               float64
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

```

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

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 27323 entries, 0 to 27322
Data columns (total 24 columns):
 #   Column                                Non-Null Count  Dtype
---  -
 0   data                                  27323 non-null  object
 1   trip_creation_time                    27323 non-null  object
 2   route_schedule_uuid                  27323 non-null  object
 3   route_type                           27323 non-null  object
 4   trip_uuid                            27323 non-null  object
 5   source_center                        27323 non-null  object
 6   source_name                          27244 non-null  object
 7   destination_center                   27322 non-null  object
 8   destination_name                     27267 non-null  object
 9   od_start_time                        27322 non-null  object
10   od_end_time                          27322 non-null  object
11   start_scan_to_end_scan                27322 non-null  float64
12   is_cutoff                            27322 non-null  object
13   cutoff_factor                        27322 non-null  float64
14   cutoff_timestamp                     27322 non-null  object
15   actual_distance_to_destination        27322 non-null  float64
16   actual_time                          27322 non-null  float64
17   osrm_time                            27322 non-null  float64
18   osrm_distance                        27322 non-null  float64
19   factor                              27322 non-null  float64
20   segment_actual_time                  27322 non-null  float64
21   segment_osrm_time                    27322 non-null  float64
22   segment_osrm_distance                 27322 non-null  float64
23   segment_factor                       27322 non-null  float64
dtypes: float64(11), object(13)

```

memory usage: 5.0+ MB

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

```
[13]:
```

	data	trip_creation_time	\
count	27323	27323	
unique	2	2852	
top	training	2018-09-28 05:23:15.359220	
freq	19801	101	

	route_schedule_uuid	route_type	\
count	27323	27323	
unique	1102	2	
top	thanos::sroute:0456b740-1dad-4929-bbe0-87d8843...	FTL	
freq	462	18559	

	trip_uuid	source_center	source_name	\
count	27323	27323	27244	
unique	2852	1169	1164	
top	trip-153811219535896559	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	
freq	101	4699	4699	

	destination_center	destination_name	\
count	27322	27267	
unique	1155	1148	
top	IND000000ACB	Gurgaon_Bilaspur_HB (Haryana)	
freq	2231	2231	

	od_start_time	od_end_time	is_cutoff	\
count	27322	27322	27322	
unique	5188	5188	2	
top	2018-10-03 04:55:30.039225	2018-10-05 11:15:01.115906	True	
freq	79	79	22183	

	cutoff_timestamp	
count	27322	
unique	23824	
top	2018-09-24 05:19:20	
freq	8	

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

```
[14]:
```

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

4 training 2018-09-20 02:35:36.476840

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

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

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

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

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

	factor	segment_actual_time	segment_osrm_time	segment_osrm_distance	\
0	1.272727	14.0	11.0	11.9653	
1	1.200000	10.0	9.0	9.7590	
2	1.428571	16.0	7.0	10.8152	
3	1.550000	21.0	12.0	13.0224	
4	1.545455	6.0	5.0	3.9153	

	segment_factor
0	1.272727
1	1.111111

```

2      2.285714
3      1.750000
4      1.200000

```

[5 rows x 24 columns]

Datetime Formatting for datetime columns

```

[15]: df['trip_creation_time']=pd.to_datetime(df['trip_creation_time'])
      df['od_start_time']=pd.to_datetime(df['od_start_time'])
      df['od_end_time']=pd.to_datetime(df['od_end_time'])
      df['cutoff_timestamp']=pd.to_datetime(df['cutoff_timestamp'], format='mixed')

```

```

[16]: df.head()

```

```

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

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

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

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

      od_start_time  ...      cutoff_timestamp \
0 2018-09-20 03:21:32.418600  ... 2018-09-20 04:27:55.000000
1 2018-09-20 03:21:32.418600  ... 2018-09-20 04:17:55.000000
2 2018-09-20 03:21:32.418600  ... 2018-09-20 04:01:19.505586

```

```
3 2018-09-20 03:21:32.418600 ... 2018-09-20 03:39:57.000000
4 2018-09-20 03:21:32.418600 ... 2018-09-20 03:33:55.000000
```

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

	factor	segment_actual_time	segment_osrm_time	segment_osrm_distance	\
0	1.272727	14.0	11.0	11.9653	
1	1.200000	10.0	9.0	9.7590	
2	1.428571	16.0	7.0	10.8152	
3	1.550000	21.0	12.0	13.0224	
4	1.545455	6.0	5.0	3.9153	

	segment_factor
0	1.272727
1	1.111111
2	2.285714
3	1.750000
4	1.200000

[5 rows x 24 columns]

Missing Value Analysis

```
[17]: df.isna().sum()
```

```
[17]: data
trip_creation_time      0
route_schedule_uuid     0
route_type              0
trip_uuid               0
source_center           0
source_name             79
destination_center      1
destination_name        56
od_start_time           1
od_end_time             1
start_scan_to_end_scan  1
is_cutoff               1
cutoff_factor           1
cutoff_timestamp        1
actual_distance_to_destination  1
actual_time             1
```

```

osrm_time            1
osrm_distance        1
factor              1
segment_actual_time  1
segment_osrm_time    1
segment_osrm_distance 1
segment_factor       1
dtype: int64

```

Observations¶ Source name and Destination name have missing values as we have very records having missing value, we can drop those rows from analysis

```

[18]: # drop missing values
df.drop(df[df['source_name'].isna() | df['destination_name'].isna()].index ,_
        inplace=True)

```

```

[19]: df.isna().sum()

```

```

[19]: data            0
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

```

2. Trip Segment Analysis delivery details of one package is divided into several rows (think of it as connecting flights to reach a particular destination) we should find way to identify unique trip details for source_destination to understand delivery data. we can create unique identifier for different segment of a trip based on combination of trip_uuid, source_center

and destination_center, we will call it as segment key

Based on segment_key we will create new aggregated columns segment_actual_time, segment_osrm_distance, segment_osrm_time

```
[20]: df['segment_key']=df.apply(lambda x:"#".  
    ↪join([x['trip_uuid'],x['source_center'],x['destination_center']] ), axis=1)
```

```
[21]: df['segment_actual_time_cumsum']=df.  
    ↪groupby('segment_key')['segment_actual_time'].transform(lambda x:x.cumsum())  
  
df['segment_osrm_time_cumsum']=df.groupby('segment_key')['segment_osrm_time'].  
    ↪transform(lambda x:x.cumsum())  
  
df['segment_osrm_distance_cumsum']=df.  
    ↪groupby('segment_key')['segment_osrm_distance'].transform(lambda x:x.  
    ↪cumsum())
```

```
[22]: # df.groupby('segment_key')['segment_actual_time_sum'].nth(-1)
```

```
[23]: df['segment_actual_time_sum']=df.  
    ↪groupby('segment_key')['segment_actual_time_cumsum'].transform(lambda x:x.  
    ↪iloc[-1])  
df['segment_osrm_time_sum']=df.  
    ↪groupby('segment_key')['segment_osrm_time_cumsum'].transform(lambda x:x.  
    ↪iloc[-1])  
df['segment_osrm_distance_sum']=df.  
    ↪groupby('segment_key')['segment_osrm_distance_cumsum'].transform(lambda x:x.  
    ↪iloc[-1])
```

```
[24]: df=df.sort_values(by=['segment_key','od_end_time'])  
df
```

```
[24]:      data      trip_creation_time \  
21691  training 2018-09-12 00:06:39.565253  
21692  training 2018-09-12 00:06:39.565253  
20850  training 2018-09-12 00:11:40.783923  
1546   training 2018-09-12 00:25:19.499696  
1547   training 2018-09-12 00:25:19.499696  
...     ...  
18729   test 2018-10-03 23:57:44.429324  
11572   test 2018-10-03 23:59:42.701692  
11573   test 2018-10-03 23:59:42.701692  
11570   test 2018-10-03 23:59:42.701692  
11571   test 2018-10-03 23:59:42.701692
```

```
                                route_schedule_uuid route_type \  
21691  thanos::sroute:a10888ff-f794-41e1-9b7a-7f62ef6...   Carting
```


21692	thanos::sroute:a10888ff-f794-41e1-9b7a-7f62ef6...	Carting
20850	thanos::sroute:c2ee580f-f4b2-4fa5-98ab-0c5b327...	Carting
1546	thanos::sroute:0ac760f3-96cb-4046-bfd0-8bc4678...	FTL
1547	thanos::sroute:0ac760f3-96cb-4046-bfd0-8bc4678...	FTL
...
18729	thanos::sroute:5609c268-e436-4e0a-8180-3db4a74...	Carting
11572	thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...	FTL
11573	thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...	FTL
11570	thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...	FTL
11571	thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...	FTL

	trip_uuid	source_center	\
21691	trip-153671079956500691	IND110024AAA	
21692	trip-153671079956500691	IND110024AAA	
20850	trip-153671110078355292	IND121004AAB	
1546	trip-153671191949943656	IND487001AAB	
1547	trip-153671191949943656	IND487001AAB	
...	
18729	trip-153861106442901555	IND209304AAA	
11572	trip-153861118270144424	IND583119AAA	
11573	trip-153861118270144424	IND583119AAA	
11570	trip-153861118270144424	IND583201AAA	
11571	trip-153861118270144424	IND583201AAA	

	source_name	destination_center	\
21691	Delhi_Lajpat_IP (Delhi)	IND110014AAA	
21692	Delhi_Lajpat_IP (Delhi)	IND110014AAA	
20850	FBD_Balabgarh_DPC (Haryana)	IND121001AAA	
1546	Narsinghpur_KndliDPP_D (Madhya Pradesh)	IND487551AAA	
1547	Narsinghpur_KndliDPP_D (Madhya Pradesh)	IND487551AAA	
...	
18729	Kanpur_Central_H_6 (Uttar Pradesh)	IND208006AAA	
11572	Sandur_WrdN1DPP_D (Karnataka)	IND583101AAA	
11573	Sandur_WrdN1DPP_D (Karnataka)	IND583101AAA	
11570	Hospet (Karnataka)	IND583119AAA	
11571	Hospet (Karnataka)	IND583119AAA	

	destination_name	od_start_time	...	\
21691	Delhi_Bhogal (Delhi)	2018-09-12 00:06:39.565253	...	
21692	Delhi_Bhogal (Delhi)	2018-09-12 00:06:39.565253	...	
20850	Faridabad (Haryana)	2018-09-12 00:11:40.783923	...	
1546	Gadarwara_MPward_D (Madhya Pradesh)	2018-09-12 04:33:09.726898	...	
1547	Gadarwara_MPward_D (Madhya Pradesh)	2018-09-12 04:33:09.726898	...	
...	
18729	Kanpur_GovndNgr_DC (Uttar Pradesh)	2018-10-03 23:57:44.429324	...	
11572	Bellary_Dc (Karnataka)	2018-10-04 03:58:40.726547	...	
11573	Bellary_Dc (Karnataka)	2018-10-04 03:58:40.726547	...	

11570	Sandur_WrdN1DPP_D (Karnataka) 2018-10-04 02:51:44.712656	...
11571	Sandur_WrdN1DPP_D (Karnataka) 2018-10-04 02:51:44.712656	...

	segment_osrm_time	segment_osrm_distance	segment_factor \
21691	8.0	10.0040	1.500000
21692	6.0	6.0820	1.833333
20850	9.0	10.8159	1.888889
1546	25.0	27.9776	1.920000
1547	16.0	20.4353	1.875000
...
18729	0.0	0.1847	-1.000000
11572	17.0	23.9866	2.647059
11573	25.0	28.5437	7.520000
11570	21.0	23.2367	1.428571
11571	4.0	4.8117	2.750000

	segment_key \
21691	trip-153671079956500691#IND110024AAA#IND110014AAA
21692	trip-153671079956500691#IND110024AAA#IND110014AAA
20850	trip-153671110078355292#IND121004AAB#IND121001AAA
1546	trip-153671191949943656#IND487001AAB#IND487551AAA
1547	trip-153671191949943656#IND487001AAB#IND487551AAA
...	...
18729	trip-153861106442901555#IND209304AAA#IND208006AAA
11572	trip-153861118270144424#IND583119AAA#IND583101AAA
11573	trip-153861118270144424#IND583119AAA#IND583101AAA
11570	trip-153861118270144424#IND583201AAA#IND583119AAA
11571	trip-153861118270144424#IND583201AAA#IND583119AAA

	segment_actual_time_cumsum	segment_osrm_time_cumsum \
21691	12.0	8.0
21692	23.0	14.0
20850	17.0	9.0
1546	48.0	25.0
1547	78.0	41.0
...
18729	91.0	52.0
11572	45.0	17.0
11573	233.0	42.0
11570	30.0	21.0
11571	41.0	25.0

	segment_osrm_distance_cumsum	segment_actual_time_sum \
21691	10.0040	23.0
21692	16.0860	23.0
20850	10.8159	17.0
1546	27.9776	78.0

1547	48.4129	78.0
...
18729	69.4497	91.0
11572	23.9866	233.0
11573	52.5303	233.0
11570	23.2367	41.0
11571	28.0484	41.0

	segment_osrm_time_sum	segment_osrm_distance_sum
21691	14.0	16.0860
21692	14.0	16.0860
20850	9.0	10.8159
1546	41.0	48.4129
1547	41.0	48.4129
...
18729	52.0	69.4497
11572	42.0	52.5303
11573	42.0	52.5303
11570	25.0	28.0484
11571	25.0	28.0484

