```
In [1]: import numpy as np
    import pandas as pd
    import matplotlib.pyplot as plt
    import seaborn as sns
    import datetime
    sns.set()
```

```
/opt/conda/lib/python3.10/site-packages/scipy/__init__.py:146: UserWarning: A NumPy version >=1.16.5 and <1. 23.0 is required for this version of SciPy (detected version 1.23.5 warnings.warn(f"A NumPy version >={np_minversion} and <{np_maxversion}"
```

This is taxi fare data where we are analysing the highest trip days finding out the outliers

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```
In [5]: df = pd.read_csv('/kaggle/input/nyc-taxi-trip-data-google-public-data/taxi_trip_data.csv')
```

```
In [6]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000000 entries, 0 to 9999999
Data columns (total 17 columns):
     Column
                          Dtype
     -----
                          ----
     vendor_id
                         int64
     pickup_datetime
                         object
 1
     dropoff_datetime
                         object
     passenger_count
                         int64
 3
    trip_distance
 4
                         float64
 5
    rate_code
                         int64
 6
    store_and_fwd_flag
                         object
 7
     payment_type
                         int64
 8
    fare_amount
                         float64
 9
                         float64
     extra
                         float64
 10 mta_tax
 11 tip amount
                         float64
 12 tolls_amount
                         float64
 13 imp_surcharge
                         float64
 14 total_amount
                         float64
 15 pickup location id
                         int64
 16 dropoff_location_id int64
dtypes: float64(8), int64(6), object(3)
memory usage: 1.3+ GB
```

In [7]: df.describe()

Out[7]:

	vendor_id	passenger_count	trip_distance	rate_code	payment_type	fare_amount	extra	mta_tax	tip_a
count	1.000000e+07	1.000000e+07	1.000000e+07	1.000000e+07	1.000000e+07	1.000000e+07	1.000000e+07	1.000000e+07	1.0000
mean	1.614328e+00	1.602949e+00	8.849280e+00	1.201239e+00	1.189299e+00	3.165255e+01	3.383781e-01	4.819289e-01	5.5985
std	5.146576e-01	1.245782e+00	5.882028e+00	1.250733e+00	4.339876e-01	1.606011e+02	5.512911e-01	1.207282e-01	4.8405
min	1.000000e+00	0.000000e+00	0.000000e+00	1.000000e+00	1.000000e+00	-8.000000e+02	-8.000000e+01	-5.000000e- 01	-3.2242
25%	1.000000e+00	1.000000e+00	5.820000e+00	1.000000e+00	1.000000e+00	2.350000e+01	0.000000e+00	5.000000e-01	2.0000
50%	2.000000e+00	1.000000e+00	8.480000e+00	1.000000e+00	1.000000e+00	2.850000e+01	0.000000e+00	5.000000e-01	5.5600
75%	2.000000e+00	2.000000e+00	1.110000e+01	1.000000e+00	1.000000e+00	3.700000e+01	5.000000e-01	5.000000e-01	7.9600
max	4.000000e+00	9.000000e+00	7.655760e+03	9.900000e+01	5.000000e+00	3.984600e+05	8.400000e+01	1.500000e+02	4.9600

In [8]: df.isnull().sum()

Out[8]: vendor_id

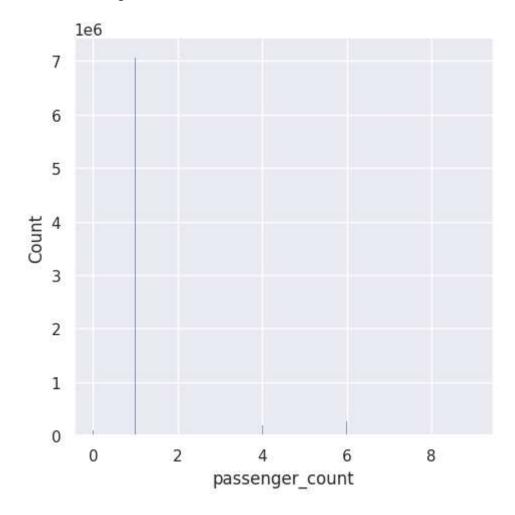
0 pickup_datetime 0 dropoff_datetime 0 passenger_count 0 trip_distance 0 rate_code store_and_fwd_flag payment_type 0 fare_amount 0 extra mta_tax tip_amount 0 tolls_amount 0 imp_surcharge total_amount pickup_location_id 0 dropoff_location_id 0 dtype: int64

