

Importing Essential Libraries

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from matplotlib import rcParams
from matplotlib.cm import rainbow
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score
from sklearn.metrics import classification_report
from sklearn import *

from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
```

Importing Dataset

```
heartData=pd.read_csv(r"C:\Users\navya\Downloads\hearts.csv");
```

```
heartData.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1025 entries, 0 to 1024
Data columns (total 14 columns):
#   Column      Non-Null Count  Dtype
---  -
0   age         1025 non-null   int64
1   sex         1025 non-null   int64
2   cp          1025 non-null   int64
3   trestbps    1025 non-null   int64
4   chol        1025 non-null   int64
5   fbs         1025 non-null   int64
6   restecg     1025 non-null   int64
7   thalach     1025 non-null   int64
8   exang       1025 non-null   int64
9   oldpeak     1025 non-null   float64
10  slope       1025 non-null   int64
11  ca          1025 non-null   int64
12  thal        1025 non-null   int64
```



```

2      70      1      0      145      174      0      1      125      1
2.6
3      61      1      0      148      203      0      1      161      0
0.0
4      62      0      0      138      294      1      1      106      0
1.9
...      ...      ...      ..      ...      ...      ...      ...      ...
...
1020    59      1      1      140      221      0      1      164      1
0.0
1021    60      1      0      125      258      0      0      141      1
2.8
1022    47      1      0      110      275      0      0      118      1
1.0
1023    50      0      0      110      254      0      0      159      0
0.0
1024    54      1      0      120      188      0      1      113      0
1.4

```

```

      slope  ca  thal  target
0         2   2    3       0
1         0   0    3       0
2         0   0    3       0
3         2   1    3       0
4         1   3    2       0
...      ...  ..   ...    ...
1020      2   0    2       1
1021      1   1    3       0
1022      1   1    2       0
1023      2   0    2       1
1024      1   1    3       0

```

```
[1025 rows x 14 columns]
```

Description

```
heartData.describe()
```

	age	sex	cp	trestbps	chol
\count	1025.000000	1025.000000	1025.000000	1025.000000	1025.000000
mean	54.434146	0.695610	0.942439	131.611707	246.000000
std	9.072290	0.460373	1.029641	17.516718	51.59251
min	29.000000	0.000000	0.000000	94.000000	126.00000

25%	48.000000	0.000000	0.000000	120.000000	211.000000
50%	56.000000	1.000000	1.000000	130.000000	240.000000
75%	61.000000	1.000000	2.000000	140.000000	275.000000
max	77.000000	1.000000	3.000000	200.000000	564.000000
	fbs	restecg	thalach	exang	oldpeak
\					
count	1025.000000	1025.000000	1025.000000	1025.000000	1025.000000
mean	0.149268	0.529756	149.114146	0.336585	1.071512
std	0.356527	0.527878	23.005724	0.472772	1.175053
min	0.000000	0.000000	71.000000	0.000000	0.000000
25%	0.000000	0.000000	132.000000	0.000000	0.000000
50%	0.000000	1.000000	152.000000	0.000000	0.800000
75%	0.000000	1.000000	166.000000	1.000000	1.800000
max	1.000000	2.000000	202.000000	1.000000	6.200000
	slope	ca	thal	target	
count	1025.000000	1025.000000	1025.000000	1025.000000	
mean	1.385366	0.754146	2.323902	0.513171	
std	0.617755	1.030798	0.620660	0.500070	
min	0.000000	0.000000	0.000000	0.000000	
25%	1.000000	0.000000	2.000000	0.000000	
50%	1.000000	0.000000	2.000000	1.000000	
75%	2.000000	1.000000	3.000000	1.000000	
max	2.000000	4.000000	3.000000	1.000000	

Missing Values

```
heartData.isnull().sum()
```

```
age      0
sex      0
cp       0
trestbps 0
chol     0
fbs      0
restecg  0
```

```

thalach      0
exang        0
oldpeak      0
slope        0
ca           0
thal         0
target       0
dtype: int64

missing_data= heartData.isnull().sum()
total_percentage = (missing_data.sum()/heartData.shape[0]) * 100
print(f'Total percentage of missing data is
{round(total_percentage,2)}%')
duplicate=heartData[heartData.duplicated()]
print("Duplicate rows:")
duplicate
#drop duplicate rows
heartData=heartData.drop_duplicates()

Total percentage of missing data is 0.0%
Duplicate rows:

```

Analysing the 'target variable'

```

heartData["target"].describe()

count      302.000000
mean       0.543046
std        0.498970
min        0.000000
25%        0.000000
50%        1.000000
75%        1.000000
max        1.000000
Name: target, dtype: float64

print(heartData.corr()["target"].abs().sort_values(ascending=False))

