D20

In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

In [2]: df=pd.read_csv(r"C:\Users\user\Downloads\23_Vande Bharat.csv")
df

Out[2]:

| | Sr. No. | Train Name | Train Number | Originating City | Originating Station | Terminal City | Т |
|----|------------|--|-----------------|---------------------------|------------------------------------|--------------------|-------|
| 0 | 1 | New Delhi - Varanasi Vande Bharat Express | 22435/22436 | Delhi | New Delhi | Varanasi | V |
| 1 | 2 | New Delhi - Shri Mata Vaishno Devi Katra Vande | 22439/22440 | Delhi | New Delhi | Katra | SI |
| 2 | 3 | Mumbai Central - Gandhinagar Capital Vande Bha | 20901/20902 | Mumbai | Mumbai Central | Gandhinagar | Gan |
| 3 | 4 | New Delhi - Amb Andaura Vande Bharat Express | 22447/22448 | Delhi | New Delhi | Andaura | |
| 4 | 5 | MGR Chennai Central - Mysuru Vande Bharat Express | 20607/20608 | Chennai | Chennai Central | Mysuru | |
| 5 | 6 | Bilaspur - Nagpur Vande Bharat Express | 20825/20826 | Bilaspur, Chhattisgarh | Bilaspur Junction | Nagpur | |
| 6 | 7 | Howrah - New Jalpaiguri Vande Bharat Express | 22301/22302 | Kolkata | Howrah Junction | Siliguri | |
| 7 | 8 | Visakhapatnam - Secunderabad Vande Bharat Express | 20833/20834 | Visakhapatnam | Visakhapatnam Junction | Hyderabad | |
| 8 | 9 | Mumbai CSMT - Solapur Vande Bharat Express | 22225/22226 | Mumbai | Chhatrapati Shivaji Terminus | Solapur | |
| 9 | 10 | Mumbai CSMT - Sainagar Shirdi Vande Bharat Exp | 22223/22224 | Mumbai | Chhatrapati Shivaji Terminus | Shirdi | |
| 10 | 11 | Rani Kamalapati (Habibganj) - Hazrat Nizamuddi | 20171/20172 | Bhopal | Habibganj (Rani Kamalapati) | Delhi | На |
| 11 | 12 | Secunderabad - Tirupati Vande Bharat Express | 20701/20702 | Hyderabad | Secunderabad Junction | Tirupati | |
| 12 | 13 | MGR Chennai Central - Coimbatore Vande Bharat | 20643/20644 | Chennai | Chennai Central | Coimbatore | Coir |
| 13 | 14 | Delhi Cantonment - Ajmer Vande Bharat Express | 20977/20978 | Delhi | Delhi Cantonment | Ajmer | |
| 14 | 15 | Kasaragod - Thiruvananthapuram Vande Bharat Ex | 20633/20634 | Kasaragod | Kasaragod | Thiruvananthapuram | Thirı |
| 15 | 16 | Howrah - Puri Vande Bharat Express | 22895/22896 | Kolkata | Howrah Junction | Puri | |

| | Sr. No. | Train Name | Train Number | Originating City | Originating Station | Terminal City | т |
|----|------------|--|-----------------|---------------------|------------------------------------|-------------------|-----|
| 16 | 17 | Anand Vihar Terminal - Dehradun Vande Bharat E | 22457/22458 | Delhi | Anand Vihar Terminal | Dehradun | De |
| 17 | 18 | New Jalpaiguri - Guwahati Vande Bharat Express | 22227/22228 | Siliguri | New Jalpaiguri Junction | Guwahati | |
| 18 | 19 | Mumbai CSMT - Madgaon Vande Bharat Express | 22229/22230 | Mumbai | Chhatrapati Shivaji Terminus | Madgaon | M |
| 19 | 19 | Mumbai CSMT - Madgaon Vande Bharat Express | 22229/22230 | Mumbai | Chhatrapati Shivaji Terminus | Madgaon | M |
| 20 | 20 | Patna - Ranchi Vande Bharat Express | 22349/22350 | Patna | Patna Junction | Ranchi | |
| 21 | 21 | KSR Bengaluru - Dharwad Vande Bharat Express | 20661/20662 | Bangalore | Bangalore City | Hubbali - Dharwad | |
| 22 | 22 | Rani Kamalapati (Habibganj) - Jabalpur Vande B | 20173/20174 | Bhopal | Habibganj (Rani Kamalapati) | Jabalpur | J |
| 23 | 23 | Indore - Bhopal Vande Bharat Express | 20911/20912 | Indore | Indore Junction | Bhopal | |
| 24 | 24 | Jodhpur - Sabarmati (Ahmedabad) Vande Bharat E | 12461/12462 | Jodhpur | Jodhpur Junction | Ahmedabad | Sa |
| 25 | 25 | Gorakhpur - Lucknow Charbagh Vande Bharat Express | 22549/22550 | Gorakhpur | Gorakhpur Junction | Charbagh | Luc |