```
In [9]: df.head()
 Out[9]:
              vendor_id pickup_datetime dropoff_datetime passenger_count trip_distance rate_code store_and_fwd_flag payment_type fare_am
                             2018-03-29
                                             2018-03-29
           0
                     2
                                                                              18.15
                                                                                           3
                                                                     1
                                                                                                             Ν
                                                                                                                          1
                               13:37:13
                                               14:17:01
                             2018-03-29
                                             2018-03-29
           1
                     2
                                                                     1
                                                                               4.59
                                                                                           1
                                                                                                             Ν
                                                                                                                           1
                               13:37:18
                                               14:15:33
                             2018-03-29
                                             2018-03-29
           2
                     2
                                                                     1
                                                                               0.30
                                                                                           1
                                                                                                             Ν
                                                                                                                           1
                               13:26:57
                                               13:28:03
                             2018-03-29
                                             2018-03-29
           3
                     2
                                                                     2
                                                                              16.97
                                                                                           1
                                                                                                             Ν
                                                                                                                           1
                               13:07:48
                                               14:03:05
                             2018-03-29
                                             2018-03-29
                     2
                                                                     5
                                                                                                             Ν
                                                                              14.45
                                                                                           1
                                                                                                                          1
                                               15:19:59
                               14:19:11
In [10]: df.columns
Out[10]: Index(['vendor_id', 'pickup_datetime', 'dropoff_datetime', 'passenger_count',
                  'trip_distance', 'rate_code', 'store_and_fwd_flag', 'payment_type',
                  'fare_amount', 'extra', 'mta_tax', 'tip_amount', 'tolls_amount',
                  'imp surcharge', 'total amount', 'pickup location id',
                  'dropoff location id'],
                 dtype='object')
          df['pickup datetime']=pd.to datetime(df['pickup datetime'])
In [11]:
          df['dropoff datetime']=pd.to datetime(df['dropoff datetime'])
```

We will now create a displot which is used to show univariant set of collected data show passenger count data distribution as we want to show one variable against another variable.

```
In [13]: sns.displot(df['passenger_count'])
```

Out[13]: <seaborn.axisgrid.FacetGrid at 0x7de8a7a95b10>



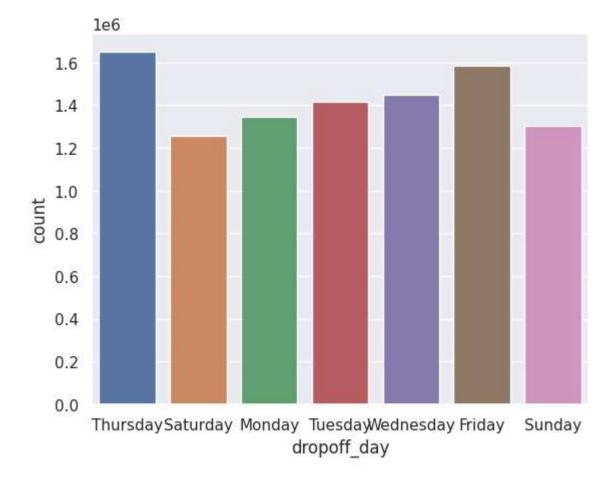
We will now convert the dates into days of the week to see on which day the most and least trips are taken

```
In [20]: df['pickup_day']=df['pickup_datetime'].dt.day_name()
df['dropoff_day']=df['dropoff_datetime'].dt.day_name()
```

```
In [21]: df['pickup_day'].value_counts()
Out[21]: Thursday
                      1664099
         Friday
                      1588090
         Wednesday
                      1451808
         Tuesday
                      1422420
         Monday
                      1343747
         Sunday
                      1284224
         Saturday
                      1245612
         Name: pickup_day, dtype: int64
In [22]: df['dropoff_day'].value_counts()
Out[22]: Thursday
                      1650531
         Friday
                      1584692
         Wednesday
                      1447014
         Tuesday
                      1418264
         Monday
                      1343468
         Sunday
                      1301192
         Saturday
                      1254839
         Name: dropoff_day, dtype: int64
```

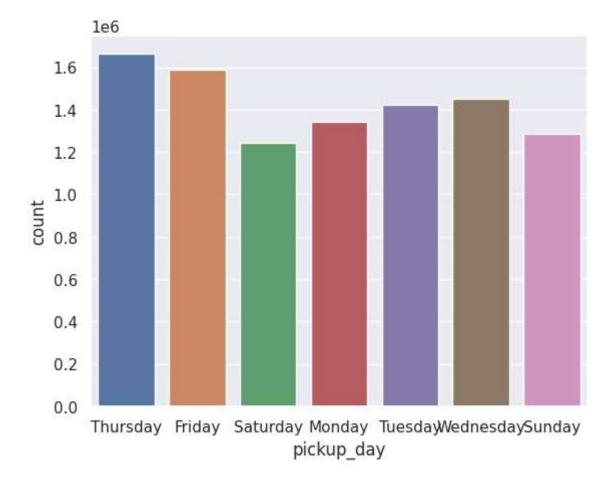
```
In [24]: sns.countplot(x='dropoff_day', data=df)
```

Out[24]: <AxesSubplot: xlabel='dropoff_day', ylabel='count'>