[27188 rows x 31 columns]

3. Feature Engineering

```
[25]: df['od_time_diff_hour']=(df['od_end_time']-df['od_start_time'])/pd.
      ↪Timedelta(hours=1)
```

Extract City, Place, Code and State Information¶

```
[26]: df.columns
```

```
[26]: Index(['data', 'trip_creation_time', 'route_schedule_uuid', 'route_type',
      'trip_uuid', 'source_center', 'source_name', 'destination_center',
      'destination_name', 'od_start_time', 'od_end_time',
      'start_scan_to_end_scan', 'is_cutoff', 'cutoff_factor',
      'cutoff_timestamp', 'actual_distance_to_destination', 'actual_time',
      'osrm_time', 'osrm_distance', 'factor', 'segment_actual_time',
      'segment_osrm_time', 'segment_osrm_distance', 'segment_factor',
      'segment_key', 'segment_actual_time_cumsum', 'segment_osrm_time_cumsum',
      'segment_osrm_distance_cumsum', 'segment_actual_time_sum',
      'segment_osrm_time_sum', 'segment_osrm_distance_sum',
      'od_time_diff_hour'],
      dtype='object')
```

```
[27]: df['trip_creation_year']=df['trip_creation_time'].dt.year
      df['trip_creation_month']=df['trip_creation_time'].dt.month
      df['trip_creation_day']=df['trip_creation_time'].dt.day
```

```
[28]: def get_state(name):
    pattern="\([A-Za-z]+\s?\w+\)"
    pattern="\([A-Za-z &]+\s?\w+\)"
    state=re.findall(pattern, name)[0]
    state=state.replace("(", "")
    state=state.replace(")", "")
    return state
```

```
[29]: def get_city(name):
    pattern="\([A-Za-z &]+\s?\w+\)"
    state=re.findall(pattern, name)[0]
    city_place_code=name.replace(state, '')
    city_place_code_parts=city_place_code.split("_")
    if len(city_place_code_parts)==1 :
        city=city_place_code_parts[0].strip()
    elif len(city_place_code_parts)==2:
        city=city_place_code.strip()
    elif len(city_place_code_parts)==3 or len(city_place_code_parts)==4:
        city=city_place_code_parts[0].strip()
    else:
        city=city_place_code
    return city
```

```
[34]: def get_place(name):
    pattern="\([A-Za-z &]+\s?\w+\)"
    try:
        state=re.findall(pattern, name)[0]
        city_place_code=name.replace(state, '')
        city_place_code_parts=city_place_code.split("_")
        if len(city_place_code_parts)==3 or len(city_place_code_parts)==4:
            place=city_place_code_parts[1].strip()
        else:
            place=None
        return place
    except Exception as exp:
        return None
```

```
[30]: def get_code(name):
    pattern="\([A-Za-z &]+\s?\w+\)"
    try:
        state=re.findall(pattern, name)[0]
        city_place_code=name.replace(state, '')
        city_place_code_parts=city_place_code.split("_")
        if len(city_place_code_parts)==3 :
            code=city_place_code_parts[2].strip()
        elif len(city_place_code_parts)==4:
            code="_".join(city_place_code_parts[2:]).strip()
```

```

    else:
        code=None
    return code
except Exception as exp:
    return None

```

```

[35]: from time import time
import re

```

```

[36]: df['destination_state']=df['destination_name'].map(get_state)
df['source_state']=df['source_name'].map(get_state)
df['destination_city']=df['destination_name'].map(get_city)
df['source_city']=df['source_name'].map(get_state)
df['destination_place']=df['destination_name'].map(get_place)
df['source_place']=df['source_name'].map(get_place)
df['destination_code']=df['destination_name'].map(get_code)
df['source_code']=df['source_name'].map(get_code)

```

4. In-depth analysis: Grouping and Aggregating at Trip-level¶

- a. Groups the segment data by the trip_uuid column to focus on aggregating data at the trip level.
- b. Apply suitable aggregation functions like first, last, and sum specified in the create_trip_dict dictionary to calculate summary statistics for each trip.

```

[37]: trip_actual_time_sum_df=df.groupby(['trip_uuid','segment_key']).nth(-1).
↳groupby(['trip_uuid']).agg(trip_actual_time_sum=('actual_time','sum')).
↳reset_index()
trip_actual_time_sum_df

```

```

[37]:
      trip_uuid  trip_actual_time_sum
0  trip-153671079956500691          23.0
1  trip-153671110078355292          17.0
2  trip-153671191949943656         253.0
3  trip-153671237597058150         241.0
4  trip-153671262893947351         195.0
...
2840 trip-153861034802474617         123.0
2841 trip-153861052318770017          47.0
2842 trip-153861089872028474          62.0
2843 trip-153861106442901555        282.0
2844 trip-153861118270144424        275.0

```

[2845 rows x 2 columns]

```

[38]:

```

```

trip_segment_actual_time_sum_df=df.groupby(['trip_uuid','segment_key']).nth(-1).
↳groupby(['trip_uuid']).
↳agg(trip_segment_actual_time_sum=('segment_actual_time_sum','sum')).
↳reset_index()
trip_segment_actual_time_sum_df

```

```

[38]:
      trip_uuid  trip_segment_actual_time_sum
0  trip-153671079956500691          23.0
1  trip-153671110078355292          17.0
2  trip-153671191949943656         252.0
3  trip-153671237597058150         240.0
4  trip-153671262893947351         195.0
...
2840 trip-153861034802474617         121.0
2841 trip-153861052318770017          47.0
2842 trip-153861089872028474          61.0
2843 trip-153861106442901555         281.0
2844 trip-153861118270144424         274.0

[2845 rows x 2 columns]

```

```

[39]: trip_segment_osrm_time_sum_df=df.groupby(['trip_uuid','segment_key']).nth(-1).
↳groupby(['trip_uuid']).
↳agg(trip_segment_osrm_time_sum=('segment_osrm_time_sum','sum')).reset_index()
trip_segment_osrm_time_sum_df

```

```

[39]:
      trip_uuid  trip_segment_osrm_time_sum
0  trip-153671079956500691          14.0
1  trip-153671110078355292           9.0
2  trip-153671191949943656         132.0
3  trip-153671237597058150          49.0
4  trip-153671262893947351          26.0
...
2840 trip-153861034802474617          50.0
2841 trip-153861052318770017          13.0
2842 trip-153861089872028474          33.0
2843 trip-153861106442901555          88.0
2844 trip-153861118270144424          67.0

[2845 rows x 2 columns]

```

```

[40]: trip_osrm_time_sum_df=df.groupby(['trip_uuid','segment_key']).nth(-1).
↳groupby(['trip_uuid']).agg(trip_osrm_time_sum=('osrm_time','sum')).
↳reset_index()
trip_osrm_time_sum_df

```

```
[40]:
```

	trip_uuid	trip_osrm_time_sum
0	trip-153671079956500691	8.0
1	trip-153671110078355292	9.0
2	trip-153671191949943656	105.0
3	trip-153671237597058150	33.0
4	trip-153671262893947351	26.0
...
2840	trip-153861034802474617	50.0
2841	trip-153861052318770017	13.0
2842	trip-153861089872028474	28.0
2843	trip-153861106442901555	48.0
2844	trip-153861118270144424	68.0

[2845 rows x 2 columns]

```
[41]: trip_osrm_distance_sum_df=df.groupby(['trip_uuid','segment_key']).nth(-1).
      ↳groupby(['trip_uuid']).agg(trip_osrm_distance_sum=('osrm_distance','sum')).
      ↳reset_index()
      trip_osrm_distance_sum_df
```

```
[41]:
```

	trip_uuid	trip_osrm_distance_sum
0	trip-153671079956500691	9.9566
1	trip-153671110078355292	10.8159
2	trip-153671191949943656	124.5063
3	trip-153671237597058150	46.9087
4	trip-153671262893947351	30.4646
...
2840	trip-153861034802474617	46.5093
2841	trip-153861052318770017	15.7803
2842	trip-153861089872028474	38.2867
2843	trip-153861106442901555	58.9037
2844	trip-153861118270144424	80.5787

[2845 rows x 2 columns]

```
[42]: trip_segment_osrm_distance_sum_df=df.groupby(['trip_uuid','segment_key']).
      ↳nth(-1).groupby(['trip_uuid']).
      ↳agg(trip_segment_osrm_distance_sum=('segment_osrm_distance_sum','sum')).
      ↳reset_index()
      trip_segment_osrm_distance_sum_df
```

```
[42]:
```

	trip_uuid	trip_segment_osrm_distance_sum
0	trip-153671079956500691	16.0860
1	trip-153671110078355292	10.8159
2	trip-153671191949943656	140.5623
3	trip-153671237597058150	56.7577
4	trip-153671262893947351	30.4646

```

...
2840 trip-153861034802474617 48.5795
2841 trip-153861052318770017 15.7803
2842 trip-153861089872028474 33.6400
2843 trip-153861106442901555 104.8866
2844 trip-153861118270144424 80.5787

```

[2845 rows x 2 columns]

```

[43]: trip_total_time_in_hrs_df=df.groupby(['trip_uuid','segment_key']).nth(-1).
      ↳groupby(['trip_uuid']).
      ↳agg(trip_total_time_in_hrs=('od_time_diff_hour','sum')).reset_index()
trip_total_time_in_hrs_df

```

```

[43]:
      trip_uuid  trip_total_time_in_hrs
0  trip-153671079956500691      0.822223
1  trip-153671110078355292      0.641675
2  trip-153671191949943656      4.844371
3  trip-153671237597058150      4.201283
4  trip-153671262893947351      4.500402
...
2840 trip-153861034802474617      2.479912
2841 trip-153861052318770017      1.395993
2842 trip-153861089872028474      3.040278
2843 trip-153861106442901555      7.035331
2844 trip-153861118270144424      5.906793

```

[2845 rows x 2 columns]

Combine all trip aggregated dataframes to single trip aggregations dataframe

```

[44]: trip_aggregated_df=pd.concat(
      objs=(
          iDF.set_index('trip_uuid') for iDF in
      ↳(trip_actual_time_sum_df,trip_osrm_time_sum_df, trip_osrm_distance_sum_df,
          ↳
      ↳trip_segment_actual_time_sum_df,trip_segment_osrm_time_sum_df,trip_segment_osrm_distance_sum_df,
          trip_total_time_in_hrs_df)
      ),
      axis=1,
      join='inner'
    ).reset_index()
trip_aggregated_df

```

```

[44]:
      trip_uuid  trip_actual_time_sum  trip_osrm_time_sum \
0  trip-153671079956500691          23.0           8.0
1  trip-153671110078355292          17.0           9.0

```


2	trip-153671191949943656	253.0	105.0
3	trip-153671237597058150	241.0	33.0
4	trip-153671262893947351	195.0	26.0
...
2840	trip-153861034802474617	123.0	50.0
2841	trip-153861052318770017	47.0	13.0
2842	trip-153861089872028474	62.0	28.0
2843	trip-153861106442901555	282.0	48.0
2844	trip-153861118270144424	275.0	68.0

	trip_osrm_distance_sum	trip_segment_actual_time_sum \
0	9.9566	23.0
1	10.8159	17.0
2	124.5063	252.0
3	46.9087	240.0
4	30.4646	195.0
...
2840	46.5093	121.0
2841	15.7803	47.0
2842	38.2867	61.0
2843	58.9037	281.0
2844	80.5787	274.0

	trip_segment_osrm_time_sum	trip_segment_osrm_distance_sum \
0	14.0	16.0860
1	9.0	10.8159
2	132.0	140.5623
3	49.0	56.7577
4	26.0	30.4646
...
2840	50.0	48.5795
2841	13.0	15.7803
2842	33.0	33.6400
2843	88.0	104.8866
2844	67.0	80.5787

	trip_total_time_in_hrs
0	0.822223
1	0.641675
2	4.844371
3	4.201283
4	4.500402
...	...
2840	2.479912
2841	1.395993
2842	3.040278
2843	7.035331