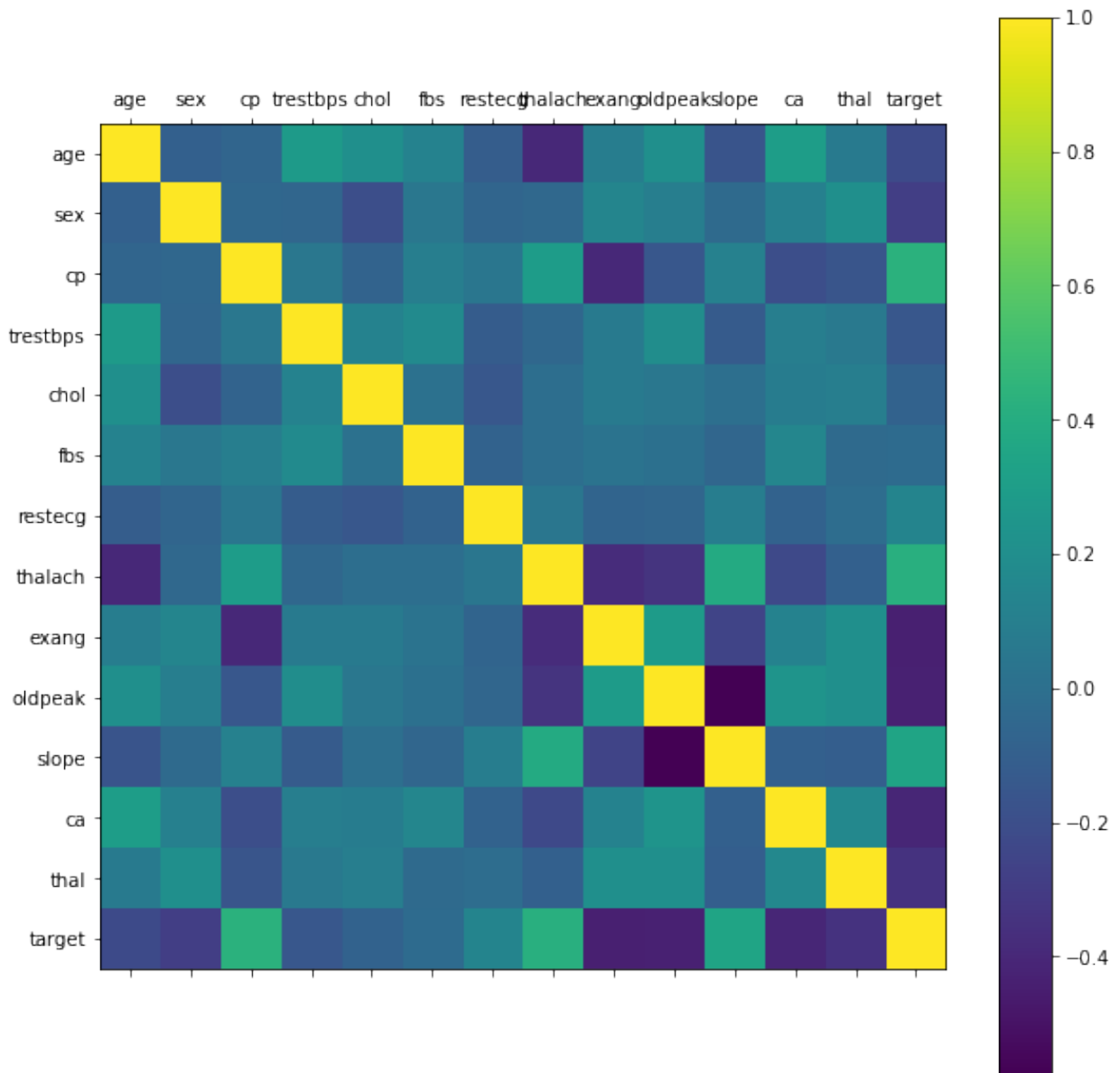
target      1.000000
exang       0.435601
cp          0.432080
oldpeak     0.429146
thalach     0.419955
ca          0.408992
slope       0.343940
thal        0.343101
sex         0.283609
age         0.221476
trestbps    0.146269

```

```
restecg    0.134874
chol       0.081437
fbs        0.026826
Name: target, dtype: float64
```

```
rcParams['figure.figsize'] = 10,10
plt.matshow(heartData.corr())
plt.yticks(np.arange(heartData.shape[1]), heartData.columns)
plt.xticks(np.arange(heartData.shape[1]), heartData.columns)
plt.colorbar()
```

```
<matplotlib.colorbar.Colorbar at 0x1d5bf82afd0>
```

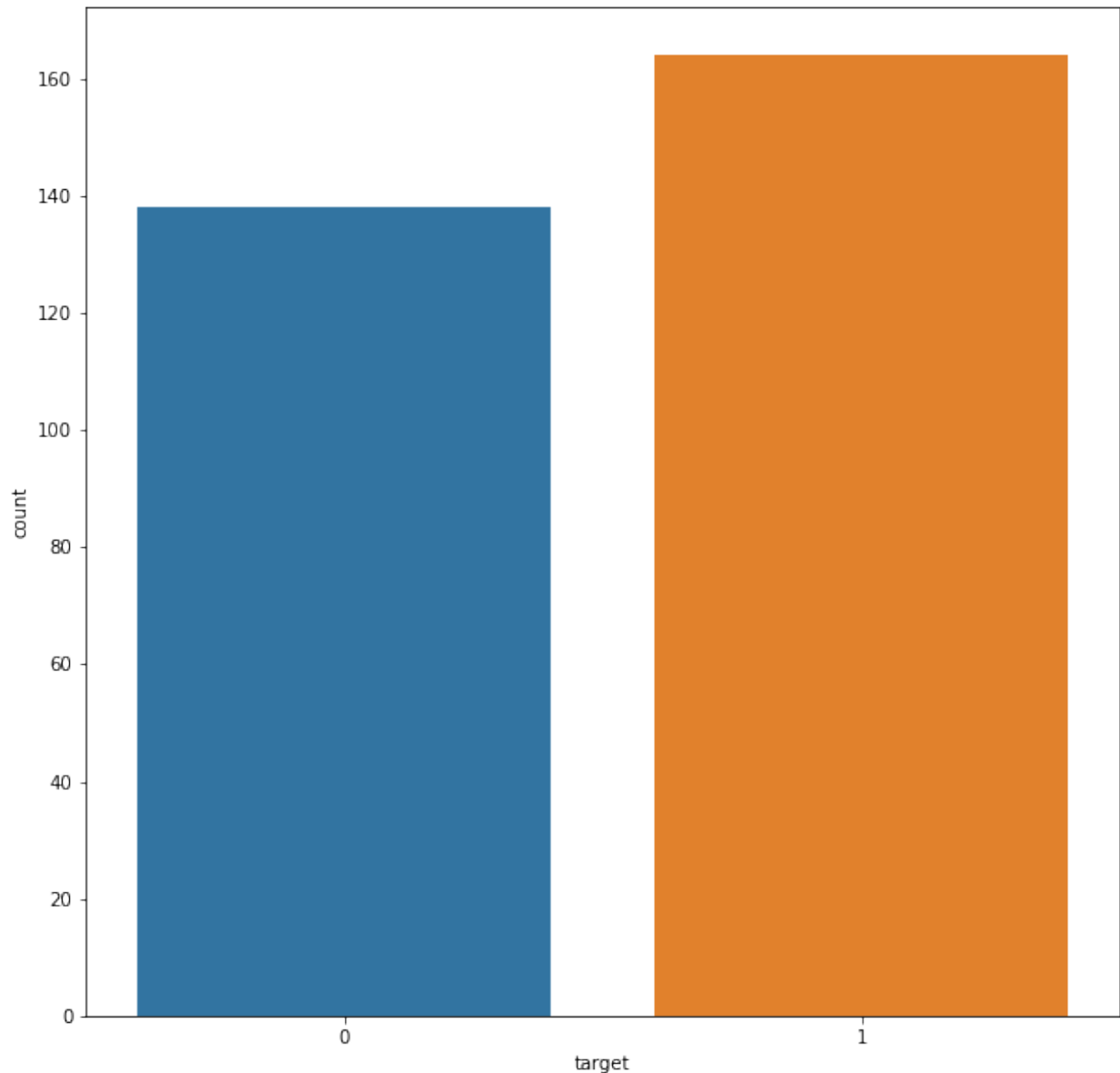


checking correlation between columns

