```
import pandas as pd
import numpy as np
import seaborn as sns
from matplotlib import pyplot as plt
import warnings
warnings.filterwarnings("ignore")
```

In [2]: data= pd.read_csv('IMDB-Movie-Data.csv')
 data.head()

Out[2]:		Rank	Title	Genre	Description	Director	Actors	Year	F (IV
	Guardians 1 of the Galaxy		of the	Action,Adventure,Sci-Fi	A group of intergalactic criminals are forced	James Gunn	Chris Pratt, Vin Diesel, Bradley Cooper, Zoe S	2014	
	1 2 Prometheus		Adventure, Mystery, Sci-Fi	Following clues to the origin of mankind, a te	Ridley Scott	Noomi Rapace, Logan Marshall- Green, Michael Fa	2012		
	2 3 Split		Horror,Thriller	Three girls are kidnapped by a man with a diag	M. Night Shyamalan	James McAvoy, Anya Taylor-Joy, Haley Lu Richar	2016		
	3	4	Sing	Animation,Comedy,Family	In a city of humanoid animals, a hustling thea	Christophe Lourdelet	Matthew McConaughey,Reese Witherspoon, Seth Ma	2016	
	4	5	Suicide Squad	Action, Adventure, Fantasy	A secret government agency recruits some of th	David Ayer	Will Smith, Jared Leto, Margot Robbie, Viola D	2016	

In [3]: data.tail()

Out[3]: **Runtime** Actors Year Rank Title Description Director Genre (Minutes) A tight-knit Secret Chiwetel Ejiofor, team of in Nicole Kidman, 2015 995 996 Crime, Drama, Mystery rising Billy Ray 111 Their Julia Roberts... investigators, Eyes alo...

	Rank	Title	Genre	Description	Director	Actors	Year	Runtime (Minutes)
996	997	Hostel: Part II	Horror	Three American college students studying abroa	Eli Roth	Lauren German, Heather Matarazzo, Bijou Philli	2007	94
997	998	Step Up 2: The Streets	Drama,Music,Romance	Romantic sparks occur between two dance studen	Jon M. Chu	Robert Hoffman, Briana Evigan, Cassie Ventura,	2008	98
998	999	Search Party	Adventure,Comedy	A pair of friends embark on a mission to reuni	Scot Armstrong	Adam Pally, T.J. Miller, Thomas Middleditch,Sh	2014	93
999	1000	Nine Lives	Comedy, Family, Fantasy	A stuffy businessman finds himself trapped ins	Barry Sonnenfeld	Kevin Spacey, Jennifer Garner, Robbie Amell,Ch	2016	87

In [4]: data.shape

Out[4]: (1000, 12)

In [5]:

data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	Rank	1000 non-null	int64
1	Title	1000 non-null	object
2	Genre	1000 non-null	object
3	Description	1000 non-null	object
4	Director	1000 non-null	object
5	Actors	1000 non-null	object
6	Year	1000 non-null	int64
7	Runtime (Minutes)	1000 non-null	int64
8	Rating	1000 non-null	float64
9	Votes	1000 non-null	int64
10	Revenue (Millions)	872 non-null	float64
11	Metascore	936 non-null	float64
dtyp	es: float64(3), int6	4(4), object(5)	

In [6]:

data.describe()

memory usage: 93.9+ KB

Out[6]:

	Rank	Year	Runtime (Minutes)	Rating	Votes	Revenue (Millions)	Metascore
cou	int 1000.000000	1000.000000	1000.000000	1000.000000	1.000000e+03	872.000000	936.000000
me	an 500.500000	2012.783000	113.172000	6.723200	1.698083e+05	82.956376	58.985043

	Rank	Year	Runtime (Minutes)	Rating	Votes	Revenue (Millions)	Metascore
std	288.819436	3.205962	18.810908	0.945429	1.887626e+05	103.253540	17.194757
min	1.000000	2006.000000	66.000000	1.900000	6.100000e+01	0.000000	11.000000
25%	250.750000	2010.000000	100.000000	6.200000	3.630900e+04	13.270000	47.000000
50%	500.500000	2014.000000	111.000000	6.800000	1.107990e+05	47.985000	59.500000
75%	750.250000	2016.000000	123.000000	7.400000	2.399098e+05	113.715000	72.000000
max	1000.000000	2016.000000	191.000000	9.000000	1.791916e+06	936.630000	100.000000