2844

5.906793

[2845 rows x 8 columns]

```
[45]: df=pd.merge(df,trip_aggregated_df, on='trip_uuid' )
df
```

```
[45]:
```

	data	trip_creation_time	\
0	training	2018-09-12 00:06:39.565253	
1	training	2018-09-12 00:06:39.565253	
2	training	2018-09-12 00:11:40.783923	
3	training	2018-09-12 00:25:19.499696	
4	training	2018-09-12 00:25:19.499696	
...	
27183	test	2018-10-03 23:57:44.429324	
27184	test	2018-10-03 23:59:42.701692	
27185	test	2018-10-03 23:59:42.701692	
27186	test	2018-10-03 23:59:42.701692	
27187	test	2018-10-03 23:59:42.701692	

	route_schedule_uuid	route_type	\
0	thanos::sroute:a10888ff-f794-41e1-9b7a-7f62ef6...	Carting	
1	thanos::sroute:a10888ff-f794-41e1-9b7a-7f62ef6...	Carting	
2	thanos::sroute:c2ee580f-f4b2-4fa5-98ab-0c5b327...	Carting	
3	thanos::sroute:0ac760f3-96cb-4046-bfd0-8bc4678...	FTL	
4	thanos::sroute:0ac760f3-96cb-4046-bfd0-8bc4678...	FTL	
...	
27183	thanos::sroute:5609c268-e436-4e0a-8180-3db4a74...	Carting	
27184	thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...	FTL	
27185	thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...	FTL	
27186	thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...	FTL	
27187	thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...	FTL	

	trip_uuid	source_center	\
0	trip-153671079956500691	IND110024AAA	
1	trip-153671079956500691	IND110024AAA	
2	trip-153671110078355292	IND121004AAB	
3	trip-153671191949943656	IND487001AAB	
4	trip-153671191949943656	IND487001AAB	
...	
27183	trip-153861106442901555	IND209304AAA	
27184	trip-153861118270144424	IND583119AAA	
27185	trip-153861118270144424	IND583119AAA	
27186	trip-153861118270144424	IND583201AAA	
27187	trip-153861118270144424	IND583201AAA	

	source_name	destination_center	\
--	-------------	--------------------	---

0	Delhi_Lajpat_IP (Delhi)	IND110014AAA
1	Delhi_Lajpat_IP (Delhi)	IND110014AAA
2	FBD_Balabhgarh_DPC (Haryana)	IND121001AAA
3	Narsinghpur_KndliDPP_D (Madhya Pradesh)	IND487551AAA
4	Narsinghpur_KndliDPP_D (Madhya Pradesh)	IND487551AAA
...
27183	Kanpur_Central_H_6 (Uttar Pradesh)	IND208006AAA
27184	Sandur_WrdN1DPP_D (Karnataka)	IND583101AAA
27185	Sandur_WrdN1DPP_D (Karnataka)	IND583101AAA
27186	Hospet (Karnataka)	IND583119AAA
27187	Hospet (Karnataka)	IND583119AAA

	destination_name	od_start_time	...	\
0	Delhi_Bhogal (Delhi)	2018-09-12 00:06:39.565253	...	
1	Delhi_Bhogal (Delhi)	2018-09-12 00:06:39.565253	...	
2	Faridabad (Haryana)	2018-09-12 00:11:40.783923	...	
3	Gadarwara_MPward_D (Madhya Pradesh)	2018-09-12 04:33:09.726898	...	
4	Gadarwara_MPward_D (Madhya Pradesh)	2018-09-12 04:33:09.726898	...	
...
27183	Kanpur_GovndNgr_DC (Uttar Pradesh)	2018-10-03 23:57:44.429324	...	
27184	Bellary_Dc (Karnataka)	2018-10-04 03:58:40.726547	...	
27185	Bellary_Dc (Karnataka)	2018-10-04 03:58:40.726547	...	
27186	Sandur_WrdN1DPP_D (Karnataka)	2018-10-04 02:51:44.712656	...	
27187	Sandur_WrdN1DPP_D (Karnataka)	2018-10-04 02:51:44.712656	...	

	source_place	destination_code	source_code	trip_actual_time_sum	\
0	Lajpat	None	IP	23.0	
1	Lajpat	None	IP	23.0	
2	Balabhgarh	None	DPC	17.0	
3	KndliDPP	D	D	253.0	
4	KndliDPP	D	D	253.0	
...
27183	Central	DC	H_6	282.0	
27184	WrdN1DPP	None	D	275.0	
27185	WrdN1DPP	None	D	275.0	
27186	None	D	None	275.0	
27187	None	D	None	275.0	

	trip_osrm_time_sum	trip_osrm_distance_sum	\
0	8.0	9.9566	
1	8.0	9.9566	
2	9.0	10.8159	
3	105.0	124.5063	
4	105.0	124.5063	
...
27183	48.0	58.9037	
27184	68.0	80.5787	

27185	68.0	80.5787
27186	68.0	80.5787
27187	68.0	80.5787

	trip_segment_actual_time_sum	trip_segment_osrm_time_sum \
0	23.0	14.0
1	23.0	14.0
2	17.0	9.0
3	252.0	132.0
4	252.0	132.0
...
27183	281.0	88.0
27184	274.0	67.0
27185	274.0	67.0
27186	274.0	67.0
27187	274.0	67.0

	trip_segment_osrm_distance_sum	trip_total_time_in_hrs
0	16.0860	0.822223
1	16.0860	0.822223
2	10.8159	0.641675
3	140.5623	4.844371
4	140.5623	4.844371
...
27183	104.8866	7.035331
27184	80.5787	5.906793
27185	80.5787	5.906793
27186	80.5787	5.906793
27187	80.5787	5.906793

[27188 rows x 50 columns]

Outlier Detection & Treatment¶ a. Find any existing outliers in numerical features.

b. Visualize the outlier values using Boxplot.

c. Handle the outliers using the IQR method.

Perform one-hot encoding on categorical features. Normalize/ Standardize the numerical features using MinMaxScaler or StandardScaler.

```
[46]: trip_aggregated_df
```

```
[46]:
```

	trip_uuid	trip_actual_time_sum	trip_osrm_time_sum \
0	trip-153671079956500691	23.0	8.0
1	trip-153671110078355292	17.0	9.0
2	trip-153671191949943656	253.0	105.0
3	trip-153671237597058150	241.0	33.0
4	trip-153671262893947351	195.0	26.0

...
2840	trip-153861034802474617	123.0	50.0
2841	trip-153861052318770017	47.0	13.0
2842	trip-153861089872028474	62.0	28.0
2843	trip-153861106442901555	282.0	48.0
2844	trip-153861118270144424	275.0	68.0

	trip_osrm_distance_sum	trip_segment_actual_time_sum \
0	9.9566	23.0
1	10.8159	17.0
2	124.5063	252.0
3	46.9087	240.0
4	30.4646	195.0
...
2840	46.5093	121.0
2841	15.7803	47.0
2842	38.2867	61.0
2843	58.9037	281.0
2844	80.5787	274.0

	trip_segment_osrm_time_sum	trip_segment_osrm_distance_sum \
0	14.0	16.0860
1	9.0	10.8159
2	132.0	140.5623
3	49.0	56.7577
4	26.0	30.4646
...
2840	50.0	48.5795
2841	13.0	15.7803
2842	33.0	33.6400
2843	88.0	104.8866
2844	67.0	80.5787

	trip_total_time_in_hrs
0	0.822223
1	0.641675
2	4.844371
3	4.201283
4	4.500402
...	...
2840	2.479912
2841	1.395993
2842	3.040278
2843	7.035331
2844	5.906793

[2845 rows x 8 columns]

```

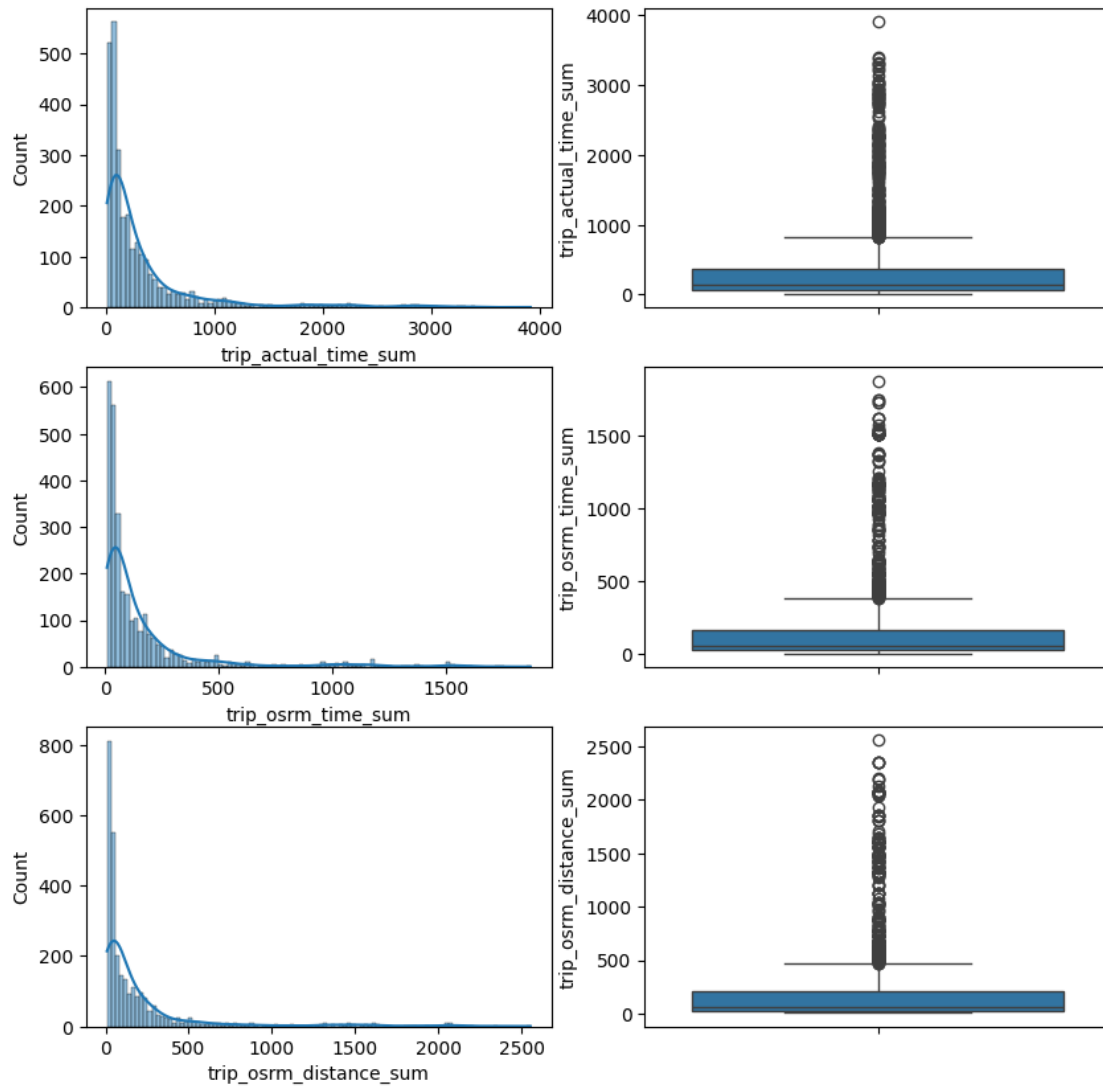
[47]: import warnings
warnings.filterwarnings("ignore")
fig, axs = plt.subplots(3, 2, figsize=(10,10))
sns.histplot(ax=axs[0,0],data=□
    ↳trip_aggregated_df['trip_actual_time_sum'],kde=True)
sns.boxplot(ax=axs[0,1],data=trip_aggregated_df['trip_actual_time_sum'])

sns.histplot(ax=axs[1,0],data=□
    ↳trip_aggregated_df['trip_osrm_time_sum'],kde=True)
sns.boxplot(ax=axs[1,1],data= trip_aggregated_df['trip_osrm_time_sum'])

sns.histplot(ax=axs[2,0],data=□
    ↳trip_aggregated_df['trip_osrm_distance_sum'],kde=True)
sns.boxplot(ax=axs[2,1],data= trip_aggregated_df['trip_osrm_distance_sum'])

plt.show()