```
corr=heartData.corr()  
corr.style.background_gradient(cmap='coolwarm')  
<pandas.io.formats.style.Styler at 0x1d5bf13d100>
```

EXPLORATORY DATA ANALYSIS

```
y = heartData["target"]  
sns.countplot(y)  
  
target_temp = heartData.target.value_counts()  
print(target_temp)  
1    164  
0    138  
Name: target, dtype: int64
```



```
print("Percentage of patience without heart problems:  
"+str(round(target_temp[0]*100/1025,2)))  
print("Percentage of patience with heart problems:  
"+str(round(target_temp[1]*100/1025,2)))
```

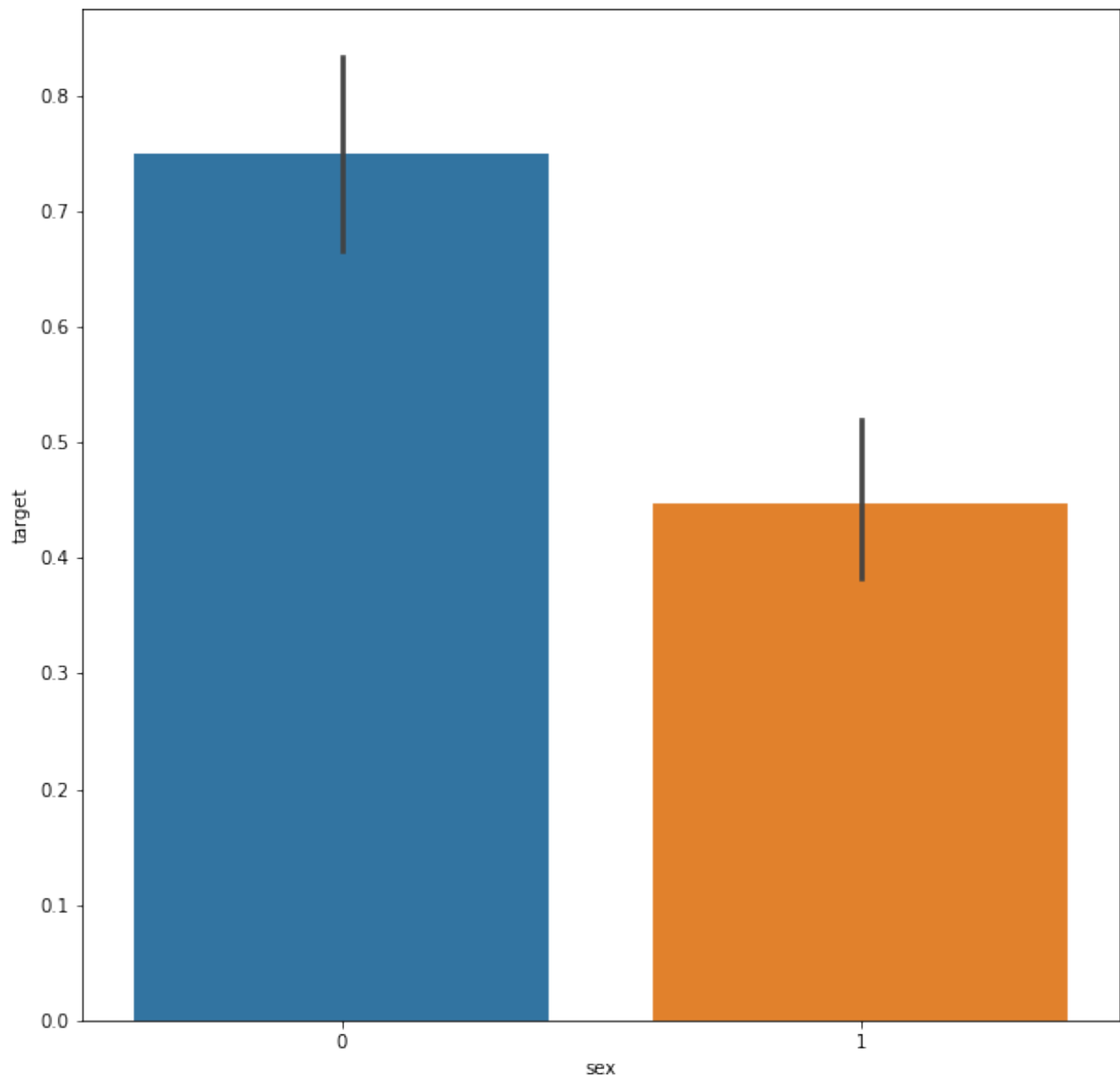
Percentage of patience without heart problems: 13.46
Percentage of patience with heart problems: 16.0

Analysing the 'Sex' feature

