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

```
In [2]: #to import dataset
data1=pd.read_csv(r"C:\Users\user\Downloads\23_Vande Bharat - 23_Vande Bharat.csv")
data1
```

Out[2]:

	Sr. No.	Train Name	Train Number	Originating City	Originating Station	Terminal City	Ter
0	1	New Delhi - Varanasi Vande Bharat Express	22435/22436	Delhi	New Delhi	Varanasi	Vara
1	2	New Delhi - Shri Mata Vaishno Devi Katra Vande...	22439/22440	Delhi	New Delhi	Katra	Shri
2	3	Mumbai Central - Gandhinagar Capital Vande Bha...	20901/20902	Mumbai	Mumbai Central	Gandhinagar	Gandh
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	My
5	6	Bilaspur - Nagpur Vande Bharat Express	20825/20826	Bilaspur, Chhattisgarh	Bilaspur Junction	Nagpur	Na
6	7	Howrah - New Jalpaiguri Vande Bharat Express	22301/22302	Kolkata	Howrah Junction	Siliguri	N
7	8	Visakhapatnam - Secunderabad Vande Bharat Express	20833/20834	Visakhapatnam	Visakhapatnam Junction	Hyderabad	S
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	S
10	11	Rani Kamalapati (Habibganj) - Hazrat Nizamuddi...	20171/20172	Bhopal	Habibganj (Rani Kamalapati)	Delhi	Hazra
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	Coimbi
13	14	Delhi Cantonment - Ajmer Vande Bharat Express	20977/20978	Delhi	Delhi Cantonment	Ajmer	A

Sr. No.		Train Name	Train Number	Originating City	Originating Station	Terminal City	Ter
14	15	Kasaragod - Thiruvananthapuram Vande Bharat Ex...	20633/20634	Kasaragod	Kasaragod	Thiruvananthapuram	Thiruv
15	16	Howrah - Puri Vande Bharat Express	22895/22896	Kolkata	Howrah Junction	Puri	
16	17	Anand Vihar Terminal - Dehradun Vande Bharat E...	22457/22458	Delhi	Anand Vihar Terminal	Dehradun	Dehr
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	Mad
19	19	Mumbai CSMT - Madgaon Vande Bharat Express	22229/22230	Mumbai	Chhatrapati Shivaji Terminus	Madgaon	Mad
20	20	Patna - Ranchi Vande Bharat Express	22349/22350	Patna	Patna Junction	Ranchi	Ra
21	21	KSR Bengaluru - Dharwad Vande Bharat Express	20661/20662	Bangalore	Bangalore City	Hubballi - Dharwad	
22	22	Rani Kamalapati (Habibganj) - Jabalpur Vande B...	20173/20174	Bhopal	Habibganj (Rani Kamalapati)	Jabalpur	Jab
23	23	Indore - Bhopal Vande Bharat Express	20911/20912	Indore	Indore Junction	Bhopal	Bh
24	24	Jodhpur - Sabarmati (Ahmedabad) Vande Bharat E...	12461/12462	Jodhpur	Jodhpur Junction	Ahmedabad	Saba
25	25	Gorakhpur - Lucknow Charbagh Vande Bharat Express	22549/22550	Gorakhpur	Gorakhpur Junction	Charbagh	Luckr

```
In [3]: #to display top 5 rows
data=data1.head()
data
```

Out[3]:

	Sr. No.	Train Name	Train Number	Originating City	Originating Station	Terminal City	Terminal Station	Operator	No of Car
0	1	New Delhi - Varanasi Vande Bharat Express	22435/22436	Delhi	New Delhi	Varanasi	Varanasi Junction	NR	1
1	2	New Delhi - Shri Mata Vaishno Devi Katra Vande...	22439/22440	Delhi	New Delhi	Katra	Shri Mata Vaishno Devi Katra	NR	1
2	3	Mumbai Central - Gandhinagar Capital Vande Bha...	20901/20902	Mumbai	Mumbai Central	Gandhinagar	Gandhinagar Capital	WR	1
3	4	New Delhi - Amb Andaura Vande Bharat Express	22447/22448	Delhi	New Delhi	Andaura	Amb Andaura	NR	1
4	5	MGR Chennai Central - Mysuru Vande Bharat Express	20607/20608	Chennai	Chennai Central	Mysuru	Mysore Junction	SR	1