```



```
[48]: import warnings
warnings.filterwarnings("ignore")
fig, axs = plt.subplots(3, 2, figsize=(10,10))

sns.histplot(ax=axs[0,0],data=
    ↳trip_aggregated_df['trip_segment_actual_time_sum'],kde=True)
sns.boxplot(ax=axs[0,1],data=
    ↳trip_aggregated_df['trip_segment_actual_time_sum'])

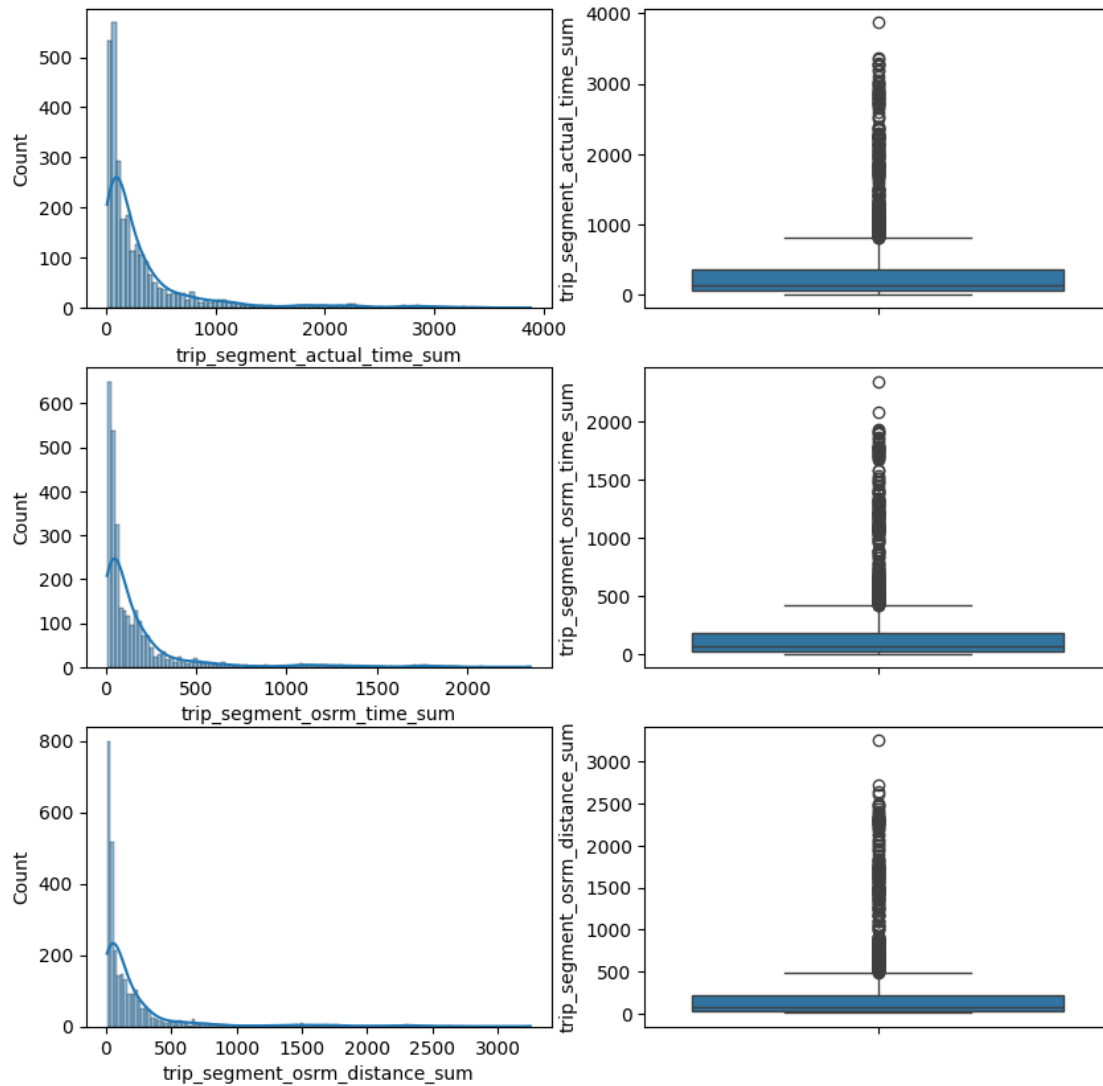
sns.histplot(ax=axs[1,0],data=
    ↳trip_aggregated_df['trip_segment_osrm_time_sum'],kde=True)
sns.boxplot(ax=axs[1,1],data= trip_aggregated_df['trip_segment_osrm_time_sum'])
```

```

sns.histplot(ax=axes[2,0],data=
↳trip_aggregated_df['trip_segment_osrm_distance_sum'],kde=True)
sns.boxplot(ax=axes[2,1],data=
↳trip_aggregated_df['trip_segment_osrm_distance_sum'])

plt.show()

```



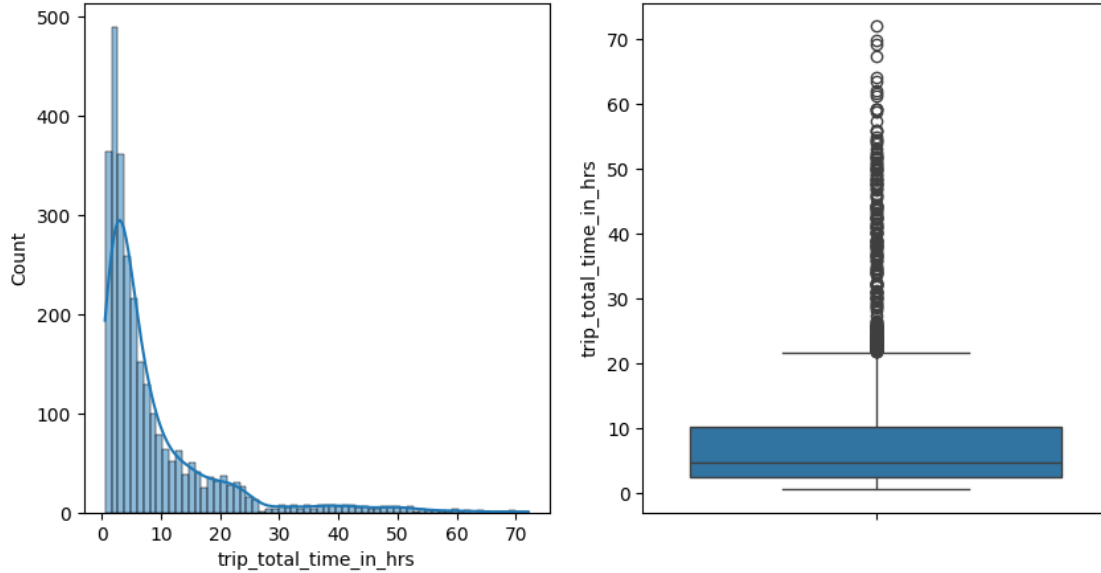
```

[49]: import warnings
warnings.filterwarnings("ignore")
fig, axes = plt.subplots(ncols= 2,  figsize=(10,5))

```

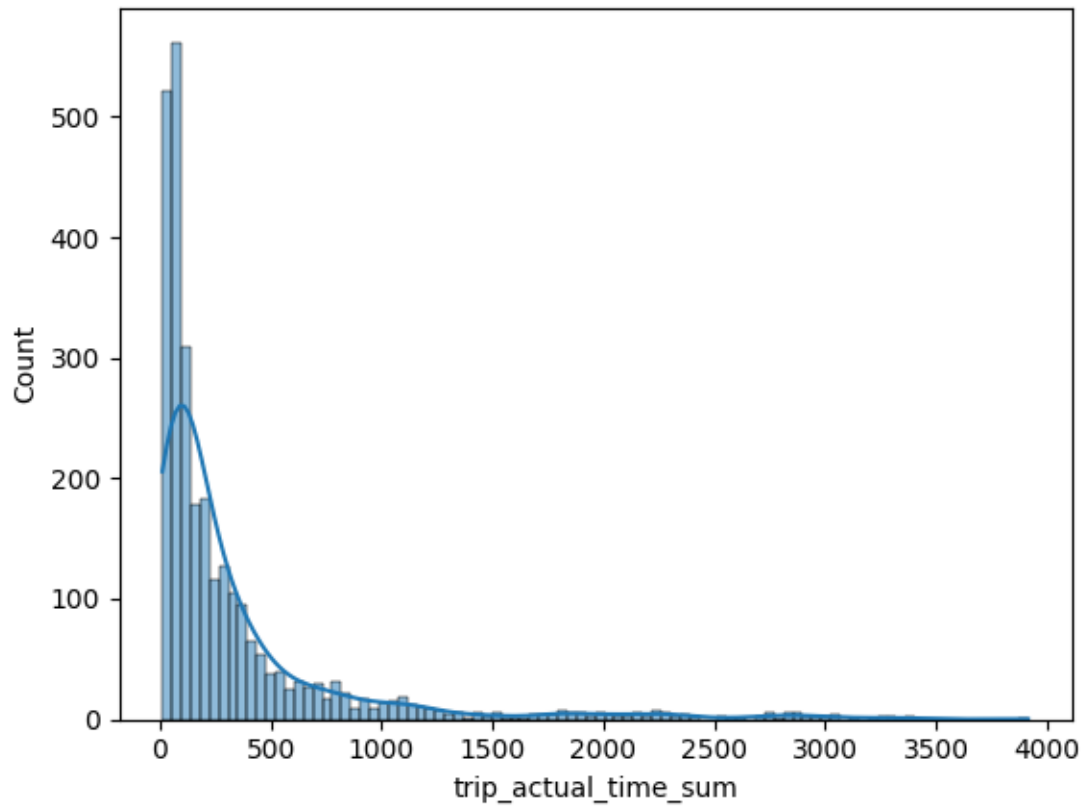


```
sns.histplot(ax=axes[0],data=trip_aggregated_df['trip_total_time_in_hrs'],kde=True)
sns.boxplot(ax=axes[1],data=trip_aggregated_df['trip_total_time_in_hrs'])
plt.show()
```



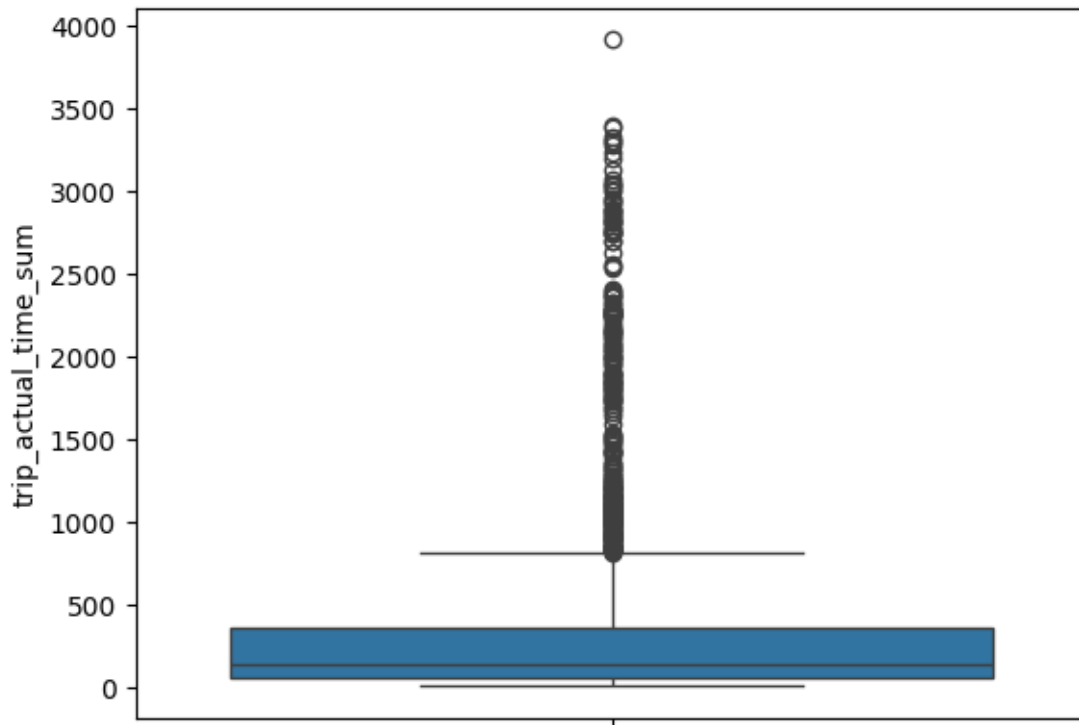
```
[50]: sns.histplot(data=trip_aggregated_df['trip_actual_time_sum'],kde=True)
```

```
[50]: <Axes: xlabel='trip_actual_time_sum', ylabel='Count'>
```



```
[51]: sns.boxplot(data=trip_aggregated_df['trip_actual_time_sum'])
```

```
[51]: <Axes: ylabel='trip_actual_time_sum'>
```



Observations we have lots of outliers let's remove those outliers Based on IQR range, we will maxout values based on IQR range

```
[52]: def clip_value_helper(row,cl, Q1,Q3, minval, maxval):
#     Q1=row[cl].quantile(0.25)
#     Q3=row[cl].quantile(0.75)
#     minval=min(row[cl])
#     maxval=max(row[cl])
IQR=Q3-Q1
if row[cl]<Q1-1.5*IQR:
    return min(minval, Q1-1.5*IQR)
elif row[cl] > Q3+1.5*IQR:
    return min(maxval, Q3+1.5*IQR)
else:
    return row[cl]
```

```
[53]: trip_aggregated_df.columns
```

```
[53]: Index(['trip_uuid', 'trip_actual_time_sum', 'trip_osrm_time_sum',
'trip_osrm_distance_sum', 'trip_segment_actual_time_sum',
'trip_segment_osrm_time_sum', 'trip_segment_osrm_distance_sum',
'trip_total_time_in_hrs'],
dtype='object')
```

```
[54]: for cl in ['trip_actual_time_sum', 'trip_osrm_time_sum',
↳ 'trip_osrm_distance_sum',
↳ 'trip_segment_actual_time_sum', 'trip_segment_osrm_time_sum',
↳ 'trip_segment_osrm_distance_sum', 'trip_total_time_in_hrs']:
    Q1=trip_aggregated_df[cl].quantile(0.25)
    Q3=trip_aggregated_df[cl].quantile(0.75)
    minval=min(trip_aggregated_df[cl])
    maxval=max(trip_aggregated_df[cl])
    trip_aggregated_df[cl]=trip_aggregated_df.apply(lambda row:
↳ clip_value_helper(row,cl,
                                                                   
↳ Q1,Q3,minval, maxval) , axis=1)
    trip_aggregated_df
```