```
heartData["sex"].unique()
```



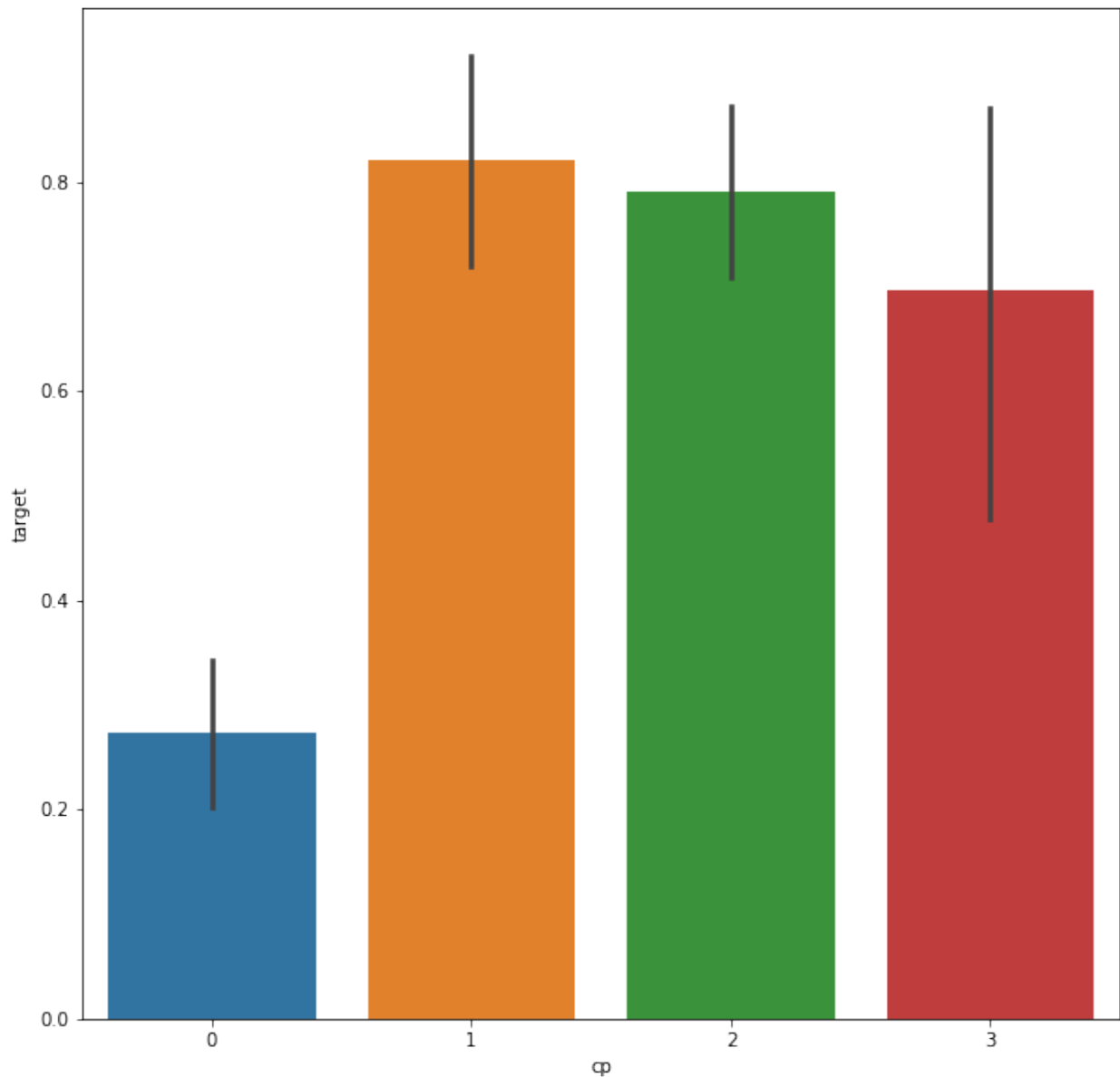
```
array([1, 0], dtype=int64)
sns.barplot(heartData["sex"],y)
<AxesSubplot:xlabel='sex', ylabel='target'>
```



Analysing the 'Chest Pain Type' feature

```
heartData["cp"].unique()
array([0, 1, 2, 3], dtype=int64)
```

```
sns.barplot(heartData["cp"],y)
<AxesSubplot:xlabel='cp', ylabel='target'>
```



Analysing the FBS feature

```
heartData["fbs"].describe()
```

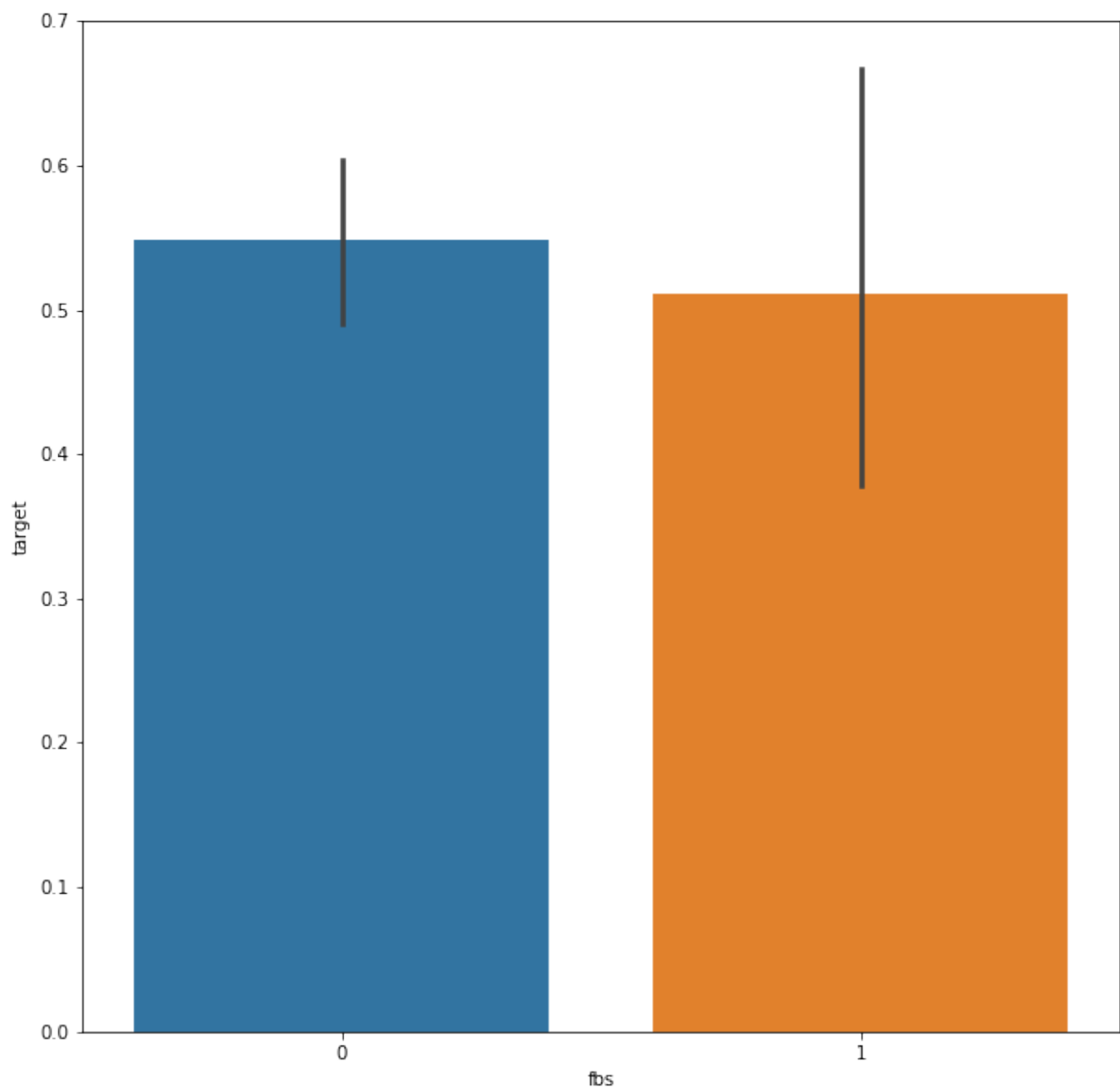
count	302.000000
mean	0.149007
std	0.356686
min	0.000000