```
In [25]: sns.countplot(x='pickup_day', data=df)
```

Out[25]: <AxesSubplot: xlabel='pickup_day', ylabel='count'>



We found out that maximum trips were taken on Thursday and least were taken on Saturday

```
In [29]: df['pickup_time']=df['pickup_datetime'].dt.time
In [30]: df['dropoff_time']=df['dropoff_datetime'].dt.time
```

```
In [32]: df.dtypes
Out[32]: vendor id
                                          int64
         pickup datetime
                                 datetime64[ns]
         dropoff datetime
                                 datetime64[ns]
         passenger count
                                          int64
         trip_distance
                                        float64
         rate code
                                          int64
         store_and_fwd_flag
                                         object
                                          int64
         payment type
                                        float64
         fare_amount
         extra
                                        float64
         mta_tax
                                        float64
         tip_amount
                                        float64
         tolls_amount
                                        float64
         imp surcharge
                                        float64
         total amount
                                        float64
         pickup location id
                                          int64
         dropoff location id
                                          int64
         pickup_day
                                         object
         dropoff_day
                                         object
         pickup_time
                                         object
         dropoff time
                                         object
         dtype: object
         def timezone(x):
In [35]:
             if x > = datetime.time(4, 0, 1) and x < = datetime.time(10, 0, 0):
                 return 'morning'
             elif x>=datetime.time(10, 0, 1) and x <=datetime.time(16, 0, 0):
                  return 'midday'
             elif x>=datetime.time(16, 0, 1) and x <=datetime.time(22, 0, 0):
                 return 'evening'
             elif x>=datetime.time(22, 0, 1) or x <=datetime.time(4, 0, 0):
                 return 'late night'
         df['pickup timezone']=df['pickup datetime'].apply(lambda x :timezone(datetime.datetime.strptime(str(x), "%Y-%
         df['dropoff timezone']=df['dropoff datetime'].apply(lambda x :timezone(datetime.datetime.strptime(str(x),
```

In [36]: df.head()

Out[36]:

	vendor_id	pickup_datetime	dropoff_datetime	passenger_count	trip_distance	rate_code	store_and_fwd_flag	payment_type	fare_am
0	2	2018-03-29 13:37:13	2018-03-29 14:17:01	1	18.15	3	N	1	
1	2	2018-03-29 13:37:18	2018-03-29 14:15:33	1	4.59	1	N	1	
2	2	2018-03-29 13:26:57	2018-03-29 13:28:03	1	0.30	1	N	1	
3	2	2018-03-29 13:07:48	2018-03-29 14:03:05	2	16.97	1	N	1	
4	2	2018-03-29 14:19:11	2018-03-29 15:19:59	5	14.45	1	N	1	

5 rows × 23 columns

In [39]: df['duration']=df['dropoff_datetime'] - df['pickup_datetime']

In [42]: df.drop(['trip_duration'], axis=1)

Out[42]:

	vendor_id	pickup_datetime	dropoff_datetime	passenger_count	trip_distance	rate_code	store_and_fwd_flag	payment_type f
0	2	2018-03-29 13:37:13	2018-03-29 14:17:01	1	18.15	3	N	1
1	2	2018-03-29 13:37:18	2018-03-29 14:15:33	1	4.59	1	N	1
2	2	2018-03-29 13:26:57	2018-03-29 13:28:03	1	0.30	1	N	1
3	2	2018-03-29 13:07:48	2018-03-29 14:03:05	2	16.97	1	N	1
4	2	2018-03-29 14:19:11	2018-03-29 15:19:59	5	14.45	1	N	1
9999995	2	2018-03-29 12:16:01	2018-03-29 13:03:31	1	3.45	1	N	1
9999996	1	2018-03-29 12:26:25	2018-03-29 13:09:54	1	6.80	1	N	1
9999997	2	2018-03-29 12:22:12	2018-03-29 13:01:07	2	8.46	1	N	1
9999998	2	2018-03-29 13:57:40	2018-03-29 15:07:42	1	14.43	1	N	1
9999999	2	2018-03-29 13:29:40	2018-03-29 14:02:51	1	4.42	1	N	1
10000000	0 rows × 24	columns						

localhost:8888/notebooks/himanshi-taxi-fare-univariate-analysis.ipynb

In [43]: df.head()

Out[43]:

	vendor_id	pickup_datetime	dropoff_datetime	passenger_count	trip_distance	rate_code	store_and_fwd_flag	payment_type	fare_am
0	2	2018-03-29 13:37:13	2018-03-29 14:17:01	1	18.15	3	N	1	
1	2	2018-03-29 13:37:18	2018-03-29 14:15:33	1	4.59	1	N	1	
2	2	2018-03-29 13:26:57	2018-03-29 13:28:03	1	0.30	1	N	1	
3	2	2018-03-29 13:07:48	2018-03-29 14:03:05	2	16.97	1	N	1	
4	2	2018-03-29 14:19:11	2018-03-29 15:19:59	5	14.45	1	N	1	

5 rows × 25 columns

4

In [51]: df.drop(["duration"], axis = 1, inplace = True)

In [52]: df.head()

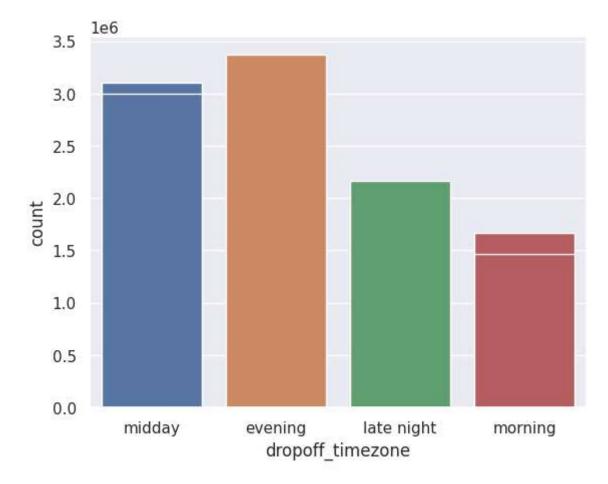
Out[52]:

	vendor_id	pickup_datetime	dropoff_datetime	passenger_count	trip_distance	rate_code	store_and_fwd_flag	payment_type	fare_am
0	2	2018-03-29 13:37:13	2018-03-29 14:17:01	1	18.15	3	N	1	
1	2	2018-03-29 13:37:18	2018-03-29 14:15:33	1	4.59	1	N	1	
2	2	2018-03-29 13:26:57	2018-03-29 13:28:03	1	0.30	1	N	1	
3	2	2018-03-29 13:07:48	2018-03-29 14:03:05	2	16.97	1	N	1	
4	2	2018-03-29 14:19:11	2018-03-29 15:19:59	5	14.45	1	N	1	

5 rows × 24 columns