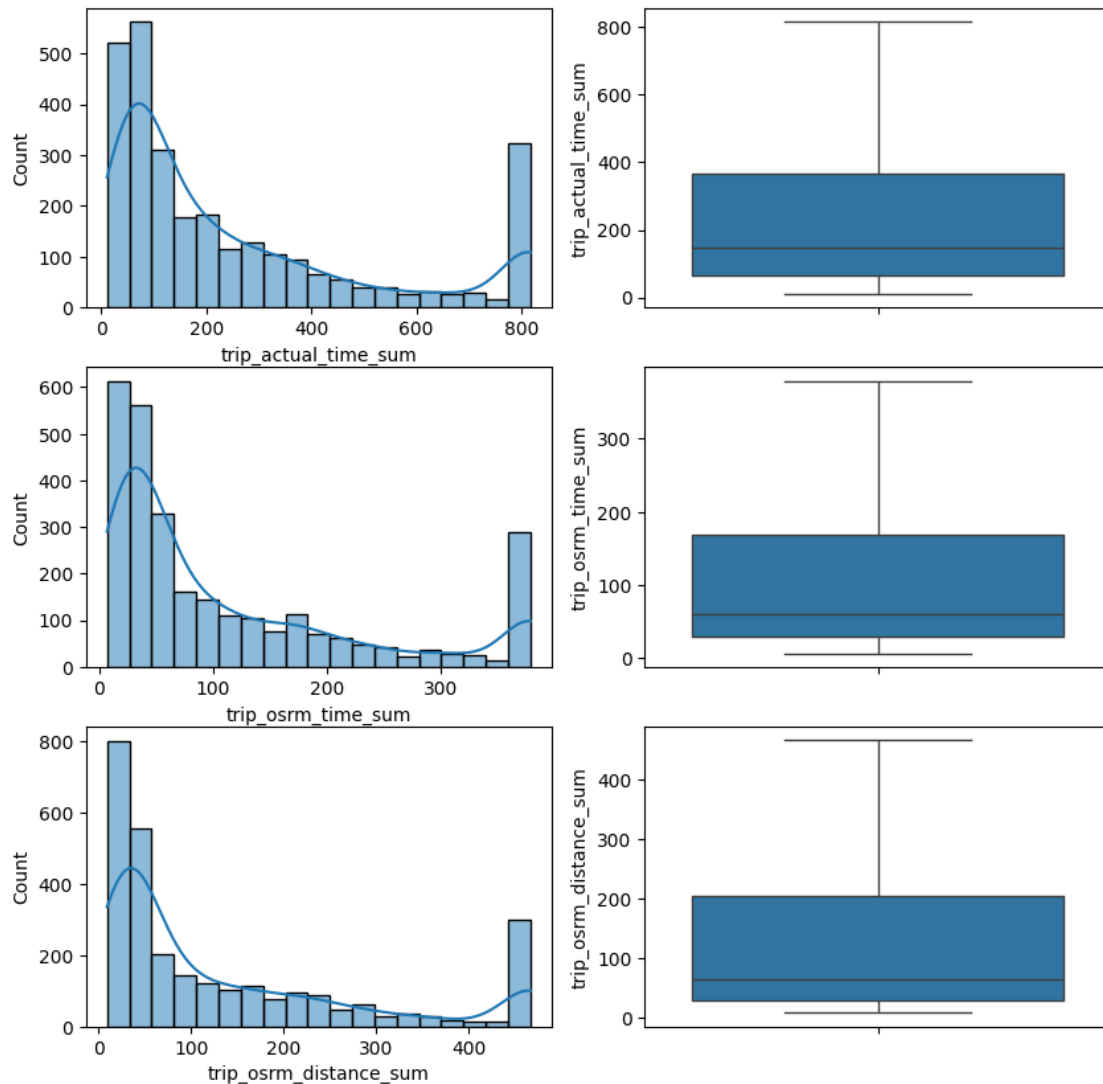
After Clipping Outliers Based On IQR Distribution¶

```
[55]: import warnings
warnings.filterwarnings("ignore")
fig, axs = plt.subplots(3, 2, figsize=(10,10))
sns.histplot(ax=axs[0,0],data=
↳ trip_aggregated_df['trip_actual_time_sum'],kde=True)
sns.boxplot(ax=axs[0,1],data=trip_aggregated_df['trip_actual_time_sum'])

sns.histplot(ax=axs[1,0],data=
↳ trip_aggregated_df['trip_osrm_time_sum'],kde=True)
sns.boxplot(ax=axs[1,1],data= trip_aggregated_df['trip_osrm_time_sum'])

sns.histplot(ax=axs[2,0],data=
↳ trip_aggregated_df['trip_osrm_distance_sum'],kde=True)
sns.boxplot(ax=axs[2,1],data= trip_aggregated_df['trip_osrm_distance_sum'])

plt.show()
```



```
[56]: import warnings
warnings.filterwarnings("ignore")
fig, axs = plt.subplots(3, 2, figsize=(10,10))

sns.histplot(ax=axs[0,0],data=
    ↳trip_aggregated_df['trip_segment_actual_time_sum'],kde=True)
sns.boxplot(ax=axs[0,1],data=
    ↳trip_aggregated_df['trip_segment_actual_time_sum'])

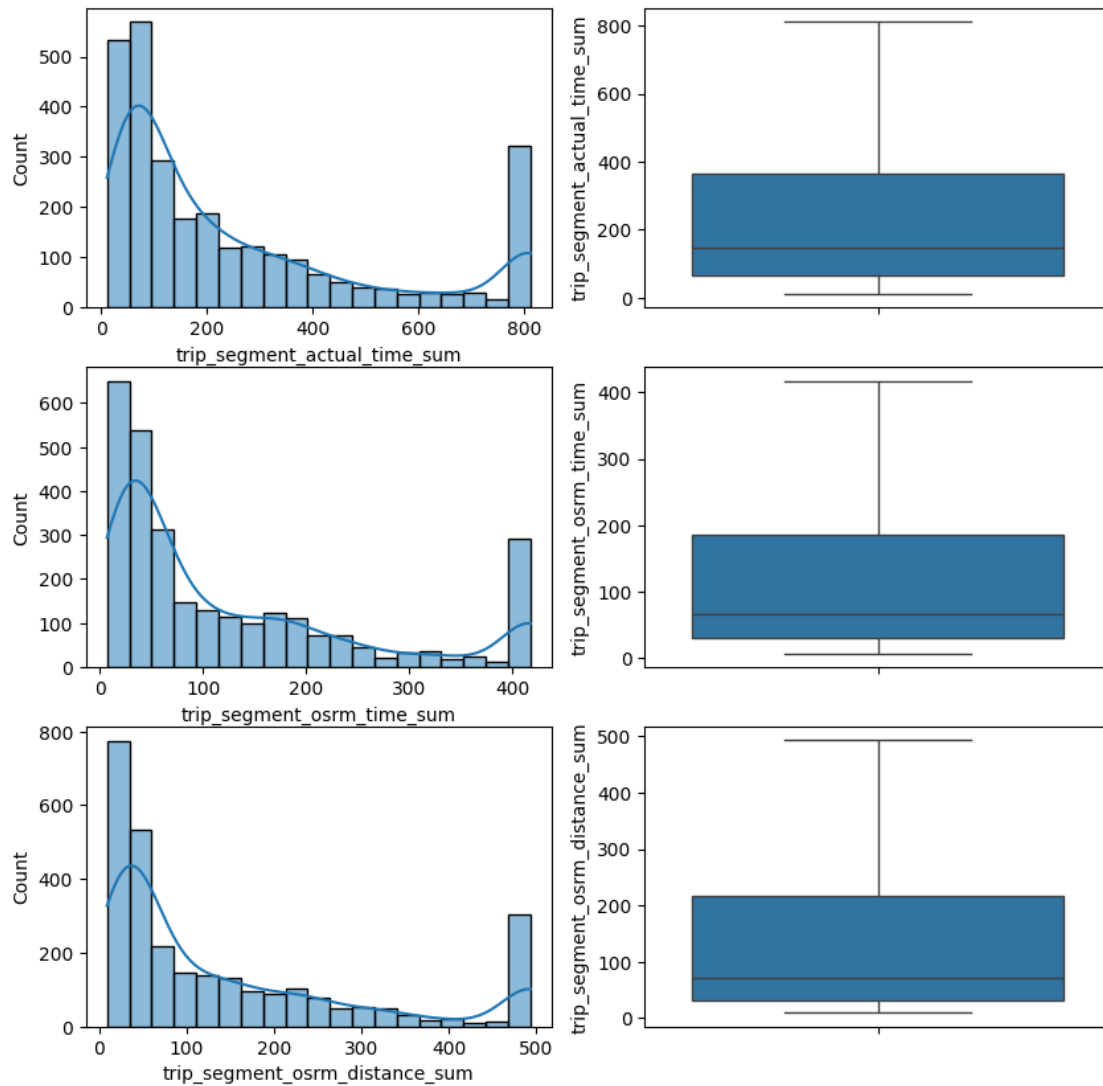
sns.histplot(ax=axs[1,0],data=
    ↳trip_aggregated_df['trip_segment_osrm_time_sum'],kde=True)
sns.boxplot(ax=axs[1,1],data= trip_aggregated_df['trip_segment_osrm_time_sum'])
```

```

sns.histplot(ax=axes[2,0],data=
    ↳trip_aggregated_df['trip_segment_osrm_distance_sum'],kde=True)
sns.boxplot(ax=axes[2,1],data=
    ↳trip_aggregated_df['trip_segment_osrm_distance_sum'])

plt.show()

```

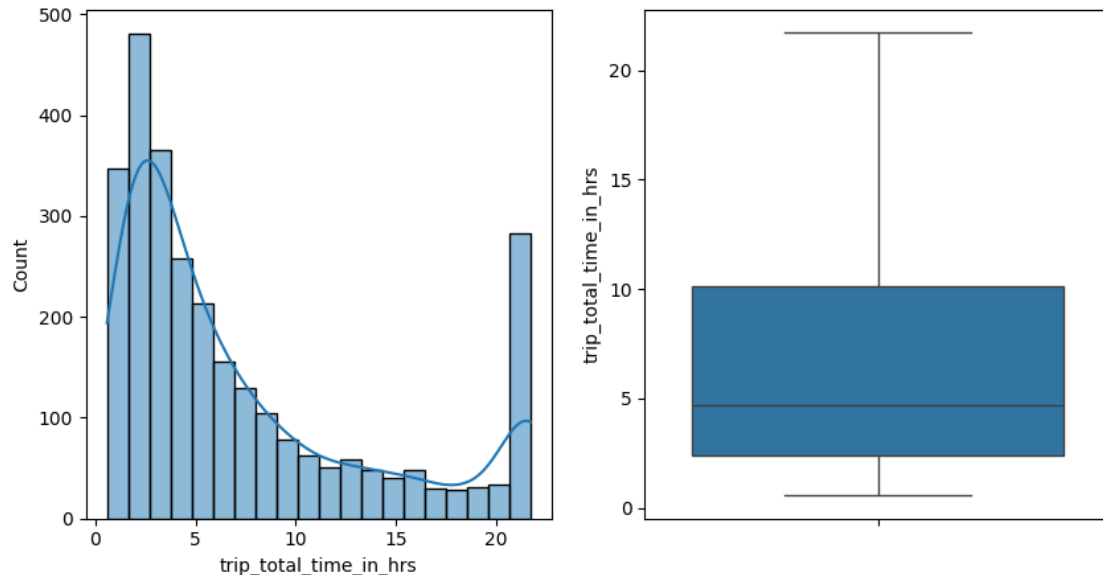


```

[57]: import warnings
warnings.filterwarnings("ignore")
fig, axes = plt.subplots(ncols= 2,  figsize=(10,5))

```

```
sns.histplot(ax=axis[0],data=
    ↳trip_aggregated_df['trip_total_time_in_hrs'],kde=True)
sns.boxplot(ax=axis[1],data= trip_aggregated_df['trip_total_time_in_hrs'])
plt.show()
```



5. Hypothesis Testing

. actual_time aggregated value and OSRM time aggregated value.

we will use ttest paired sample test to know if there is significant difference in actual trip aggregated time and OSRM trip aggregated time for each trip

HO : mean Actual time to deliver package from source to destination is lesser than OSRM time for entire trip

HA: mean Actual time to deliver package from source to destination is greater than OSRM time

```
[58]: from scipy.stats import ttest_rel, ttest_ind
stat, pval=ttest_rel(trip_aggregated_df['trip_actual_time_sum'],
                    trip_aggregated_df['trip_osrm_time_sum'],
                    alternative='greater')
print(f"stat {stat} pval {pval}")
```

stat 49.54905609405925 pval 0.0

```
[59]: if pval <0.05:
    print("We will reject H0 :")
    print("mean Actual time to deliver package from source to destination is_
    ↳greater than OSRM time")
else:
```

```

print("We fail to reject H0 :")
print("mean Actual time to deliver package from source to destination is_
↳lesser or equal than OSRM time")

```

We will reject H0 :

mean Actual time to deliver package from source to destination is greater than OSRM time

```

[60]: np.mean(trip_aggregated_df['trip_actual_time_sum']), np.
↳mean(trip_aggregated_df['trip_osrm_time_sum'])

```

```

[60]: (258.7082601054482, 115.17223198594024)

```

- b. actual_time aggregated value and segment actual time aggregated value. we will use ttest paired sample test to know if there is significant difference in actual trip aggregated time and segment actual time aggregated value for each trip

HO : mean Actual aggregated trip time to deliver package from source to destination is lesser than segment actual time aggregated value for entire trip