```
25%      0.000000
50%      0.000000
75%      0.000000
max       1.000000
Name: fbs, dtype: float64

heartData["fbs"].unique()
array([0, 1], dtype=int64)

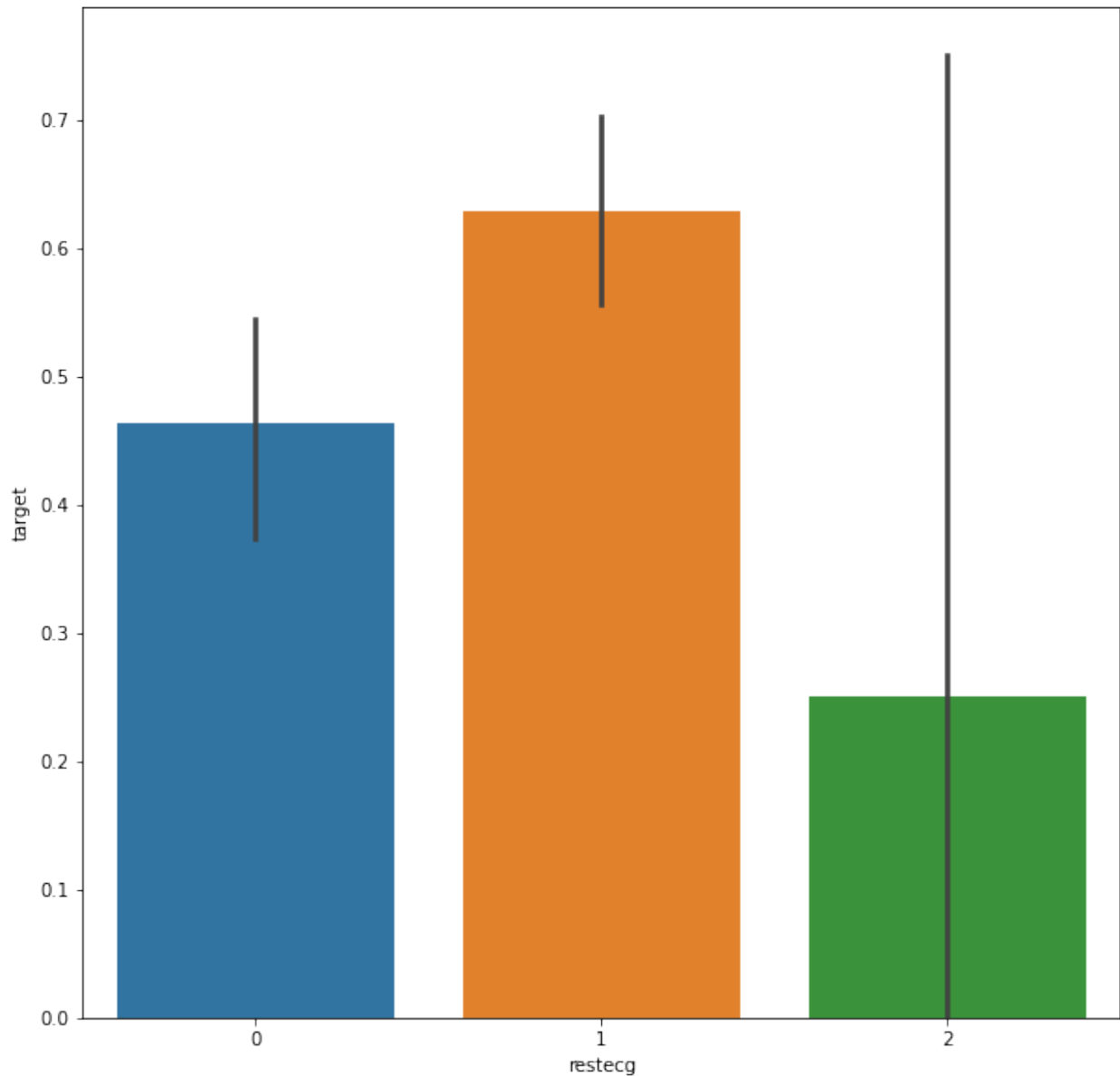
sns.barplot(heartData["fbs"],y)

<AxesSubplot:xlabel='fbs', ylabel='target'>
```



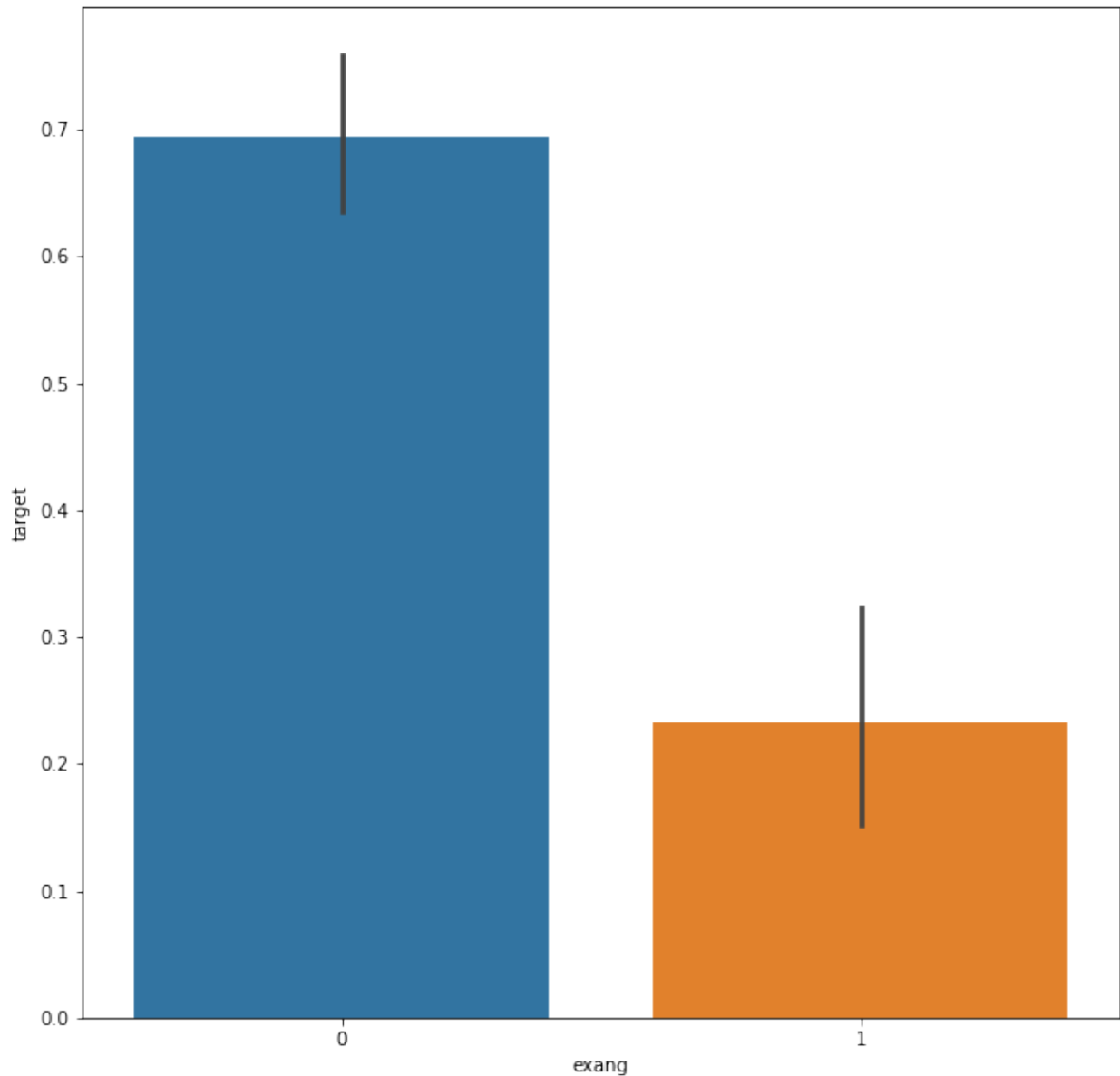
Analysing the restecg feature