In [7]: data.isnull()

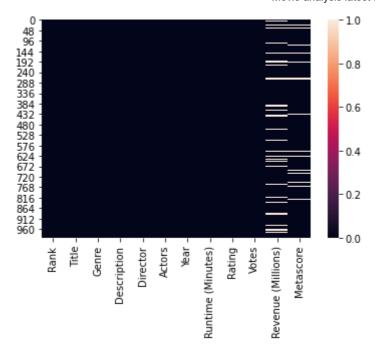
Out[7]:

	Rank	Title	Genre	Description	Director	Actors	Year	Runtime (Minutes)	Rating	Votes	Revenue (Millions)
0	False	False	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	False	False
•••											
995	False	False	False	False	False	False	False	False	False	False	True
996	False	False	False	False	False	False	False	False	False	False	False
997	False	False	False	False	False	False	False	False	False	False	False
998	False	False	False	False	False	False	False	False	False	False	True
999	False	False	False	False	False	False	False	False	False	False	False

1000 rows × 12 columns

In [8]:

```
sns.heatmap(data.isnull())
plt.show()
```



```
In [9]: data.isnull().sum()
```

0 Rank Out[9]: Title 0 Genre 0 Description 0 Director 0 Actors 0 0 Year 0 Runtime (Minutes) 0 Rating Votes 0 Revenue (Millions) 128 Metascore dtype: int64

Out[10]:		Rank	Title	Genre	Description	Director	Actors	Year	F (IV
	0	1	Guardians of the Galaxy	Action,Adventure,Sci-Fi	A group of intergalactic criminals are forced	James Gunn	Chris Pratt, Vin Diesel, Bradley Cooper, Zoe S	2014	
	1	2	Prometheus	Adventure,Mystery,Sci-Fi	Following clues to the origin of mankind, a te	Ridley Scott	Noomi Rapace, Logan Marshall- Green, Michael Fa	2012	
	2	3	Split	Horror,Thriller	Three girls are kidnapped by a man with a diag	M. Night Shyamalan	James McAvoy, Anya Taylor-Joy, Haley Lu Richar	2016	

	Rank	Title	Genre	e Description Director		Actors Yea		F (N
	3 4 Sing		Animation,Comedy,Family	In a city of humanoid animals, a hustling thea	Christophe Lourdelet	Matthew McConaughey,Reese Witherspoon, Seth Ma	2016	
	4 5	Suicide Squad	Action, Adventure, Fantasy	A secret government agency recruits some of th	David Ayer	Will Smith, Jared Leto, Margot Robbie, Viola D	2016	
In [11]:	= True)	ascore']	llions)'].fillna(df .fillna(df['Metasco				place	
Out[11]:	Rank Title Genre Description Director Actors Year Runtime (Marting Votes Revenue (Matascore dtype: int	Minutes) Millions)	0 0 0 0 0 0 0 0 0					
In [12]:	dup_data	a=data.du	uplicated().any()					
In [13]:	. ,		e any duplicate val	ues?",dup	_data)			
In [14]:	df.colum	nns						
Out[14]:	'Ru 'Me	ank', 'Tii untime (Mi etascore'] pe='object		iption', 'D Votes', 'Re	irector', venue (Mil	'Actors', 'Year', lions)',		

Movies runtime more than 3hours

82 The Wolf of Wall Street

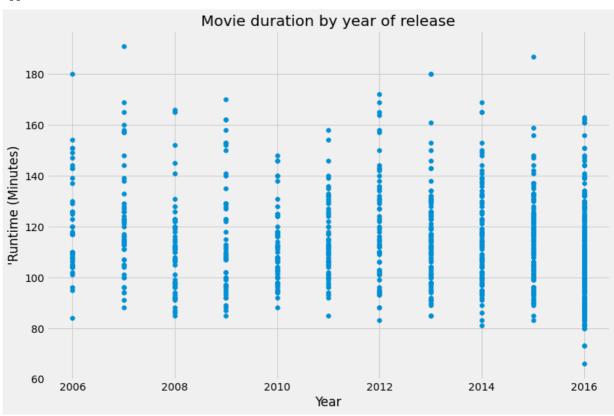
Title

```
88
                  The Hateful Eight
                    La vie d'Adèle
         311
         828
                      Grindhouse
         965
                    Inland Empire
In [16]:
          data.columns
         Out[16]:
                'Metascore'],
               dtype='object')
In [17]:
          #left
In [18]:
          data.nlargest(10, 'Runtime (Minutes)')[['Title', 'Runtime (Minutes)']]
Out[18]:
                                       Title
                                            Runtime (Minutes)
         828
                                  Grindhouse
                                                        191
          88
                              The Hateful Eight
                                                        187
                          The Wolf of Wall Street
          82
                                                        180
         311
                                 La vie d'Adèle
                                                        180
         965
                                 Inland Empire
                                                        180
                                   Cloud Atlas
                                                        172
         267
                                      3 Idiots
         430
                                                        170
          36
                                   Interstellar
                                                        169
             Pirates of the Caribbean: At World's End
                                                        169
         271
                The Hobbit: An Unexpected Journey
                                                        169
In [19]:
          plt.style.use('fivethirtyeight')
          fig=plt.figure(figsize=(12,8))
          #Create a scatter plot of duration versus release year
          plt.scatter(
          x=df['Year'],
          y=df['Runtime (Minutes)']
          #Create a title and axis labels
```

```
plt.title("Movie duration by year of release")
plt.xlabel("Year")
plt.ylabel("'Runtime (Minutes)")

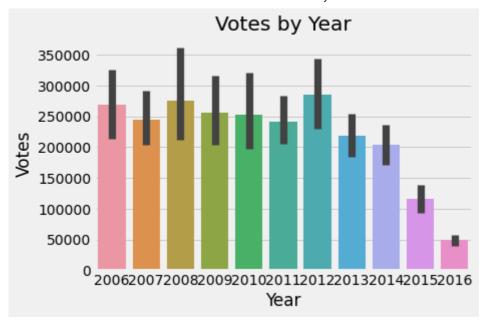
#Show the plot
plt.plot()
```

Out[19]: []



Mean of votes by years.

```
In [20]:
          data.groupby('Year')["Votes"].mean().sort_values(ascending=False)
         Year
Out[20]:
         2012
                 285226.093750
                 275505.384615
         2008
                 269289.954545
         2006
         2009
                 255780.647059
         2010
                 252782.316667
         2007
                 244331.037736
                 240790.301587
         2011
         2013
                 219049.648352
         2014
                 203930.224490
         2015
                 115726.220472
         2016
                  48591.754209
         Name: Votes, dtype: float64
In [21]:
          sns.barplot("Year",y="Votes",data=data)
          plt.title("Votes by Year")
          plt.show()
```



```
In [22]: df.drop(['Genre'],axis=1,inplace=True )
    df.drop(['Description'],axis=1,inplace=True )
    df.drop(['Director'],axis=1,inplace=True )
    df.drop(['Actors'],axis=1,inplace=True )
    df.head()
```

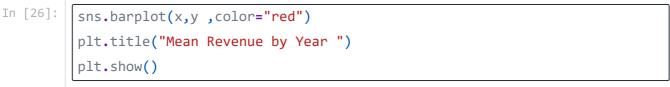
Out[22]:	l	Rank	Title	Year	Runtime (Minutes)	Rating	Votes	Revenue (Millions)	Metascore
	0	1	Guardians of the Galaxy	2014	121	8.1	757074	333.13	76.0
	1	2	Prometheus	2012	124	7.0	485820	126.46	65.0
	2	3	Split	2016	117	7.3	157606	138.12	62.0
	3	4	Sing	2016	108	7.2	60545	270.32	59.0
	4	5	Suicide Squad	2016	123	6.2	393727	325.02	40.0

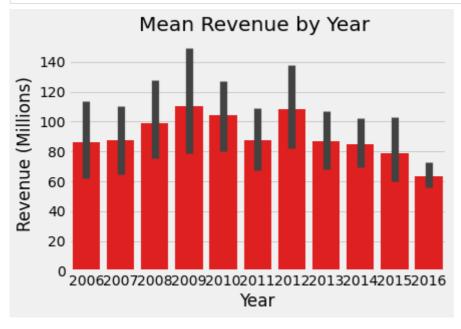
In [23]: df.head()

Out[23]:		Rank	Title	Year	Runtime (Minutes)	Rating	Votes	Revenue (Millions)	Metascore
	0	1	Guardians of the Galaxy	2014	121	8.1	757074	333.13	76.0
	1	2	Prometheus	2012	124	7.0	485820	126.46	65.0
	2	3	Split	2016	117	7.3	157606	138.12	62.0
	3	4	Sing	2016	108	7.2	60545	270.32	59.0
	4	5	Suicide Squad	2016	123	6.2	393727	325.02	40.0

Mean Revenue by years

```
5/30/22, 9:08 PM
                                                   Movie analysis latest saved
    In [24]:
               df.groupby("Year")["Revenue (Millions)"].mean()
    Out[24]:
              2006
                       86.144835
              2007
                       87.510481
              2008
                       98.772623
              2009
                      110.276186
              2010
                      103.975319
              2011
                       87.538355
              2012
                      107.973281
              2013
                       86.984496
              2014
                       84.992097
              2015
                       78.862278
              2016
                       63.446588
              Name: Revenue (Millions), dtype: float64
    In [25]:
               x=df["Year"]
               y=df["Revenue (Millions)"]
    In [26]:
               sns.barplot(x,y ,color="red")
```





Top 10 Movies by Revenue

```
In [27]:
         b=df.groupby("Title")["Revenue
         (Millions)"].max().sort_values(ascending=False).head(10)
         pd.DataFrame(b)
```

Out[27]: **Revenue (Millions)**

Title	
Star Wars: Episode VII - The Force Awakens	936.63
Avatar	760.51

Revenue (Millions)

Title	
Jurassic World	652.18
The Avengers	623.28
The Dark Knight	533.32
Rogue One	532.17
Finding Dory	486.29
Avengers: Age of Ultron	458.99
The Dark Knight Rises	448.13
The Hunger Games: Catching Fire	424.65

Top 10 Movies by Votes.

```
In [28]: b=df.groupby("Title")
    ["Votes"].max().sort_values(ascending=False).head(10)
    pd.DataFrame(b)
```

```
Out[28]: Votes
```

```
      Title

      The Dark Knight
      1791916

      Inception
      1583625

      The Dark Knight Rises
      1222645

      Interstellar
      1047747

      The Avengers
      1045588

      Django Unchained
      1039115

      Inglourious Basterds
      959065

      The Departed
      937414

      Avatar
      935408

      The Prestige
      913152
```

Top 10 by Rating.

localhost:8888/nbconvert/html/Downloads/Movie analysis latest saved.ipynb?download=false

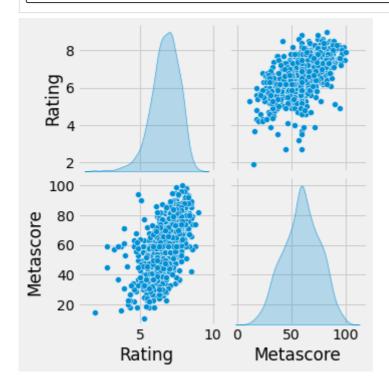
Title

Rating

Title	
The Dark Knight	9.0
Dangal	8.8
Inception	8.8
Interstellar	8.6
The Intouchables	8.6
Kimi no na wa	8.6
Taare Zameen Par	8.5
The Lives of Others	8.5
The Prestige	8.5
The Departed	8.5

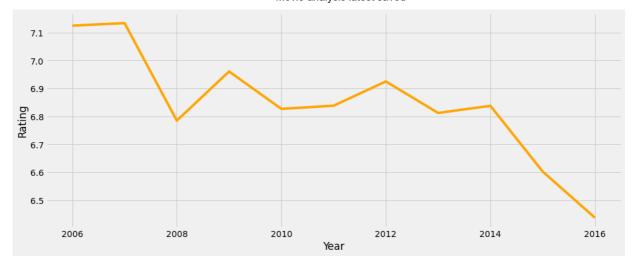
Pairplot for Rating & Metascore.

```
In [30]: sns.pairplot(df[["Rating","Metascore"]],diag_kind="kde")
plt.show()
```