HA: mean Actual aggregated trip time to deliver package from source to destination is greater than segment actual time aggregated value for entire trip

```

[61]: stat, pval=ttest_rel(trip_aggregated_df['trip_actual_time_sum'],
                           trip_aggregated_df['trip_segment_actual_time_sum'],
                           alternative='greater')
print(f"stat {stat} pval {pval}")

```

```

stat 55.445752235969664 pval 0.0

```

```

[62]: if pval <0.05:
        print("We will reject H0 :")
        print("mean Actual trip aggregated time to deliver package from source to_
↳destination is greater than segment actual aggregated time")
    else:
        print("We fail to reject H0 :")
        print("mean Actual trip aggregated time to deliver package from source to_
↳destination is lesser or equal than segment actual aggregated time")

```

We will reject H0 :

mean Actual trip aggregated time to deliver package from source to destination is greater than segment actual aggregated time

- c. OSRM distance aggregated value and segment OSRM distance aggregated value. ¶ we will use ttest paired sample test to know if there is significant difference in OSRM distance aggregated value and segment actual aggregated distance for each trip

HO : mean Actual aggregated OSRM distance for trip to deliver package from source to destination is lesser than segment actual OSRM distance aggregated value for entire trip

HA: mean Actual aggregated OSRM distance for trip to deliver package from source to destination is greater than segment actual OSRM distance aggregated value for entire trip

```
[63]: stat, pval=ttest_rel(trip_aggregated_df['trip_osrm_distance_sum'],
                           trip_aggregated_df['trip_segment_osrm_distance_sum'],
                           alternative='greater')
print(f"stat {stat} pval {pval}")
```

```
stat -21.773765031064443 pval 1.0
```

```
[64]: if pval <0.05:
    print("We will reject H0 :")
    print("mean Actual trip OSRM distance for trip to deliver package from_
    ↳source to destination is greater than segment actual aggregated OSRM_
    ↳distance for trip")
else:
    print("We fail to reject H0 :")
    print("mean Actual trip OSRM distance for trip to deliver package from_
    ↳source to destination is lesser or equal than segment actual aggregated_
    ↳OSRM distance for trip")
```

We fail to reject H0 :
 mean Actual trip OSRM distance for trip to deliver package from source to destination is lesser or equal than segment actual aggregated OSRM distance for trip

```
[65]: trip_aggregated_df
```

```
[65]:
```

	trip_uuid	trip_actual_time_sum	trip_osrm_time_sum \
0	trip-153671079956500691	23.0	8.0
1	trip-153671110078355292	17.0	9.0
2	trip-153671191949943656	253.0	105.0
3	trip-153671237597058150	241.0	33.0
4	trip-153671262893947351	195.0	26.0
...
2840	trip-153861034802474617	123.0	50.0
2841	trip-153861052318770017	47.0	13.0
2842	trip-153861089872028474	62.0	28.0
2843	trip-153861106442901555	282.0	48.0
2844	trip-153861118270144424	275.0	68.0

	trip_osrm_distance_sum	trip_segment_actual_time_sum \
0	9.9566	23.0
1	10.8159	17.0
2	124.5063	252.0
3	46.9087	240.0
4	30.4646	195.0
...

2840	46.5093	121.0
2841	15.7803	47.0
2842	38.2867	61.0
2843	58.9037	281.0
2844	80.5787	274.0

	trip_segment_osrm_time_sum	trip_segment_osrm_distance_sum \
0	14.0	16.0860
1	9.0	10.8159
2	132.0	140.5623
3	49.0	56.7577
4	26.0	30.4646
...
2840	50.0	48.5795
2841	13.0	15.7803
2842	33.0	33.6400
2843	88.0	104.8866
2844	67.0	80.5787

	trip_total_time_in_hrs
0	0.822223
1	0.641675
2	4.844371
3	4.201283
4	4.500402
...	...
2840	2.479912
2841	1.395993
2842	3.040278
2843	7.035331
2844	5.906793

[2845 rows x 8 columns]

```
[66]: np.max(trip_aggregated_df['trip_actual_time_sum']), np.
      ↪max(trip_aggregated_df['trip_segment_actual_time_sum'])
```

[66]: (817.5, 813.0)

OSRM time aggregated value and segment OSRM time aggregated value. ¶ we will use ttest paired sample test to know if there is significant difference in OSRM time aggregated value and segment OSRM aggregated time for each trip

HO : mean Actual aggregated OSRM time aggregated for trip to deliver package from source to destination is lesser than segment OSRM aggregated time value for entire trip

HA: mean Actual aggregated OSRM time aggregated for trip to deliver package from source to destination is greater than segment OSRM aggregated time value for entire trip

```
[67]: stat, pval=ttest_rel(trip_aggregated_df['trip_osrm_time_sum'],
                           trip_aggregated_df['trip_segment_osrm_time_sum'],
                           alternative='greater')
print(f"stat {stat} pval {pval}")
```

```
stat -27.296301433442657 pval 1.0
```

```
[68]: if pval <0.05:
        print("We will reject H0 :")
        print("mean Actual trip aggregated OSRM time for trip to deliver package_
        ↳from source to destination is greater than segment actual aggregated OSRM_
        ↳time for trip")
    else:
        print("We fail to reject H0 :")
        print("mean Actual trip aggregated OSRM time for trip to deliver package_
        ↳from source to destination is lesser or equal than segment actual_
        ↳aggregated OSRM time for trip")
```

We fail to reject H0 :

mean Actual trip aggregated OSRM time for trip to deliver package from source to destination is lesser or equal than segment actual aggregated OSRM time for trip

Business Insights & Recommendations

From Where the Most Orders are coming from

```
[69]: # Check from where most orders are coming from (State, Corridor, etc.)

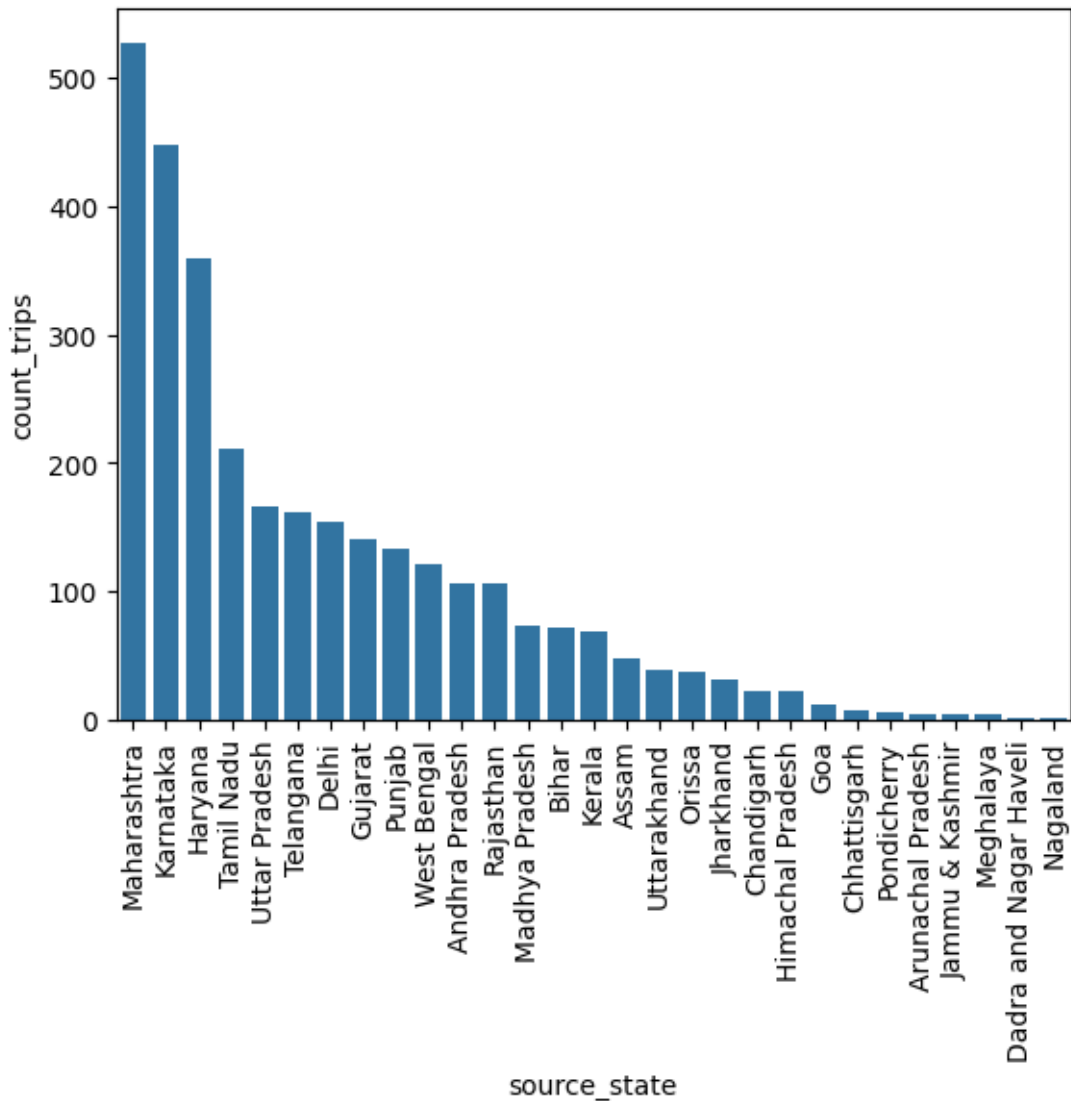
orders_from_df=df.groupby('source_state').
        ↳agg(count_trips=('trip_uuid', 'nunique')).reset_index()
orders_from_df=orders_from_df.sort_values(by='count_trips', ascending=False)
orders_from_df
```

```
[69]:
```

	source_state	count_trips
17	Maharashtra	527
14	Karnataka	448
10	Haryana	359
24	Tamil Nadu	211
26	Uttar Pradesh	166
25	Telangana	161
7	Delhi	154
9	Gujarat	141
22	Punjab	133
28	West Bengal	121
0	Andhra Pradesh	107
23	Rajasthan	107

16	Madhya Pradesh	73
3	Bihar	72
15	Kerala	69
2	Assam	48
27	Uttarakhand	39
20	Orissa	38
13	Jharkhand	31
4	Chandigarh	23
11	Himachal Pradesh	23
8	Goa	12
5	Chhattisgarh	7
21	Pondicherry	6
1	Arunachal Pradesh	5
12	Jammu & Kashmir	5
18	Meghalaya	4
6	Dadra and Nagar Haveli	2
19	Nagaland	2

```
[70]: sns.barplot(data=orders_from_df,x='source_state',y='count_trips')
plt.xticks(rotation=90)
plt.show()
```



To Which State Most orders are going

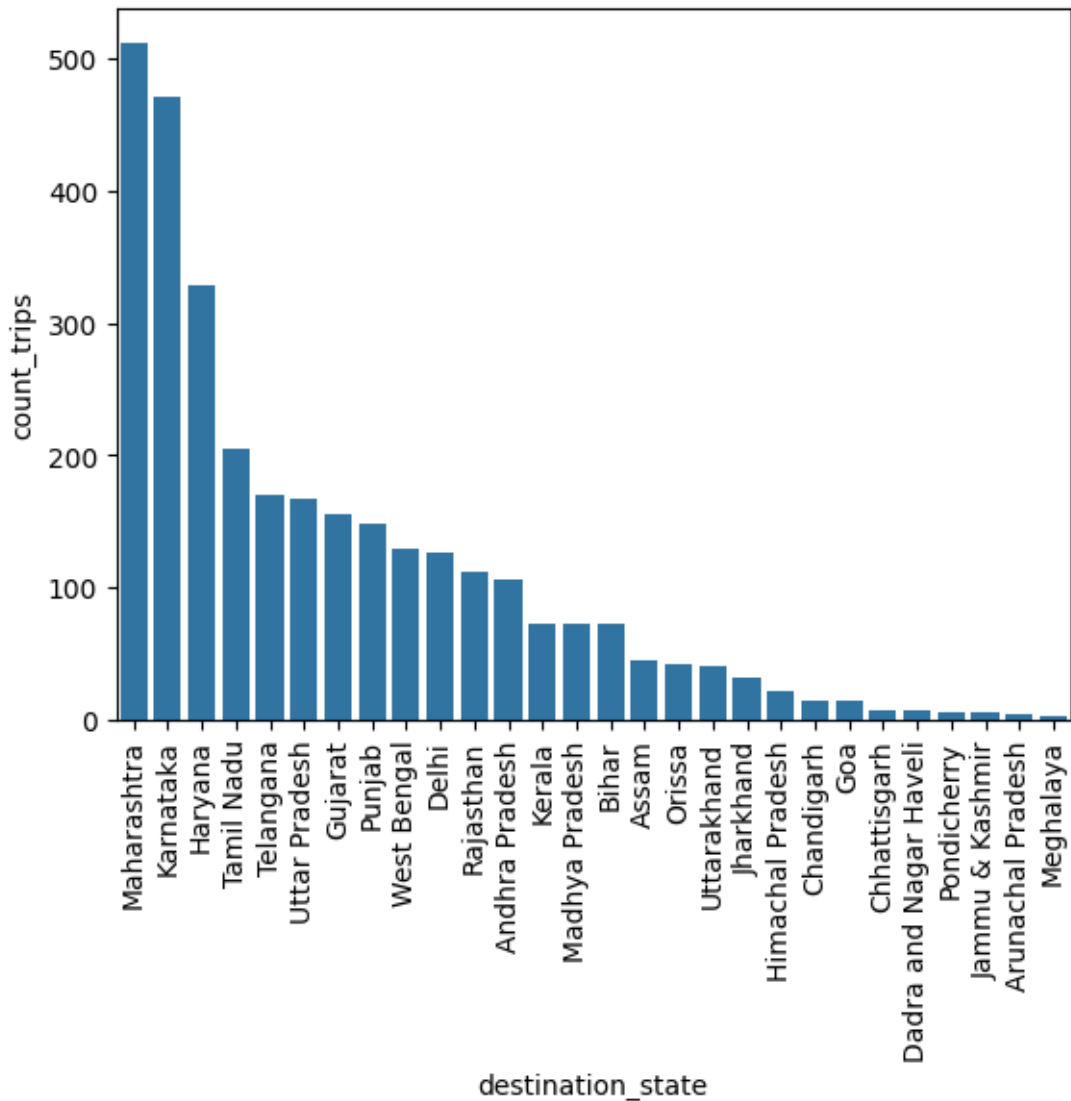
```
[71]: orders_to_df=df.groupby('destination_state').
      ↪agg(count_trips=('trip_uuid','nunique')).reset_index()
orders_to_df=orders_to_df.sort_values(by='count_trips', ascending=False)
orders_to_df
```

```
[71]:
```

	destination_state	count_trips
17	Maharashtra	512
14	Karnataka	471
10	Haryana	329
23	Tamil Nadu	205
24	Telangana	170