```
heartData["restecg"].unique()  
array([1, 0, 2], dtype=int64)  
sns.barplot(heartData["restecg"],y)  
<AxesSubplot:xlabel='restecg', ylabel='target'>
```



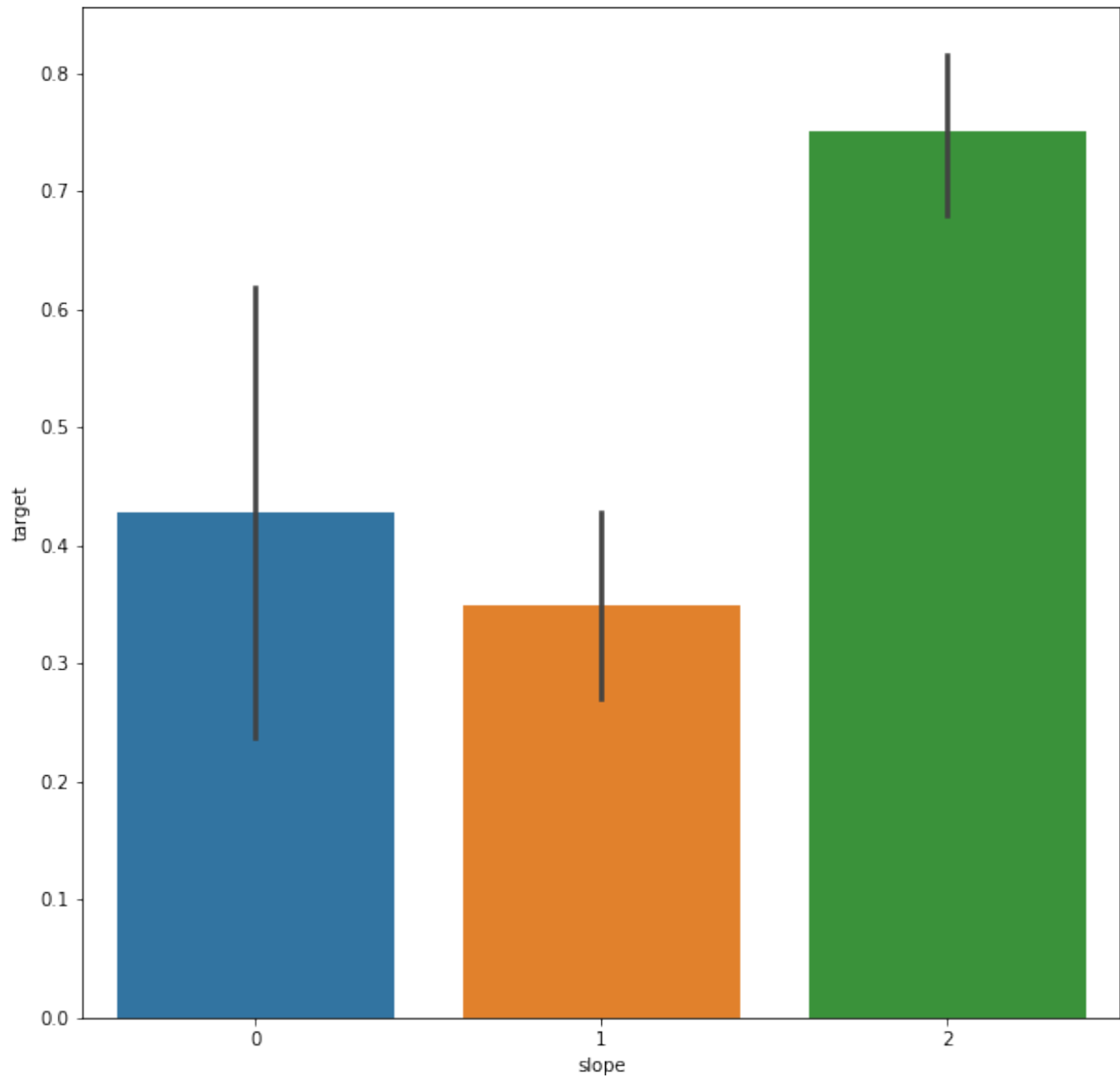
Analysing the 'exang' feature