In [3]: df.head(10)

Out[3]:

| | Sr. No. | Train Name | Train Number | Originating City | Originating Station | Terminal City | Terminal Station | 0 |
|---|------------|--|-----------------|---------------------------|------------------------------------|------------------|------------------------------------|---|
| 0 | 1 | New Delhi - Varanasi Vande Bharat Express | 22435/22436 | Delhi | New Delhi | Varanasi | Varanasi Junction | |
| 1 | 2 | New Delhi - Shri Mata Vaishno Devi Katra Vande | 22439/22440 | Delhi | New Delhi | Katra | Shri Mata Vaishno Devi Katra | |
| 2 | 3 | Mumbai Central - Gandhinagar Capital Vande Bha | 20901/20902 | Mumbai | Mumbai Central | Gandhinagar | Gandhinagar Capital | |
| 3 | 4 | New Delhi - Amb Andaura Vande Bharat Express | 22447/22448 | Delhi | New Delhi | Andaura | Amb Andaura | |
| 4 | 5 | MGR Chennai Central - Mysuru Vande Bharat Express | 20607/20608 | Chennai | Chennai Central | Mysuru | Mysore Junction | |
| 5 | 6 | Bilaspur - Nagpur Vande Bharat Express | 20825/20826 | Bilaspur, Chhattisgarh | Bilaspur Junction | Nagpur | Nagpur Junction | |
| 6 | 7 | Howrah - New Jalpaiguri Vande Bharat Express | 22301/22302 | Kolkata | Howrah Junction | Siliguri | New Jalpaiguri Junction | |
| 7 | 8 | Visakhapatnam Secunderabad Vande Bharat Express | 20833/20834 | Visakhapatnam | Visakhapatnam Junction | Hyderabad | Secunderabad Junction | |
| 8 | 9 | Mumbai CSMT - Solapur Vande Bharat Express | 22225/22226 | Mumbai | Chhatrapati Shivaji Terminus | Solapur | Solapur | |
| 9 | 10 | Mumbai CSMT - Sainagar Shirdi Vande Bharat Exp | 22223/22224 | Mumbai | Chhatrapati Shivaji Terminus | Shirdi | Sainagar Shirdi | |
| 4 | | | | | | | | • |

```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 26 entries, 0 to 25
Data columns (total 16 columns):
```

| # | Column | Nor | n-Null Count | Dtype | | |
|---|---------------------|-----|--------------|--------|--|--|
| | | | | | | |
| 0 | Sr. No. | 26 | non-null | int64 | | |
| 1 | Train Name | 26 | non-null | object | | |
| 2 | Train Number | 26 | non-null | object | | |
| 3 | Originating City | 26 | non-null | object | | |
| 4 | Originating Station | 26 | non-null | object | | |
| 5 | Terminal City | 26 | non-null | object | | |
| 6 | Terminal Station | 26 | non-null | object | | |
| 7 | Operator | 26 | non-null | object | | |
| 8 | No. of Cars | 26 | non-null | int64 | | |
| 9 | Frequency | 26 | non-null | object | | |
| 10 | Distance | 26 | non-null | object | | |
| 11 | Travel Time | 26 | non-null | object | | |
| 12 | Speed | 26 | non-null | object | | |
| 13 | Average Speed | 26 | non-null | object | | |
| 14 | Inauguration | 26 | non-null | object | | |
| 15 | Average occupancy | 26 | non-null | object | | |
| <pre>dtypes: int64(2), object(14)</pre> | | | | | | |
| | | | | | | |

memory usage: 3.4+ KB

In [6]: df.describe()

Out[6]:

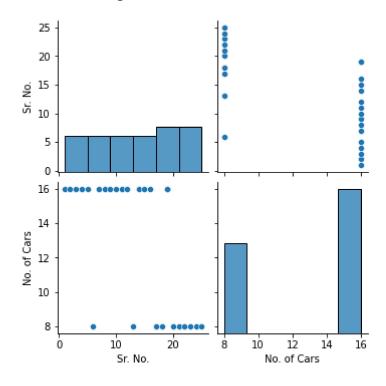
| | Sr. No. | No. of Cars |
|-------|-----------|-------------|
| count | 26.000000 | 26.000000 |
| mean | 13.230769 | 12.923077 |
| std | 7.306478 | 3.969112 |
| min | 1.000000 | 8.000000 |
| 25% | 7.250000 | 8.000000 |
| 50% | 13.500000 | 16.000000 |
| 75% | 19.000000 | 16.000000 |
| max | 25.000000 | 16.000000 |

```
In [7]: df.columns
```

```
Out[7]: Index(['Sr. No.', 'Train Name', 'Train Number', 'Originating City',
                'Originating Station', 'Terminal City', 'Terminal Station', 'Operato
        r',
               'No. of Cars', 'Frequency', 'Distance', 'Travel Time', 'Speed',
               'Average Speed', 'Inauguration', 'Average occupancy'],
              dtype='object')
```

In [8]: sns.pairplot(df)

Out[8]: <seaborn.axisgrid.PairGrid at 0x1ec95455c10>

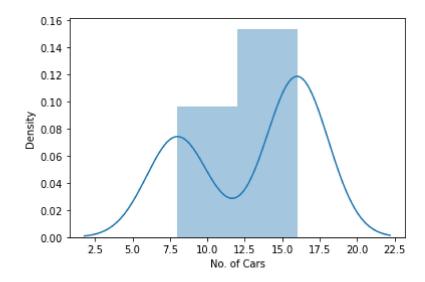


In [12]: sns.distplot(df['No. of Cars'])

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: Fut ureWarning: `distplot` is a deprecated function and will be removed in a futu re version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[12]: <AxesSubplot:xlabel='No. of Cars', ylabel='Density'>



```
In [13]: df1=df[['Sr. No.','No. of Cars']]
In [15]: | sns.heatmap(df1.corr())
Out[15]: <AxesSubplot:>
                                                        - 1.0
                                                        - 0.8
                                                        - 0.6
           Sr. No.
                                                        - 0.4
                                                        - 0.2
                                                        - 0.0
           of Cars
                                      No. of Cars
                     Sr. No.
In [19]: x=df1[['Sr. No.']]
          y=df1['No. of Cars']
In [20]: from sklearn.model_selection import train_test_split
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
In [21]: from sklearn.linear model import LinearRegression
          lr=LinearRegression()
          lr.fit(x_train,y_train)
Out[21]: LinearRegression()
In [22]: |print(lr.intercept_)
          17.200633291477864
In [23]:
          coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
          coeff
Out[23]:
                  Co-efficient
                     -0.30398
           Sr. No.
```

```
D10-20 - Jupyter Notebook
In [24]:
         prediction=lr.predict(x_test)
         plt.scatter(y_test,prediction)
Out[24]: <matplotlib.collections.PathCollection at 0x1ec97a01370>
          16
          15
          14
          13
          12
          11
                         10
                              11
                                   12
                                        13
                                             14
                                                   15
                                                        16
In [25]: print(lr.score(x_test,y_test))
         0.3658816497886015
In [26]: from sklearn.linear model import Ridge,Lasso
In [27]: rr=Ridge(alpha=10)
         rr.fit(x_train,y_train)
Out[27]: Ridge(alpha=10)
In [28]: |rr.score(x_test,y_test)
Out[28]: 0.363491971767534
In [29]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
Out[29]: Lasso(alpha=10)
In [30]: la.score(x_test,y_test)
Out[30]: 0.14517901597397775
In [31]: | from sklearn.linear_model import ElasticNet
         en=ElasticNet()
         en.fit(x_train,y_train)
Out[31]: ElasticNet()
In [32]: print(en.coef_)
```

[-0.2925483]

```
In [33]: |print(en.intercept_)
         17.055197792088315
         print(en.predict(x_train))
In [34]:
         [14.71481141 11.20423183 11.49678013 10.32658694 16.76264949 12.66697332
          13.54461822 10.03403864 15.8850046 14.12971481 15.00735971 9.74149034
          13.83716651 12.08187672 11.49678013 15.5924563 16.1775529 15.299908 ]
In [35]: print(en.score(x_train,y_train))
         0.3667895915833099
In [36]: from sklearn import metrics
In [37]: print("Mean Absolytre Error:", metrics.mean_absolute_error(y_test, prediction))
         Mean Absolytre Error: 2.8942512420156135
In [38]: print("Mean Square Error:", metrics.mean_squared_error(y_test, prediction))
         Mean Square Error: 10.145893603382376
In [39]: print("Root Mean Square Error:",np.sqrt(metrics.mean_absolute_error(y_test,pre
         Root Mean Square Error: 1.7012499058091417
In [40]:
         import pickle
In [41]: |f5="prediction"
         pickle.dump(lr,open(f5,'wb'))
 In [ ]:
```