25	Uttar Pradesh	167
9	Gujarat	155
21	Punjab	148
27	West Bengal	129
7	Delhi	126
22	Rajasthan	112
0	Andhra Pradesh	106
15	Kerala	73
16	Madhya Pradesh	73
3	Bihar	72
2	Assam	45
19	Orissa	42
26	Uttarakhand	40
13	Jharkhand	32
11	Himachal Pradesh	22
4	Chandigarh	15
8	Goa	14
5	Chhattisgarh	7
6	Dadra and Nagar Haveli	7
20	Pondicherry	6
12	Jammu & Kashmir	5
1	Arunachal Pradesh	4
18	Meghalaya	3

```
[72]: sns.barplot(data=orders_to_df,x='destination_state',y='count_trips')
plt.xticks(rotation=90)
plt.show()
```



Most Busiest corridor

```
[73]: #
      ↳corridor_df=df[['source_center','destination_center','actual_time','osrm_time',
      ↳'osrm_distance']]

df['corridor']=df.apply(lambda x: "#".
      ↳join([x['source_center'],x['destination_center']]), axis=1)
df
```

```
[73]:      data      trip_creation_time \
0      training 2018-09-12 00:06:39.565253
1      training 2018-09-12 00:06:39.565253
```

```

2      training 2018-09-12 00:11:40.783923
3      training 2018-09-12 00:25:19.499696
4      training 2018-09-12 00:25:19.499696
...
27183    test 2018-10-03 23:57:44.429324
27184    test 2018-10-03 23:59:42.701692
27185    test 2018-10-03 23:59:42.701692
27186    test 2018-10-03 23:59:42.701692
27187    test 2018-10-03 23:59:42.701692

```

```

                                route_schedule_uuid route_type \
0      thanos::sroute:a10888ff-f794-41e1-9b7a-7f62ef6...   Carting
1      thanos::sroute:a10888ff-f794-41e1-9b7a-7f62ef6...   Carting
2      thanos::sroute:c2ee580f-f4b2-4fa5-98ab-0c5b327...   Carting
3      thanos::sroute:0ac760f3-96cb-4046-bfd0-8bc4678...     FTL
4      thanos::sroute:0ac760f3-96cb-4046-bfd0-8bc4678...     FTL
...
27183  thanos::sroute:5609c268-e436-4e0a-8180-3db4a74...   Carting
27184  thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...     FTL
27185  thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...     FTL
27186  thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...     FTL
27187  thanos::sroute:412fea14-6d1f-4222-8a5f-a517042...     FTL

```

```

                                trip_uuid source_center \
0      trip-153671079956500691  IND110024AAA
1      trip-153671079956500691  IND110024AAA
2      trip-153671110078355292  IND121004AAB
3      trip-153671191949943656  IND487001AAB
4      trip-153671191949943656  IND487001AAB
...
27183  trip-153861106442901555  IND209304AAA
27184  trip-153861118270144424  IND583119AAA
27185  trip-153861118270144424  IND583119AAA
27186  trip-153861118270144424  IND583201AAA
27187  trip-153861118270144424  IND583201AAA

```

```

                                source_name destination_center \
0      Delhi_Lajpat_IP (Delhi)      IND110014AAA
1      Delhi_Lajpat_IP (Delhi)      IND110014AAA
2      FBD_Balabgarh_DPC (Haryana)  IND121001AAA
3      Narsinghpur_KndliDPP_D (Madhya Pradesh)  IND487551AAA
4      Narsinghpur_KndliDPP_D (Madhya Pradesh)  IND487551AAA
...
27183  Kanpur_Central_H_6 (Uttar Pradesh)  IND208006AAA
27184  Sandur_WrdN1DPP_D (Karnataka)  IND583101AAA
27185  Sandur_WrdN1DPP_D (Karnataka)  IND583101AAA
27186  Hospet (Karnataka)  IND583119AAA

```


	destination_name	od_start_time	...	\
0	Delhi_Bhogal (Delhi)	2018-09-12 00:06:39.565253	...	
1	Delhi_Bhogal (Delhi)	2018-09-12 00:06:39.565253	...	
2	Faridabad (Haryana)	2018-09-12 00:11:40.783923	...	
3	Gadarwara_MPward_D (Madhya Pradesh)	2018-09-12 04:33:09.726898	...	
4	Gadarwara_MPward_D (Madhya Pradesh)	2018-09-12 04:33:09.726898	...	
...	
27183	Kanpur_GovndNgr_DC (Uttar Pradesh)	2018-10-03 23:57:44.429324	...	
27184	Bellary_Dc (Karnataka)	2018-10-04 03:58:40.726547	...	
27185	Bellary_Dc (Karnataka)	2018-10-04 03:58:40.726547	...	
27186	Sandur_WrdN1DPP_D (Karnataka)	2018-10-04 02:51:44.712656	...	
27187	Sandur_WrdN1DPP_D (Karnataka)	2018-10-04 02:51:44.712656	...	

	destination_code	source_code	trip_actual_time_sum	trip_osrm_time_sum	\
0	None	IP	23.0	8.0	
1	None	IP	23.0	8.0	
2	None	DPC	17.0	9.0	
3	D	D	253.0	105.0	
4	D	D	253.0	105.0	
...	
27183	DC	H_6	282.0	48.0	
27184	None	D	275.0	68.0	
27185	None	D	275.0	68.0	
27186	D	None	275.0	68.0	
27187	D	None	275.0	68.0	

	trip_osrm_distance_sum	trip_segment_actual_time_sum	\
0	9.9566	23.0	
1	9.9566	23.0	
2	10.8159	17.0	
3	124.5063	252.0	
4	124.5063	252.0	
...	
27183	58.9037	281.0	
27184	80.5787	274.0	
27185	80.5787	274.0	
27186	80.5787	274.0	
27187	80.5787	274.0	

	trip_segment_osrm_time_sum	trip_segment_osrm_distance_sum	\
0	14.0	16.0860	
1	14.0	16.0860	
2	9.0	10.8159	
3	132.0	140.5623	
4	132.0	140.5623	

...
27183	88.0	104.8866
27184	67.0	80.5787
27185	67.0	80.5787
27186	67.0	80.5787
27187	67.0	80.5787

	trip_total_time_in_hrs	corridor
0	0.822223	IND110024AAA#IND110014AAA
1	0.822223	IND110024AAA#IND110014AAA
2	0.641675	IND121004AAB#IND121001AAA
3	4.844371	IND487001AAB#IND487551AAA
4	4.844371	IND487001AAB#IND487551AAA
...
27183	7.035331	IND209304AAA#IND208006AAA
27184	5.906793	IND583119AAA#IND583101AAA
27185	5.906793	IND583119AAA#IND583101AAA
27186	5.906793	IND583201AAA#IND583119AAA
27187	5.906793	IND583201AAA#IND583119AAA

[27188 rows x 51 columns]

```
[74]: corridor_total_trips=df.groupby(['trip_uuid','corridor']).nth(-1).
      ↳groupby(['corridor']).agg(total_trips=('trip_uuid','nunique')).reset_index()
corridor_total_trips
```

```
[74]:
```

	corridor	total_trips
0	IND000000AAL#IND411033AAA	4
1	IND000000AAS#IND783370AAC	2
2	IND000000ABA#IND683565AAA	2
3	IND000000ABD#IND562132AAA	2
4	IND000000ABG#IND501359AAF	3
...
1791	IND854326AAB#IND854105AAB	3
1792	IND854326AAB#IND854311AAA	5
1793	IND854334AAA#IND852118AAA	2
1794	IND854335AAA#IND852111AAA	3
1795	IND854335AAA#IND854326AAB	1

[1796 rows x 2 columns]

```
[75]: corridor_actual_time_mean_df=df.groupby(['trip_uuid','corridor']).nth(-1).
      ↳groupby(['corridor']).
      ↳agg(corridor_actual_time_mean=('segment_actual_time_cumsum','mean')).
      ↳reset_index()
corridor_actual_time_mean_df
```

```
[75]:
```

	corridor	corridor_actual_time_mean
0	IND000000AAL#IND411033AAA	81.750000
1	IND000000AAS#IND783370AAC	63.000000
2	IND000000ABA#IND683565AAA	48.000000
3	IND000000ABD#IND562132AAA	620.500000
4	IND000000ABG#IND501359AAF	208.666667
...
1791	IND854326AAB#IND854105AAB	56.333333
1792	IND854326AAB#IND854311AAA	68.400000
1793	IND854334AAA#IND852118AAA	26.000000
1794	IND854335AAA#IND852111AAA	37.666667
1795	IND854335AAA#IND854326AAB	197.000000

[1796 rows x 2 columns]

```
[76]: corridor_osrm_time_mean_df=df.groupby(['trip_uuid','corridor']).nth(-1).
      ↪groupby(['corridor']).
      ↪agg(corridor_osrm_time_mean=('segment_osrm_time_cumsum','mean')).
      ↪reset_index()
      corridor_osrm_time_mean_df
```

```
[76]:
```

	corridor	corridor_osrm_time_mean
0	IND000000AAL#IND411033AAA	28.500000
1	IND000000AAS#IND783370AAC	30.000000
2	IND000000ABA#IND683565AAA	20.000000
3	IND000000ABD#IND562132AAA	402.500000
4	IND000000ABG#IND501359AAF	33.666667
...
1791	IND854326AAB#IND854105AAB	19.666667
1792	IND854326AAB#IND854311AAA	29.400000
1793	IND854334AAA#IND852118AAA	21.000000
1794	IND854335AAA#IND852111AAA	19.000000
1795	IND854335AAA#IND854326AAB	82.000000

[1796 rows x 2 columns]

```
[77]: corridor_osrm_distance_mean_df=df.groupby(['trip_uuid','corridor']).nth(-1).
      ↪groupby(['corridor']).
      ↪agg(corridor_osrm_distance_mean=('segment_osrm_distance_cumsum','mean')).
      ↪reset_index()
      corridor_osrm_distance_mean_df
```

```
[77]:
```

	corridor	corridor_osrm_distance_mean
0	IND000000AAL#IND411033AAA	28.299475
1	IND000000AAS#IND783370AAC	42.577350
2	IND000000ABA#IND683565AAA	16.638750
3	IND000000ABD#IND562132AAA	443.786100

4	IND000000ABG#IND501359AAF	35.057133
...
1791	IND854326AAB#IND854105AAB	29.209900
1792	IND854326AAB#IND854311AAA	42.575020
1793	IND854334AAA#IND852118AAA	23.600650
1794	IND854335AAA#IND852111AAA	27.882833
1795	IND854335AAA#IND854326AAB	109.264800

[1796 rows x 2 columns]

```
[78]: corridor_aggregated_df=pd.concat(
    objs=(
        iDF.set_index('corridor') for iDF in (corridor_total_trips,
        corridor_actual_time_mean_df,
        corridor_osrm_distance_mean_df,
        corridor_osrm_time_mean_df
    ),
    axis=1,
    join='inner'
).reset_index()
corridor_aggregated_df
```