```
heartData["exang"].unique()  
array([0, 1], dtype=int64)  
sns.barplot(heartData["exang"],y)  
<AxesSubplot:xlabel='exang', ylabel='target'>
```



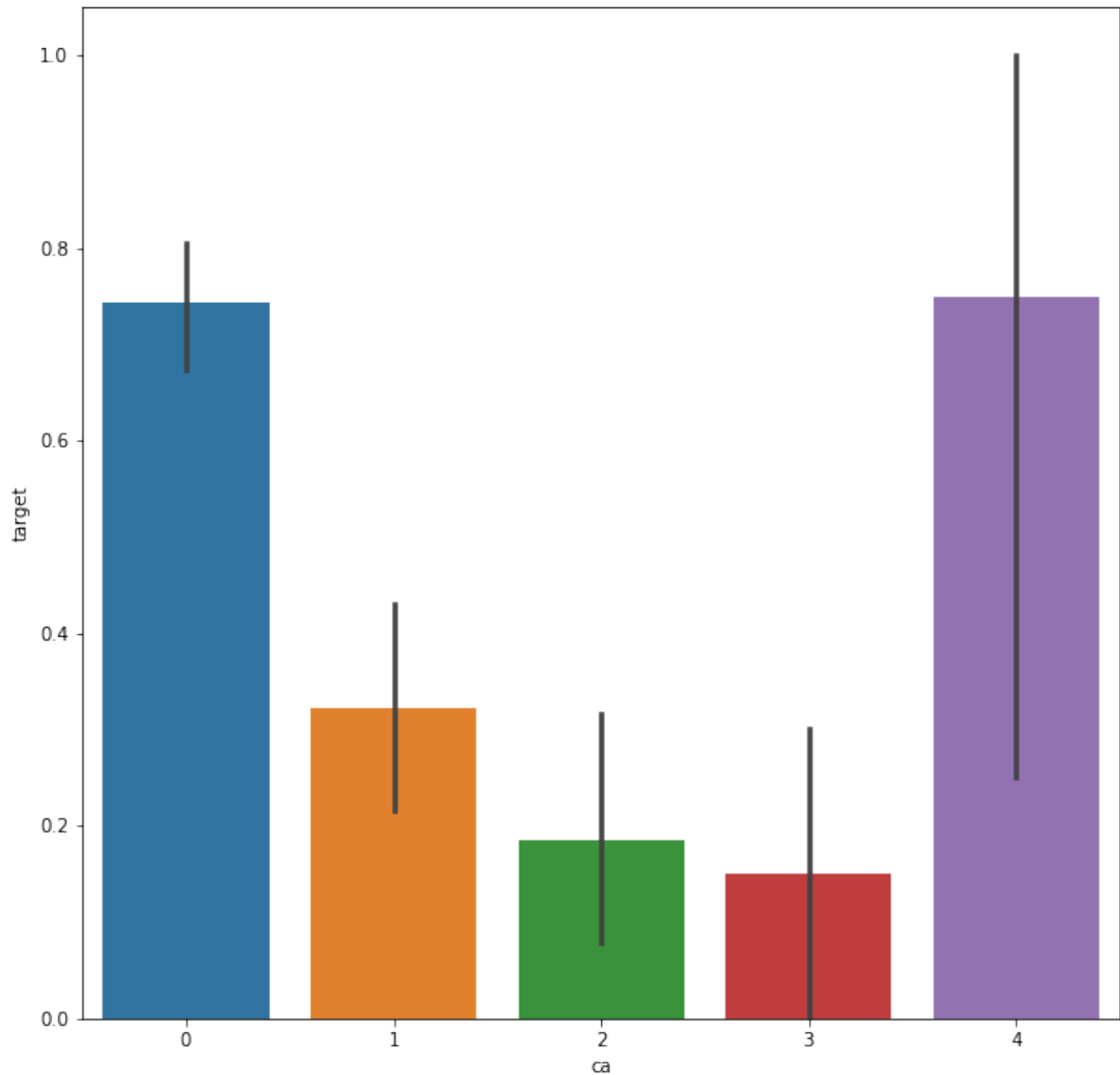
Analysing the 'Slope' feature

```
heartData["slope"].unique()  
array([2, 0, 1], dtype=int64)  
sns.barplot(heartData["slope"],y)  
<AxesSubplot:xlabel='slope', ylabel='target'>
```



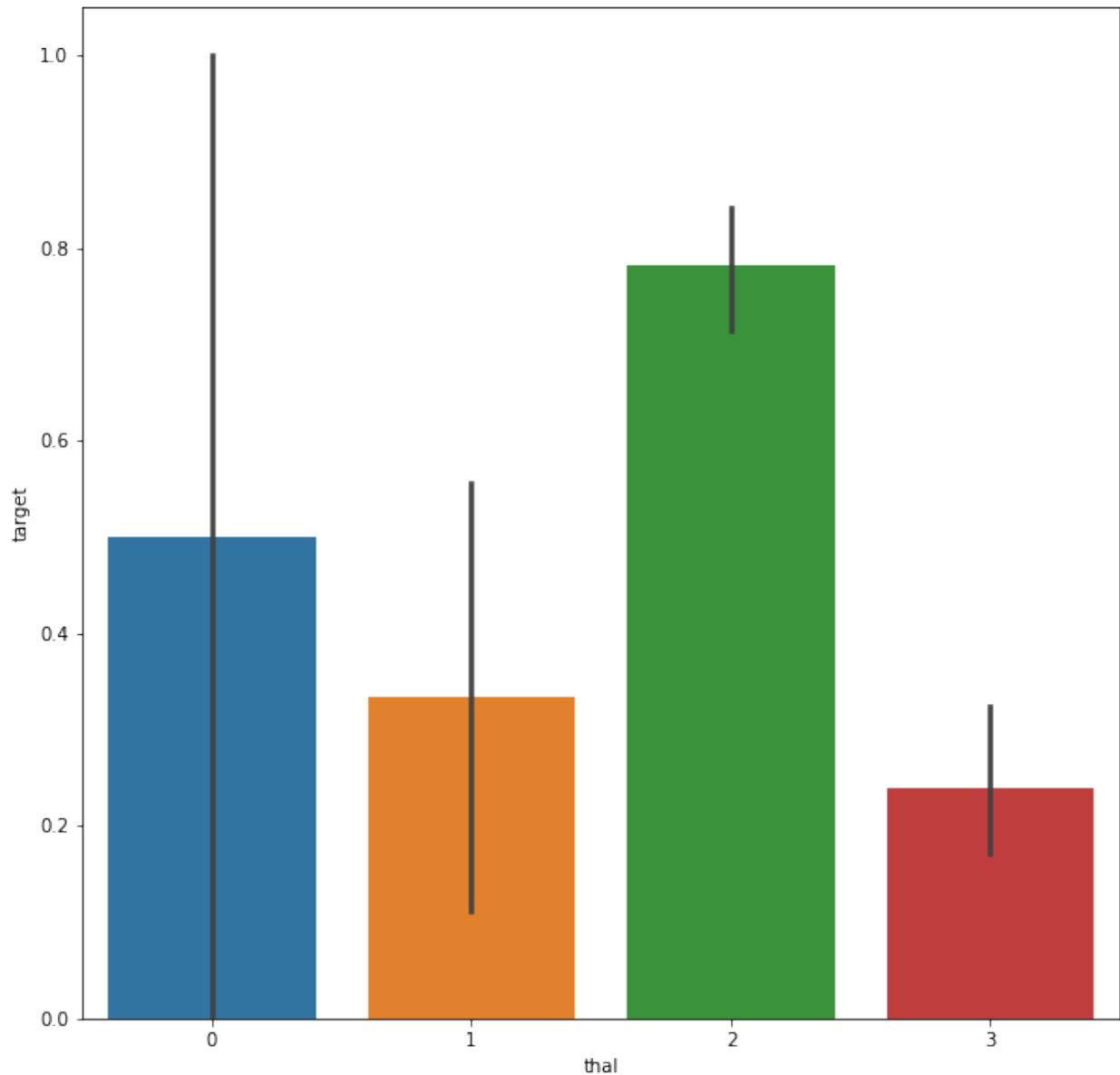
Analysing the 'ca' feature

