Importing Reqired Libraries

```
import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
from sklearn.metrics import roc_curve
from sklearn.metrics import roc_auc_score
import warnings
warnings.filterwarnings('ignore')
```

Import Data

```
In [6]:
    election = pd.read_csv("election_data.csv")
    election
```

Out[6]:		Election-id	Result	Year	Amount Spent	Popularity Rank
	0	NaN	NaN	NaN	NaN	NaN
	1	122.0	0.0	32.0	3.81	3.0
	2	315.0	1.0	48.0	6.32	2.0
	3	201.0	1.0	51.0	3.67	1.0
	4	965.0	0.0	40.0	2.93	4.0
	5	410.0	1.0	52.0	3.60	1.0
	6	150.0	0.0	35.0	4.20	4.0
	7	743.0	1.0	39.0	5.66	2.0
	8	612.0	1.0	42.0	4.32	3.0
	9	206.0	1.0	44.0	3.26	3.0
	10	792.0	0.0	50.0	4.52	4.0

Initial Analysis

```
float64
          Popularity Rank
          dtype: object
 In [9]:
           election.isna().sum()
          Election-id
                               1
 Out[9]:
          Result
                               1
          Year
                               1
          Amount Spent
                               1
          Popularity Rank
                                1
          dtype: int64
In [10]:
           election.describe()
                                            Year
Out[10]:
                 Election-id
                                Result
                                                  Amount Spent Popularity Rank
           count
                   10.000000
                             10.000000
                                       10.000000
                                                      10.000000
                                                                      10.000000
                 451.600000
                              0.600000
                                       43.300000
                                                       4.229000
                                                                       2.700000
           mean
                 304.266404
                              0.516398
                                        6.912951
                                                       1.055351
                                                                       1.159502
             std
                 122.000000
                              0.000000
                                       32.000000
                                                       2.930000
                                                                       1.000000
            min
            25%
                 202.250000
                              0.000000
                                       39.250000
                                                       3.617500
                                                                       2.000000
            50%
                 362.500000
                                      43.000000
                                                       4.005000
                                                                       3.000000
                              1.000000
                                                                       3.750000
            75%
                 710.250000
                              1.000000
                                       49.500000
                                                       4.470000
            max 965.000000
                              1.000000 52.000000
                                                       6.320000
                                                                       4.000000
In [11]:
           election.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 11 entries, 0 to 10
          Data columns (total 5 columns):
           #
                Column
                                  Non-Null Count Dtype
           0
                Election-id
                                   10 non-null
                                                     float64
           1
                Result
                                   10 non-null
                                                     float64
           2
                Year
                                   10 non-null
                                                     float64
           3
                Amount Spent
                                   10 non-null
                                                     float64
                Popularity Rank 10 non-null
                                                     float64
          dtypes: float64(5)
          memory usage: 568.0 bytes
In [12]:
           election.dropna(axis = 0,inplace = True)
In [15]:
           election.head(10)
Out[15]:
               Election-id Result Year
                                      Amount Spent Popularity Rank
            1
                    122.0
                             0.0
                                 32.0
                                                3.81
                                                                 3.0
                             1.0
            2
                    315.0
                                 48.0
                                                6.32
                                                                 2.0
            3
                    201.0
                             1.0
                                 51.0
                                                3.67
                                                                 1.0
```

0.0

1.0

40.0

52.0

2.93

3.60

4.0

1.0

965.0

410.0

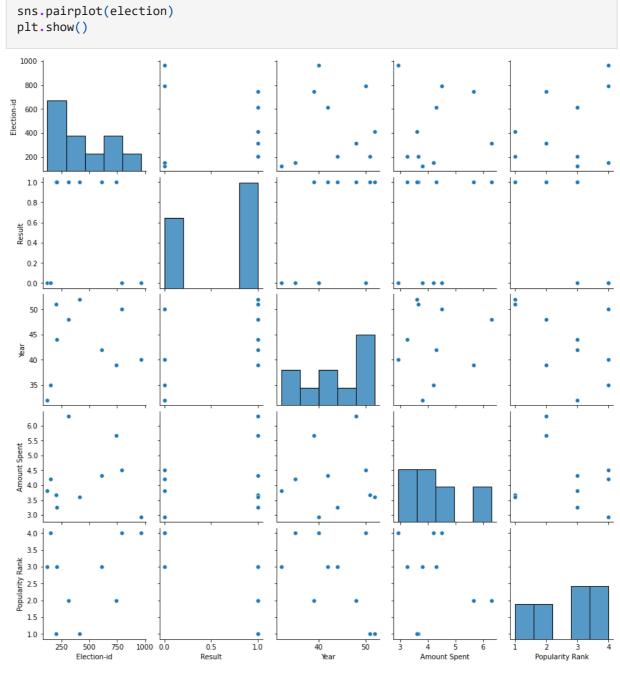
5

	Election-id	Result	Year	Amount Spent	Popularity Rank
6	150.0	0.0	35.0	4.20	4.0
7	743.0	1.0	39.0	5.66	2.0
8	612.0	1.0	42.0	4.32	3.0
9	206.0	1.0	44.0	3.26	3.0
10	792.0	0.0	50.0	4.52	4.0