DATA CLEANING AND PREPROCESSING

In [4]: data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5 entries, 0 to 4
Data columns (total 16 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Sr. No.                5 non-null      int64
1   Train Name             5 non-null      object
2   Train Number           5 non-null      object
3   Originating City       5 non-null      object
4   Originating Station    5 non-null      object
5   Terminal City          5 non-null      object
6   Terminal Station       5 non-null      object
7   Operator               5 non-null      object
8   No. of Cars            5 non-null      int64
9   Frequency              5 non-null      object
10  Distance               5 non-null      object
11  Travel Time            5 non-null      object
12  Speed                  5 non-null      object
13  Average Speed          5 non-null      object
14  Inauguration           5 non-null      object
15  Average occupancy      5 non-null      object
dtypes: int64(2), object(14)
memory usage: 768.0+ bytes
```

In [5]: *#to display summary of statistics*
data.describe()

Out[5]:

	Sr. No.	No. of Cars
count	5.000000	5.0
mean	3.000000	16.0
std	1.581139	0.0
min	1.000000	16.0
25%	2.000000	16.0
50%	3.000000	16.0
75%	4.000000	16.0
max	5.000000	16.0

In [6]: `data.isnull()`

Out[6]:

	Sr. No.	Train Name	Train Number	Originating City	Originating Station	Terminal City	Terminal Station	Operator	No. of Cars	Frequency
0	False	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	False

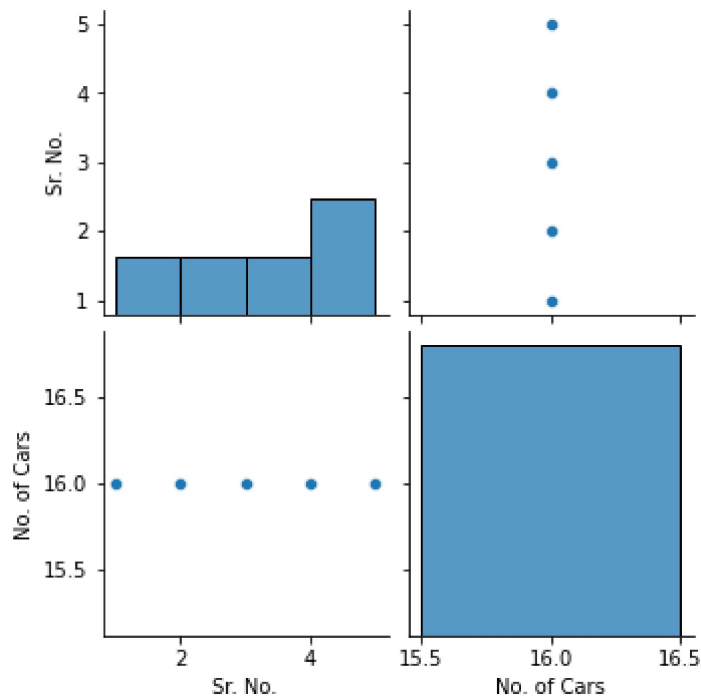
In [7]: *#to display the column heading*
`data.columns`

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

EDA and DATA VISUALIZATION

```
In [12]: sns.pairplot(data)
```

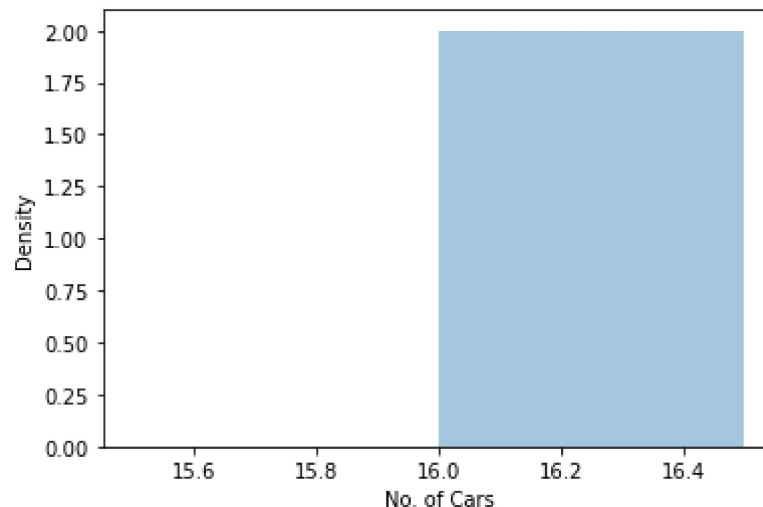
```
Out[12]: <seaborn.axisgrid.PairGrid at 0x2a12f0dedc0>
```



```
In [14]: sns.distplot(data['No. of Cars'])
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:306: UserWarning: Dataset has 0 variance; skipping density estimate.
warnings.warn(msg, UserWarning)

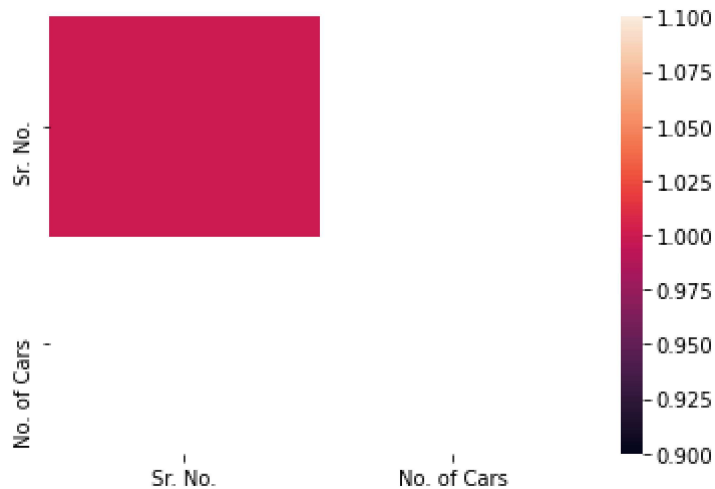
```
Out[14]: <AxesSubplot:xlabel='No. of Cars', ylabel='Density'>
```



```
In [8]: df=data[['Sr. No.', 'Train Name', 'Train Number', 'Originating City',  
                'Originating Station', 'Terminal City', 'Terminal Station', 'Operator',  
                'No. of Cars', 'Frequency', 'Distance', 'Travel Time', 'Speed',  
                'Average Speed', 'Inauguration', 'Average occupancy']]
```

```
In [15]: sns.heatmap(df.corr())
```

```
Out[15]: <AxesSubplot:>
```



TO TRAIN MODEL

```
In [20]: x=df[['Sr. No.']]
         y=df['No. of Cars']
```

```
In [21]: #to split my dataset into training and test

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 [22]: from sklearn.linear_model import LinearRegression

lr=LinearRegression()
lr.fit(x_train,y_train)
```

```
Out[22]: LinearRegression()
```

```
In [23]: #to find intercept
         print(lr.intercept_)
```

```
16.0
```

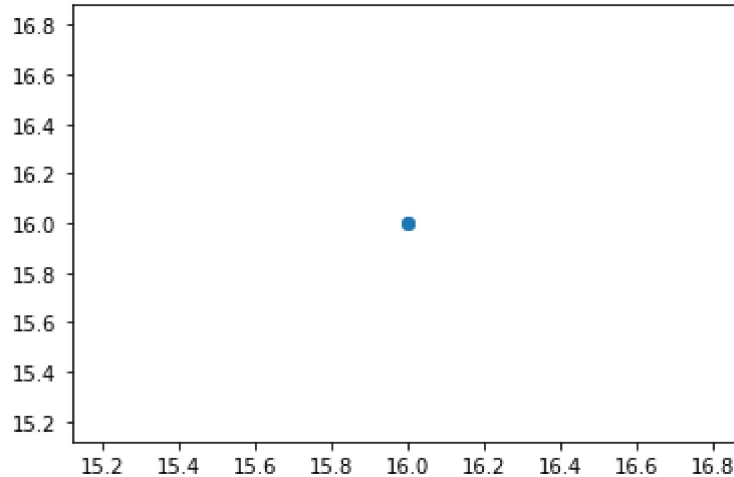
```
In [24]: coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
         coeff
```

```
Out[24]:
```

	Co-efficient
Sr. No.	0.0


```
In [25]: prediction = lr.predict(x_test)
plt.scatter(y_test, prediction)
```

Out[25]: <matplotlib.collections.PathCollection at 0x2a130271280>



```
In [26]: print(lr.score(x_test, y_test))
```

1.0

RIDGE AND LASSO REGRESSION

```
In [27]: from sklearn.linear_model import Ridge, Lasso
```

```
In [28]: rr=Ridge(alpha=10)
rr.fit(x_train, y_train)
```

Out[28]: Ridge(alpha=10)

```
In [29]: rr.score(x_test, y_test)
```

Out[29]: 1.0

```
In [30]: la=Lasso(alpha=10)
la.fit(x_train, y_train)
```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear_model_coordinate_descent.py:530: ConvergenceWarning: Objective did not converge. You might want to increase the number of iterations. Duality gap: 0.0, tolerance: 0.0
model = cd_fast.enet_coordinate_descent(

Out[30]: Lasso(alpha=10)

```
In [31]: la.score(x_test, y_test)
```

Out[31]: 1.0

```
In [32]: from sklearn.linear_model import ElasticNet
en=ElasticNet()
en.fit(x_train,y_train)
```

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\linear_model_coordinate_descent.py:530: ConvergenceWarning: Objective did not converge. You might want to increase the number of iterations. Duality gap: 0.0, tolerance: 0.0
model = cd_fast.enet_coordinate_descent(

Out[32]: ElasticNet()

```
In [33]: print(en.coef_)

[0.]
```

```
In [34]: print(en.predict(x_test))

[16. 16.]
```

```
In [35]: print(en.score(x_test,y_test))

1.0
```

EVALUATION METRICS

```
In [36]: from sklearn import metrics
```

```
In [37]: print("Mean Absolute error",metrics.mean_absolute_error(y_test,prediction))

Mean Absolute error 0.0
```

```
In [38]: print("Mean Squared error",metrics.mean_squared_error(y_test,prediction))

Mean Squared error 0.0
```

```
In [39]: print("Root Mean Absolute error",np.sqrt(metrics.mean_squared_error(y_test,prediction)))

Root Mean Absolute error 0.0
```

MODEL SAVING

```
In [40]: import pickle
```

```
In [42]: filename='predict4'
pickle.dump(lr,open(filename,'wb'))
```