```
heartData["ca"].unique()  
array([2, 0, 1, 3, 4], dtype=int64)  
sns.barplot(heartData["ca"],y)  
<AxesSubplot:xlabel='ca', ylabel='target'>
```



Analysing the 'thal' feature

```
heartData["thal"].unique()  
array([3, 2, 1, 0], dtype=int64)  
sns.barplot(heartData["thal"],y)  
<AxesSubplot:xlabel='thal', ylabel='target'>
```



Train Test Split

```
from sklearn.model_selection import train_test_split

predictors = heartData.drop("target",axis=1)
target = heartData["target"]

X_train,X_test,Y_train,Y_test =
train_test_split(predictors,target,test_size=0.20,random_state=0)

X_train.shape
(241, 13)

X_test.shape
(61, 13)

Y_train.shape
(241,)

Y_test.shape
(61,)
```

Model Fitting

```
from sklearn.metrics import accuracy_score
```

Logistic Regression

```
from sklearn.linear_model import LogisticRegression

lr = LogisticRegression()

lr.fit(X_train,Y_train)

Y_pred_lr = lr.predict(X_test)

Y_pred_lr.shape
(61,)

score_lr = round(accuracy_score(Y_pred_lr,Y_test)*100,2)

print("The accuracy score achieved using Logistic Regression is:
"+str(score_lr)+" %")
```

The accuracy score achieved using Logistic Regression is: 83.61 %

Naive Bayes

```
from sklearn.naive_bayes import GaussianNB
nb = GaussianNB()
nb.fit(X_train,Y_train)
Y_pred_nb = nb.predict(X_test)
Y_pred_nb.shape
(61,)
score_nb = round(accuracy_score(Y_pred_nb,Y_test)*100,2)
print("The accuracy score achieved using Naive Bayes is: "+str(score_nb)+" %")
The accuracy score achieved using Naive Bayes is: 80.33 %
```

K Nearest Neighbors

```
from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors=7)
knn.fit(X_train,Y_train)
Y_pred_knn=knn.predict(X_test)
Y_pred_knn.shape
(61,)
score_knn = round(accuracy_score(Y_pred_knn,Y_test)*100,2)
print("The accuracy score achieved using KNN is: "+str(score_knn)+" %")
The accuracy score achieved using KNN is: 65.57 %
```

Decision Tree

```
from sklearn.tree import DecisionTreeClassifier
max_accuracy = 0
```

```

for x in range(200):
    dt = DecisionTreeClassifier(random_state=x)
    dt.fit(X_train,Y_train)
    Y_pred_dt = dt.predict(X_test)
    current_accuracy = round(accuracy_score(Y_pred_dt,Y_test)*100,2)
    if(current_accuracy>max_accuracy):
        max_accuracy = current_accuracy
        best_x = x

#print(max_accuracy)
#print(best_x)

dt = DecisionTreeClassifier(random_state=best_x)
dt.fit(X_train,Y_train)
Y_pred_dt = dt.predict(X_test)

print(Y_pred_dt.shape)

(61,)

score_dt = round(accuracy_score(Y_pred_dt,Y_test)*100,2)

print("The accuracy score achieved using Decision Tree is:
"+str(score_dt)+" %")

The accuracy score achieved using Decision Tree is: 85.25 %

```

Random Forest

```

from sklearn.ensemble import RandomForestClassifier

max_accuracy = 0

for x in range(2000):
    rf = RandomForestClassifier(random_state=x)
    rf.fit(X_train,Y_train)
    Y_pred_rf = rf.predict(X_test)
    current_accuracy = round(accuracy_score(Y_pred_rf,Y_test)*100,2)
    if(current_accuracy>max_accuracy):
        max_accuracy = current_accuracy
        best_x = x

#print(max_accuracy)
#print(best_x)

rf = RandomForestClassifier(random_state=best_x)
rf.fit(X_train,Y_train)
Y_pred_rf = rf.predict(X_test)

```

```
Y_pred_rf.shape
```

```
(61,)
```

```
score_rf = round(accuracy_score(Y_pred_rf,Y_test)*100,2)
```

```
print("The accuracy score achieved using Random Forest Tree is:  
"+str(score_rf)+" %")
```

```
The accuracy score achieved using Decision Tree is: 86.89 %
```

Output Final score

```
scores = [score_lr,score_nb,score_knn,score_dt,score_rf]  
algorithms = ["Logistic Regression","Naive Bayes","K-Nearest  
Neighbors","Decision Tree","Random Forest"]
```

```
for i in range(len(algorithms)):  
    print("The accuracy score achieved using "+algorithms[i]+" is:  
"+str(scores[i])+" %")
```

```
The accuracy score achieved using Logistic Regression is: 83.61 %
```

```
The accuracy score achieved using Naive Bayes is: 80.33 %
```

```
The accuracy score achieved using K-Nearest Neighbors is: 65.57 %
```

```
The accuracy score achieved using Decision Tree is: 85.25 %
```

```
The accuracy score achieved using Random Forest is: 86.89 %
```

```
sns.set(rc={'figure.figsize':(15,8)})
```

```
plt.xlabel("Algorithms")
```

```
plt.ylabel("Accuracy score")
```

```
sns.barplot(algorithms,scores)
```

```
<AxesSubplot:xlabel='Algorithms', ylabel='Accuracy score'>
```