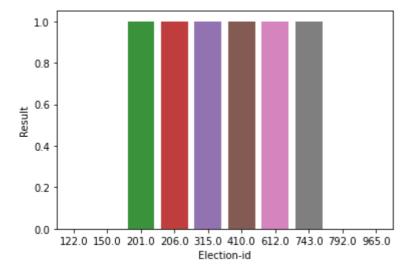
In [16]: election.shape

Out[16]: (10, 5)

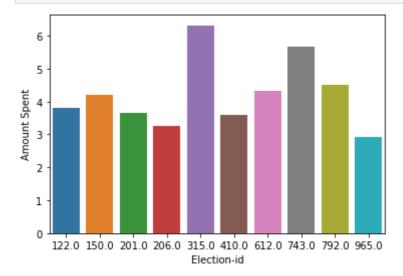
In [19]:



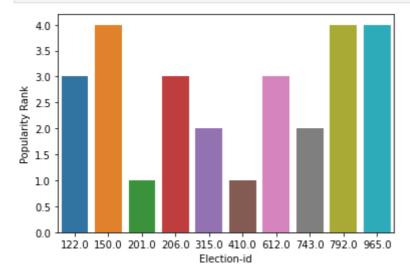
In [30]:
 sns.barplot(x = 'Election-id' , y = 'Result' ,data = election)
 plt.show()



```
In [33]:
    sns.barplot( x = 'Election-id', y = 'Amount Spent' ,data = election)
    plt.show()
```



```
In [36]:
    sns.barplot(x = 'Election-id' , y = 'Popularity Rank' ,data = election)
    plt.show()
```



Model Building

In [37]:

Model Training

```
In [45]:
          classifier = LogisticRegression()
          classifier.fit(x_train,y_train)
Out[45]: LogisticRegression()
In [73]:
          y_train_pred = classifier.predict(x_train)
          y_train_pred
Out[73]: array([315., 201., 965., 612., 792., 206., 743., 150.])
In [74]:
          accuracy_score(y_train,y_train_pred)
Out[74]: 1.0
In [75]:
          confusion_matrix(y_train,y_train_pred)
Out[75]: array([[1, 0, 0, 0, 0, 0, 0, 0],
                 [0, 1, 0, 0, 0, 0, 0, 0],
                 [0, 0, 1, 0, 0, 0, 0, 0],
                 [0, 0, 0, 1, 0, 0, 0, 0],
                 [0, 0, 0, 0, 1, 0, 0, 0],
                 [0, 0, 0, 0, 0, 1, 0, 0],
                 [0, 0, 0, 0, 0, 0, 1, 0],
                 [0, 0, 0, 0, 0, 0, 1]], dtype=int64)
In [76]:
          print(classification_report(y_train,y_train_pred))
                        precision
                                     recall f1-score
                                                         support
                 150.0
                             1.00
                                        1.00
                                                  1.00
                                                               1
                 201.0
                             1.00
                                        1.00
                                                  1.00
                                                               1
                 206.0
                             1.00
                                        1.00
                                                  1.00
                                                               1
                 315.0
                             1.00
                                        1.00
                                                  1.00
                                                               1
                 612.0
                             1.00
                                        1.00
                                                  1.00
                                                               1
                 743.0
                             1.00
                                        1.00
                                                  1.00
                                                               1
                 792.0
                             1.00
                                        1.00
                                                  1.00
                                                               1
                 965.0
                             1.00
                                        1.00
                                                  1.00
                                                               1
```

```
accuracy 1.00 8 macro avg 1.00 1.00 1.00 8 weighted avg 1.00 1.00 1.00 8
```

```
actual predicted_prob
Out[77]:
                  122.0
                                   150.0
             1
                  315.0
                                   315.0
             3
                  201.0
                                   201.0
                  965.0
                                   965.0
             5
                  410.0
                                   201.0
                  150.0
                                   150.0
                                   743.0
             7
                  743.0
             8
                  612.0
                                   612.0
             9
                  206.0
                                   206.0
                  792.0
                                   792.0
            10
```

```
In [95]: ### roc curve

fpr, tpr, thresholds = roc_curve(y_train,classifier.predict_proba (x)[:,1])
```

```
Traceback (most recent call last)
<ipython-input-95-1126a337eff0> in <module>
      3
      4
   -> 5 fpr, tpr, thresholds = roc_curve(y_train,classifier.predict_proba (x)[:,1])
~\anaconda3.01\lib\site-packages\sklearn\utils\validation.py in inner f(*args, **kwa
rgs)
     61
                    extra_args = len(args) - len(all_args)
     62
                    if extra_args <= 0:</pre>
                        return f(*args, **kwargs)
---> 63
     64
                    # extra_args > 0
~\anaconda3.01\lib\site-packages\sklearn\metrics\_ranking.py in roc_curve(y_true, y_
score, pos_label, sample_weight, drop_intermediate)
    911
            .....
    912
--> 913
            fps, tps, thresholds = _binary_clf_curve(
```

```
914
                        y_true, y_score, pos_label=pos_label, sample_weight=sample_weight)
            915
        ~\anaconda3.01\lib\site-packages\sklearn\metrics\_ranking.py in _binary_clf_curve(y_
        true, y_score, pos_label, sample_weight)
            689
                   if not (y_type == "binary" or
            690
                            (y_type == "multiclass" and pos_label is not None)):
        --> 691
                        raise ValueError("{0} format is not supported".format(y_type))
            692
            693
                    check_consistent_length(y_true, y_score, sample_weight)
        ValueError: multiclass format is not supported
In [ ]:
```