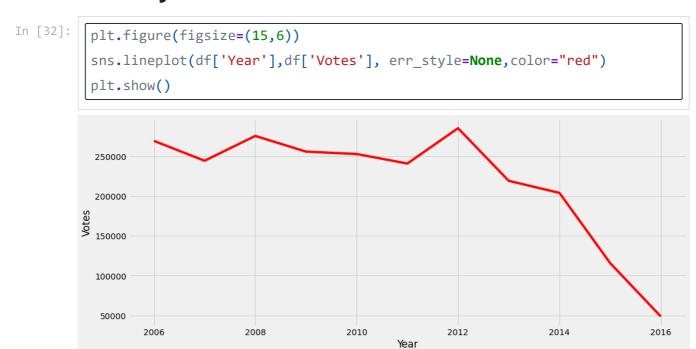


Year by rating.

```
plt.figure(figsize=(15,6))
sns.lineplot(df['Year'],df['Rating'], err_style=None,color="orange")
plt.show()
```

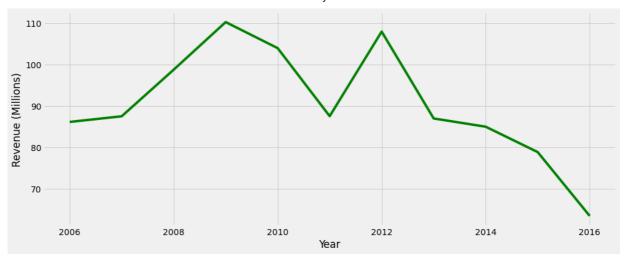


Year by Votes.