[78]:	corridor	total_trips	corridor_actual_time_mean \
0	IND000000AAL#IND411033AAA	4	81.750000
1	IND000000AAS#IND783370AAC	2	63.000000
2	IND000000ABA#IND683565AAA	2	48.000000
3	IND000000ABD#IND562132AAA	2	620.500000
4	IND000000ABG#IND501359AAF	3	208.666667
...
1791	IND854326AAB#IND854105AAB	3	56.333333
1792	IND854326AAB#IND854311AAA	5	68.400000
1793	IND854334AAA#IND852118AAA	2	26.000000
1794	IND854335AAA#IND852111AAA	3	37.666667
1795	IND854335AAA#IND854326AAB	1	197.000000

	corridor_osrm_distance_mean	corridor_osrm_time_mean
0	28.299475	28.500000
1	42.577350	30.000000
2	16.638750	20.000000
3	443.786100	402.500000
4	35.057133	33.666667
...
1791	29.209900	19.666667
1792	42.575020	29.400000
1793	23.600650	21.000000

1794	27.882833	19.000000
1795	109.264800	82.000000

[1796 rows x 5 columns]

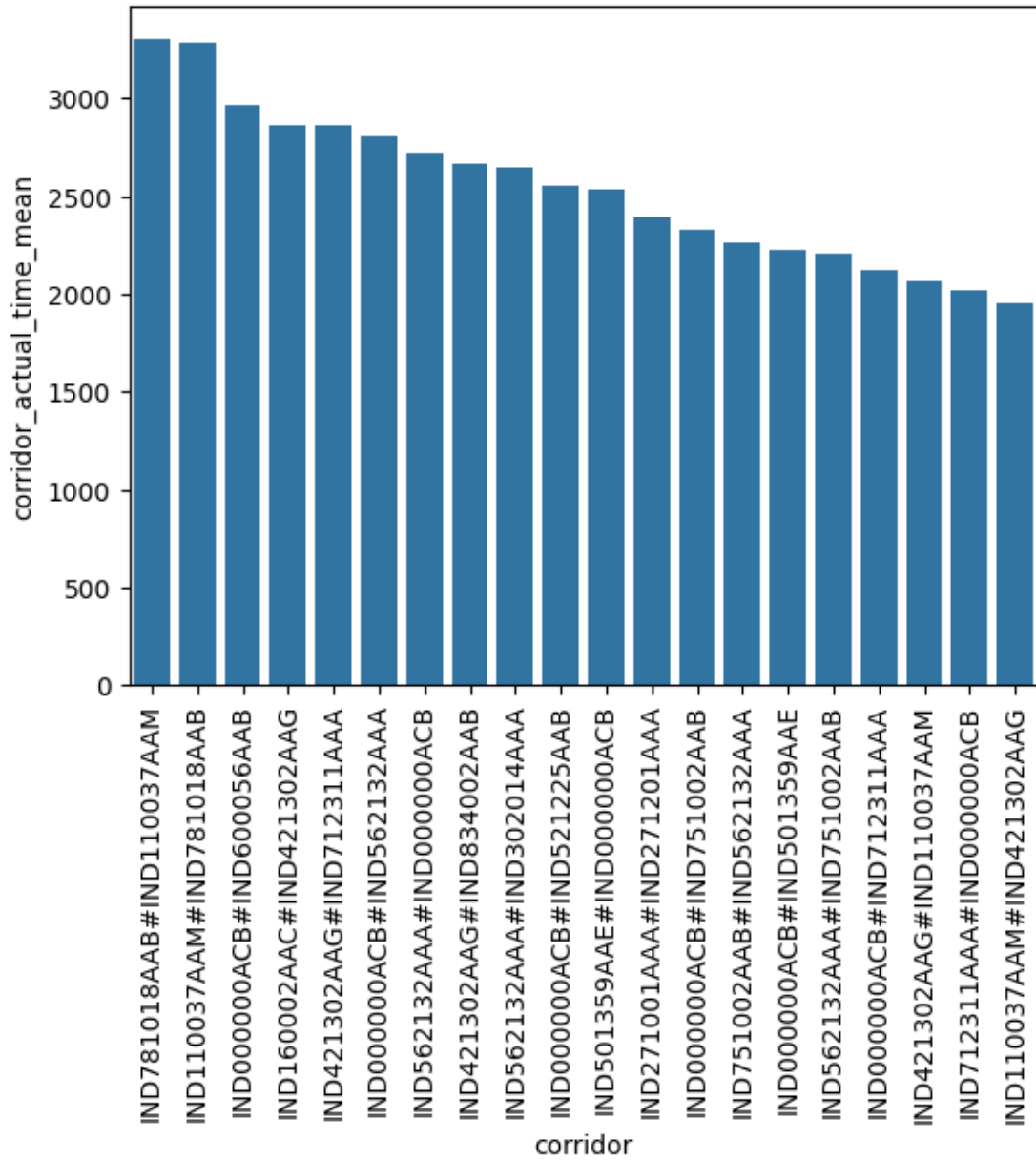
Busiest corridor Actual time

```
[79]: top_20trip_time_corridor_df=corridor_aggregated_df[['corridor','corridor_actual_time_mean']].
      ↪sort_values(by='corridor_actual_time_mean', ascending=False)[:20]
      top_20trip_time_corridor_df
```

```
[79]:
```

	corridor	corridor_actual_time_mean
1638	IND781018AAB#IND110037AAM	3302.500000
86	IND110037AAM#IND781018AAB	3281.000000
44	IND000000ACB#IND600056AAB	2963.000000
222	IND160002AAC#IND421302AAG	2867.000000
765	IND421302AAG#IND712311AAA	2867.000000
43	IND000000ACB#IND562132AAA	2808.642857
1130	IND562132AAA#IND000000ACB	2721.333333
766	IND421302AAG#IND834002AAB	2667.000000
1132	IND562132AAA#IND302014AAA	2651.000000
42	IND000000ACB#IND521225AAB	2550.333333
899	IND501359AAE#IND000000ACB	2536.000000
402	IND271001AAA#IND271201AAA	2398.000000
46	IND000000ACB#IND751002AAB	2323.750000
1601	IND751002AAB#IND562132AAA	2264.000000
41	IND000000ACB#IND501359AAE	2222.500000
1158	IND562132AAA#IND751002AAB	2208.000000
45	IND000000ACB#IND712311AAA	2121.600000
742	IND421302AAG#IND110037AAM	2064.500000
1515	IND712311AAA#IND000000ACB	2021.000000
85	IND110037AAM#IND421302AAG	1952.666667

```
[80]: sns.
      ↪barplot(data=top_20trip_time_corridor_df,x='corridor',y='corridor_actual_time_mean')
      plt.xticks(rotation=90)
      plt.show()
```



```
[81]: corridor_aggregated_df[['corridor_actual_time_mean', 'corridor_osrm_distance_mean', 'corridor_dsrm_time_mean', 'total_trips']
      <-corr()
```

```
[81]:
```

	corridor_actual_time_mean \
corridor_actual_time_mean	1.000000
corridor_osrm_distance_mean	0.937734
corridor_dsrm_time_mean	0.934818
total_trips	-0.011093

```

corridor_osrm_distance_mean \

```

corridor_actual_time_mean	0.937734
corridor_osrm_distance_mean	1.000000
corridor_dsrm_time_mean	0.995915
total_trips	0.007006

	corridor_dsrm_time_mean	total_trips
corridor_actual_time_mean	0.934818	-0.011093
corridor_osrm_distance_mean	0.995915	0.007006
corridor_dsrm_time_mean	1.000000	0.008240
total_trips	0.008240	1.000000

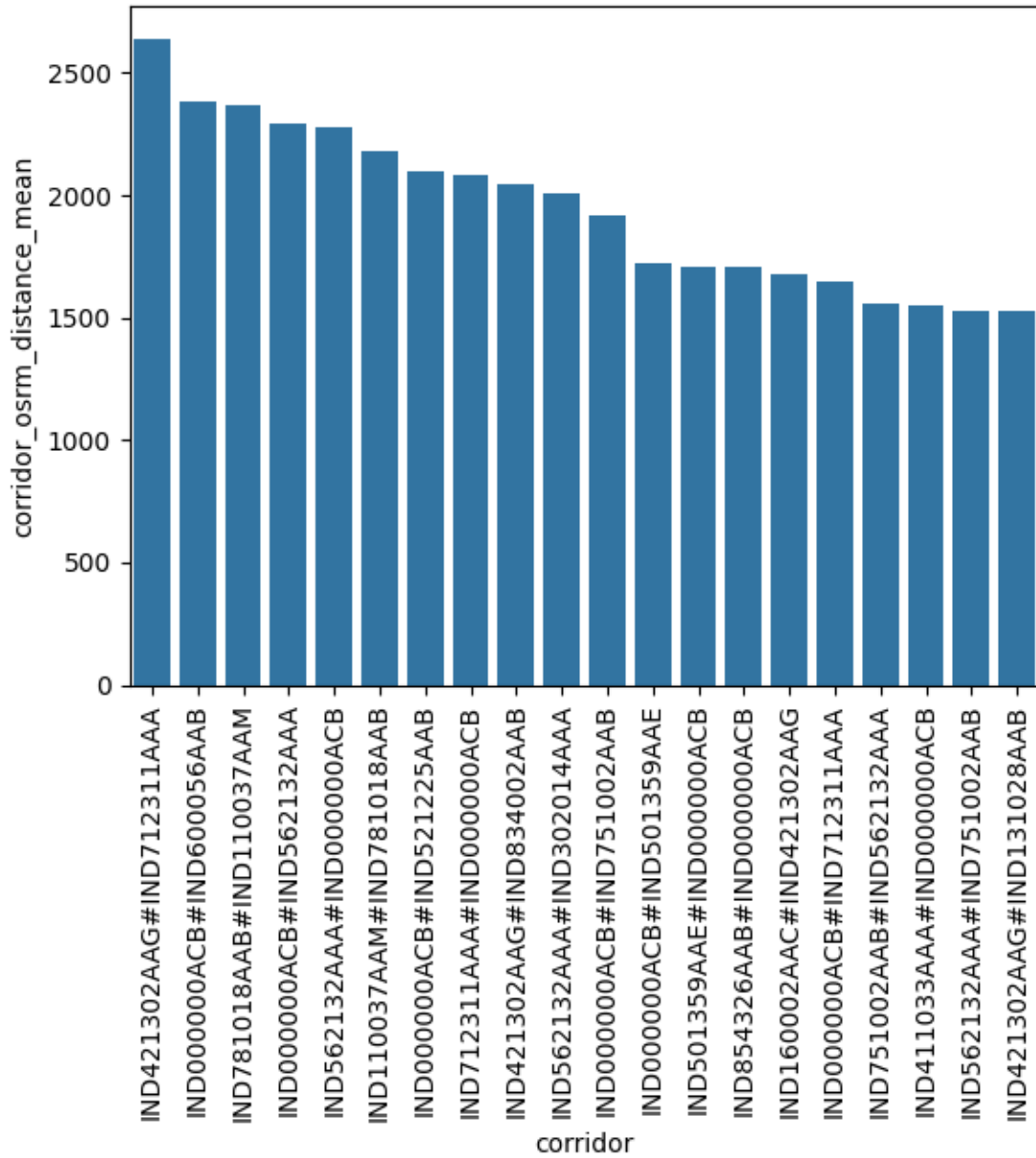
Busiest Corridor by Distance

```
[82]: top_20trip_dist_corridor_df=corridor_aggregated_df[['corridor','corridor_osrm_distance_mean']]
      ↪sort_values(by='corridor_osrm_distance_mean', ascending=False)[:20]
      top_20trip_dist_corridor_df
```

```
[82]:
```

	corridor	corridor_osrm_distance_mean
765	IND421302AAG#IND712311AAA	2640.924700
44	IND000000ACB#IND600056AAB	2384.514700
1638	IND781018AAB#IND110037AAM	2367.129350
43	IND000000ACB#IND562132AAA	2289.301729
1130	IND562132AAA#IND000000ACB	2276.539317
86	IND110037AAM#IND781018AAB	2181.460700
42	IND000000ACB#IND521225AAB	2099.734933
1515	IND712311AAA#IND000000ACB	2079.612400
766	IND421302AAG#IND834002AAB	2046.679900
1132	IND562132AAA#IND302014AAA	2010.760400
46	IND000000ACB#IND751002AAB	1916.699800
41	IND000000ACB#IND501359AAE	1721.693467
899	IND501359AAE#IND000000ACB	1708.819400
1786	IND854326AAB#IND000000ACB	1707.401300
222	IND160002AAC#IND421302AAG	1679.658000
45	IND000000ACB#IND712311AAA	1645.840390
1601	IND751002AAB#IND562132AAA	1557.065950
690	IND411033AAA#IND000000ACB	1553.552800
1158	IND562132AAA#IND751002AAB	1527.861300
743	IND421302AAG#IND131028AAB	1524.396400

```
[83]: sns.
      ↪barplot(data=top_20trip_dist_corridor_df,x='corridor',y='corridor_osrm_distance_mean')
      plt.xticks(rotation=90)
      plt.show()
```



Business Recommendations

Maharashtra, Karnataka, Tamil Nadu, Utter Pradesh, Telangana and Gujarat States are states where most delivery trips are done. Most Bussiest corridor are in these states.

Business should focus on identifying best corridors to move packages very quickly, they should focus on potential reasons for difference in actual delivery time and osrm delivery time value.

If Actual delivery time is higher than osrm time then should focus on hops which are causing delays, if delays are related to processing or logistic that should be quickly fixed.

If Issue is not related to delivery and logistic process then should focus on identifying best route to

move packages quickly.