```
In [56]: sns.countplot(x='pickup_timezone', data=df)
sns.countplot(x='dropoff_timezone', data=df)
```

Out[56]: <AxesSubplot: xlabel='dropoff_timezone', ylabel='count'>



```
In [57]: df.dtypes
Out[57]: vendor_id
                                           int64
         pickup_datetime
                                  datetime64[ns]
         dropoff_datetime
                                  datetime64[ns]
         passenger_count
                                           int64
         trip_distance
                                         float64
         rate_code
                                           int64
         store_and_fwd_flag
                                          object
         payment_type
                                           int64
         fare_amount
                                         float64
         extra
                                         float64
         mta_tax
                                         float64
         tip_amount
                                         float64
         tolls_amount
                                         float64
         imp_surcharge
                                         float64
         total amount
                                         float64
         pickup_location_id
                                           int64
         dropoff_location_id
                                           int64
         pickup_day
                                          object
         dropoff_day
                                          object
         pickup_time
                                          object
         dropoff time
                                          object
         pickup_timezone
                                          object
         dropoff timezone
                                          object
         trip duration
                                 timedelta64[ns]
         dtype: object
In [60]: | df['trip_duration_hour']=df['trip_duration']/3600
In [62]: df['triphours'] = df['trip duration hour'] / pd.Timedelta(hours=1)
```