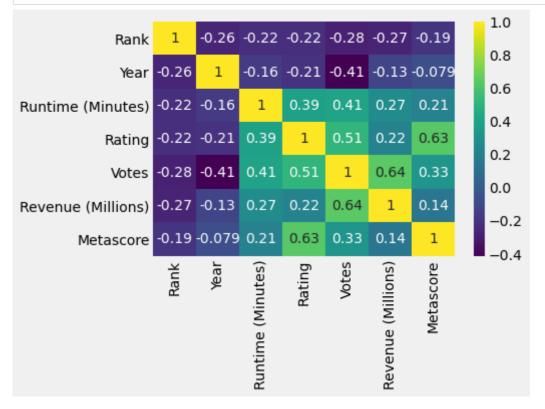


Year by Revenue.

```
plt.figure(figsize=(15,6))
sns.lineplot(df['Year'],df['Revenue (Millions)'],
err_style=None,color="green")
plt.show()
```



```
In [34]: sns.heatmap(data.corr(), annot = True, cmap = 'viridis')
plt.show()
```



```
In [35]: from sklearn import linear_model
```

```
In [36]: from sklearn.linear_model import LinearRegression
```

```
In [38]: [1r=LinearRegression()
```

```
In [39]: df.columns
```

```
In [99]: x=df[["Votes"]]
```

```
y=df[["Revenue (Millions)"]]
In [100...
          from sklearn.model_selection import train_test_split
In [101...
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_s
In [102...
          x_train.head()
Out[102...
               Votes
         105 115546
                 291
          68
         479
               41642
             113686
         399
                 223
         434
In [103...
          lr.fit(x_train,y_train)
Out[103... LinearRegression()
In [104...
          y_pred=lr.predict(x_test)
In [105...
          lr.score(x_train,y_train)
Out[105... 0.33769779991188953
In [106...
          print("Coefficient: \n", lr.coef_)
         Coefficient:
          [[0.00028465]]
In [107...
          print("Intercept :\n", lr.intercept_)
         Intercept :
          [33.61382523]
In [108...
          lr.score(y_test, y_pred)
Out[108... -0.7440714625299043
In [109...
          from sklearn.metrics import mean absolute error
          from sklearn.metrics import mean_squared_error
          from sklearn.metrics import r2_score
In [110...
```

```
print("R2Score : " ,r2_score(y_test, y_pred))
          print("mean_absolute_error : ",mean_absolute_error(y_test, y_pred))
          print("mean_squared_error : " ,mean_squared_error(y_test, y_pred))
          print("Root mean_squared_error : ",np.sqrt(mean_squared_error(y_test,
          y_pred)))
         R2Score: 0.4149415654872476
         mean_absolute_error : 52.740548367277746
         mean_squared_error : 6819.744965698331
         Root mean_squared_error : 82.58174717029382
In [111...
          a={"Actual value":x_train,"predicted value":y_pred}
Out[111... {'Actual value':
                                Votes
          105
              115546
          68
                  291
          479
                41642
          399
               113686
          434
                  223
          . .
          835
                38804
          192
               268282
          629
                74886
          559
               115355
          684 245144
          [700 rows x 1 columns],
           predicted value': array([[ 73.72104506],
                  [ 92.47750379],
                   46.11822594],
                  [ 36.22890705],
                 [133.81725919],
                  [ 88.6910862 ],
                   33.76269731],
                   34.5545943 ],
                   49.89781192],
                   93.0009756],
                  [112.85447638],
                   59.33823737],
                   42.54330348],
                  [ 61.18277097],
                  [183.23908905],
                   47.61577089],
                   37.13238694],
                   74.33845144],
                  [ 36.06523316],
                  [306.61191452],
                  [ 69.48829588],
                  [ 60.31373376],
                  [101.13884151],
                  [ 62.97549823],
                   83.11991157],
                  [ 53.29938237],
                  [146.05778911],
                  [ 35.66444561],
                  [ 59.58218263],
                  [171.90260831],
                  [ 35.28045243],
                  [129.18713827],
                  [ 90.53618911],
                   47.9149383 ],
                   91.22134225],
```

[34.79654701],

[88.32075623], [88.92962311], [90.36938406], [118.6249062], [175.61985588], [54.43656011], [121.6840424], [132.54885769], [48.11561672], [33.90729964], [33.77436797], [43.34430928], [54.41976575], [74.29034555], [134.1821808], [39.4531404], [39.79500535], [174.46417588], [42.81087472], [59.1996127], [156.59582591], [189.42738542], [43.20312276], [119.23747353], [53.1562033], [34.83895989], [103.74026012], [56.34457072], [43.69528303], [34.69521152], [50.49700069], [37.24539308], [73.41675394], [46.25941246], [38.57784089], [147.66805556], [42.05142786], [37.80046107], [59.69433483], [105.37956089], [96.42674132], [58.6308815], [45.10145526], [33.75102665], [199.14306766], [102.41635182], [62.38968802], [60.70455855], [86.84456005], [59.28586172], [45.41143937], [55.81512126], [118.69948456], [62.44917992], [69.66648693], [60.57219619], [50.1010522], [201.84866826], [47.96076699], [106.04194202], [41.16303444], [33.89961408], [33.6952352], [34.62831871], [132.14664689], [65.94980866], [158.91942588], [49.33875883], [33.97333849],

```
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[207.69538433],
[ 77.634132 ],
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[177.08865116],
[ 47.07465077],
[ 95.86085663],
[ 87.93448585],
[ 48.75237933],
[ 60.1779556 ],
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[ 35.74158583],
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 84.35073924],
 43.12626719],
[ 35.24060139],
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```

```
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train_std = sc.fit_transform(x_train) # study & exam
x_test_std = sc.transform(x_test)
```

```
In [114... | from sklearn.ensemble import RandomForestRegressor
```

```
rf_tree = RandomForestRegressor(random_state=0)
rf_tree.fit(x_train_std,y_train)
rf_tree_y_pred = rf_tree.predict(x_train_std)
print("Accuracy: {}".format(rf_tree.score(x_train_std,y_train)))
# or
print("R squared:
{}".format(r2_score(y_true=y_train,y_pred=rf_tree_y_pred)))
```

Accuracy: 0.8620132614862872 R squared: 0.8620132614862872