```
In [ ]:
```

# In [82]:

```
# IMPORT LIBRARIES
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

#### In [83]:

a=pd.read\_csv(r"C:\Users\user\Downloads\15\_Horse Racing Results.CSV - 15\_Horse Racing Results.CSV.csv
a

# Out[83]:

	Dato	Track	Race Number	Distance	Surface	Prize money	Starting position	Jockey	Jockey weight	Country	 Tra
0	03.09.2017	Sha Tin	10	1400	Gress	1310000	6	K C Leung	52	Sverige	 
1	16.09.2017	Sha Tin	10	1400	Gress	1310000	14	C Y Ho	52	Sverige	
2	14.10.2017	Sha Tin	10	1400	Gress	1310000	8	C Y Ho	52	Sverige	
3	11.11.2017	Sha Tin	9	1600	Gress	1310000	13	Brett Prebble	54	Sverige	
4	26.11.2017	Sha Tin	9	1600	Gress	1310000	9	C Y Ho	52	Sverige	
27003	14.06.2020	Sha Tin	11	1200	Gress	1450000	6	A Hamelin	59	Australia	
27004	21.06.2020	Sha Tin	2	1200	Gress	967000	7	K C Leung	57	Australia	
27005	21.06.2020	Sha Tin	4	1200	Gress	967000	6	Blake Shinn	57	Australia	 Р
27006	21.06.2020	Sha Tin	5	1200	Gress	967000	14	Joao Moreira	57	New Zealand	
27007	21.06.2020	Sha Tin	11	1200	Gress	1450000	7	C Schofield	55	New Zealand	

27008 rows × 21 columns

# In [84]:

a=a.head(10)

Out[84]:

	Dato	Track	Race Number	Distance	Surface	Prize money	Starting position	Jockey	Jockey weight	Country	 Trainerl
0	03.09.2017	Sha Tin	10	1400	Gress	1310000	6	K C Leung	52	Sverige	 С
1	16.09.2017	Sha Tin	10	1400	Gress	1310000	14	C Y Ho	52	Sverige	 С
2	14.10.2017	Sha Tin	10	1400	Gress	1310000	8	C Y Ho	52	Sverige	 С
3	11.11.2017	Sha Tin	9	1600	Gress	1310000	13	Brett Prebble	54	Sverige	 С
4	26.11.2017	Sha Tin	9	1600	Gress	1310000	9	C Y Ho	52	Sverige	 С
5	10.12.2017	Sha Tin	1	1800	Gress	1310000	4	C Y Ho	52	Sverige	 С
6	01.01.2018	Sha Tin	9	1800	Gress	1310000	9	C Schofield	54	Sverige	 С
7	04.02.2018	Sha Tin	5	1800	Gress	1310000	6	Joao Moreira	57	Sverige	 С
8	03.03.2018	Sha Tin	8	1800	Gress	1310000	3	C Y Ho	56	Sverige	 С
9	11.03.2018	Sha Tin	10	1600	Gress	1310000	8	C Y Ho	57	Sverige	 С

10 rows × 21 columns

1

#### In [85]:

# # to find a.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10 entries, 0 to 9
Data columns (total 21 columns):

#	Column	Non-Null Count	Dtype
0	Dato	10 non-null	object
1	Track	10 non-null	object
2	Race Number	10 non-null	int64
3	Distance	10 non-null	int64
4	Surface	10 non-null	object
5	Prize money	10 non-null	int64
6	Starting position	10 non-null	int64
7	Jockey	10 non-null	object
8	Jockey weight	10 non-null	int64
9	Country	10 non-null	object
10	Horse age	10 non-null	int64
11	TrainerName	10 non-null	object
12	Race time	10 non-null	object
13	Path	10 non-null	int64
14	Final place	10 non-null	int64
15	FGrating	10 non-null	int64
16	Odds	10 non-null	object
17	RaceType	10 non-null	object
18	HorseId	10 non-null	int64
19	JockeyId	10 non-null	int64
20	TrainerID	10 non-null	int64

dtypes: int64(12), object(9)

memory usage: 1.8+ KB

# In [86]:

# to display summary of statastic
a.describe()

# Out[86]:

	Race Number	Distance	Prize money	Starting position	Jockey weight	Horse age	Path	Final place	FGratin
count	10.000000	10.000000	10.0	10.000000	10.000000	10.000000	10.000000	10.000000	10.00000
mean	8.100000	1620.000000	1310000.0	8.000000	53.800000	7.400000	1.500000	4.700000	120.10000
std	2.923088	175.119007	0.0	3.527668	2.149935	0.516398	1.581139	2.496664	6.47130
min	1.000000	1400.000000	1310000.0	3.000000	52.000000	7.000000	0.000000	1.000000	107.00000
25%	8.250000	1450.000000	1310000.0	6.000000	52.000000	7.000000	0.250000	3.000000	119.00000
50%	9.000000	1600.000000	1310000.0	8.000000	53.000000	7.000000	1.000000	4.000000	123.00000
75%	10.000000	1800.000000	1310000.0	9.000000	55.500000	8.000000	2.000000	6.000000	124.00000
max	10.000000	1800.000000	1310000.0	14.000000	57.000000	8.000000	5.000000	9.000000	125.00000

**←** 

#### In [87]:

```
# to display colum heading
a.columns
```

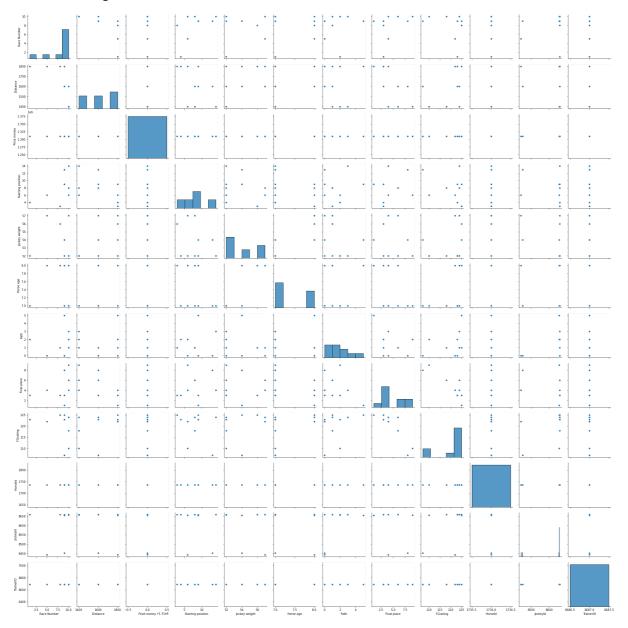
#### Out[87]:

# In [88]:

```
sns.pairplot(a)
```

#### Out[88]:

<seaborn.axisgrid.PairGrid at 0x198ba310e20>

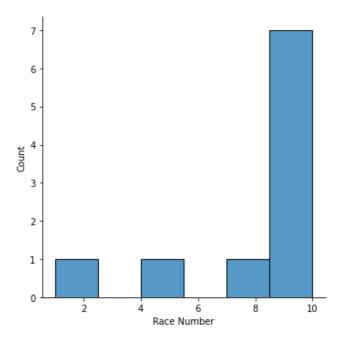


# In [89]:

```
sns.displot(a["Race Number"])
```

# Out[89]:

<seaborn.axisgrid.FacetGrid at 0x198bf882460>



# In [90]:

# Out[90]:

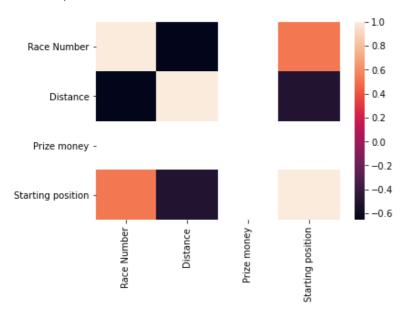
	Track	Race Number	Distance	Surface	Prize money	Starting position	Jockey
0	Sha Tin	10	1400	Gress	1310000	6	K C Leung
1	Sha Tin	10	1400	Gress	1310000	14	C Y Ho
2	Sha Tin	10	1400	Gress	1310000	8	C Y Ho
3	Sha Tin	9	1600	Gress	1310000	13	Brett Prebble
4	Sha Tin	9	1600	Gress	1310000	9	C Y Ho
5	Sha Tin	1	1800	Gress	1310000	4	C Y Ho
6	Sha Tin	9	1800	Gress	1310000	9	C Schofield
7	Sha Tin	5	1800	Gress	1310000	6	Joao Moreira
8	Sha Tin	8	1800	Gress	1310000	3	C Y Ho
9	Sha Tin	10	1600	Gress	1310000	8	C Y Ho

#### In [91]:

```
sns.heatmap(b.corr())
```

# Out[91]:

#### <AxesSubplot:>



# In [99]:

#### In [100]:

```
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 [101]:

```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

#### Out[101]:

LinearRegression()

#### In [102]:

```
lr.intercept_
```

#### Out[102]:

-1.5987211554602254e-14

```
In [103]:
```

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

#### Out[103]:

# Race Number 1.000000e+00 Distance 1.012346e-17 Prize money 0.000000e+00

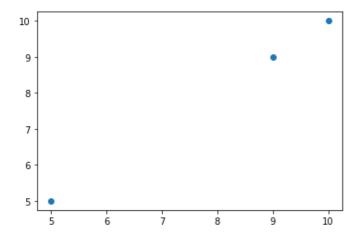
Starting position -2.192399e-16

# In [104]:

```
prediction = lr.predict(x_test)
plt.scatter(y_test,prediction)
```

#### Out[104]:

<matplotlib.collections.PathCollection at 0x198c2934dc0>



# In [105]:

```
lr.score(x_test,y_test)
```

# Out[105]:

1.0

#### In [106]:

```
lr.score(x_train,y_train)
```

### Out[106]:

1.0

#### In [107]:

```
from sklearn.linear_model import Ridge,Lasso
```

```
In [108]:
rr=Ridge(alpha=10)
rr.fit(x_test,y_test)
Out[108]:
Ridge(alpha=10)
In [109]:
rr.score(x_test,y_test)
Out[109]:
0.8871594645918177
In [110]:
la=Lasso(alpha=10)
la.fit(x_test,y_test)
Out[110]:
Lasso(alpha=10)
In [111]:
la.score(x_test,y_test)
Out[111]:
0.42616071428571434
In [112]:
from sklearn.linear_model import ElasticNet
en=ElasticNet()
en.fit(x_train,y_train)
Out[112]:
ElasticNet()
In [113]:
en.coef_
Out[113]:
array([ 0.78494747, -0.00282822, 0.
                                             , 0.
                                                          ])
In [114]:
en.intercept_
Out[114]:
6.195484221835903
In [115]:
prediction=en.predict(x_test)
prediction
Out[115]:
array([9.51981051, 5.02942959, 8.16921949])
```

```
In [116]:
en.score(x_test,y_test)
Out[116]:
0.9341682638724647
EVALUATION METRICS
In [117]:
from sklearn import metrics
In [118]:
print("Mean Absolute Error:",metrics.mean_absolute_error(y_test,prediction))
Mean Absolute Error: 0.4467998631727319
In [119]:
print("Mean Squared Error", metrics.mean_squared_error(y_test, prediction))
Mean Squared Error 0.30721476859516517
In [120]:
print("Root Mean Squared Error",np.sqrt(metrics.mean_squared_error(y_test,prediction)))
Root Mean Squared Error 0.5542695811562864
```

In [ ]: