import pandas as pd
heart_dis=pd.read_excel("heart_disease.xlsx")

heart_dis

| | age | sex | ср | trestbps | chol | fbs | restecg | thalach | exang | oldpeak | slope | ca | thal | target |
|-----|-----|-----|----|----------|------|-----|---------|---------|-------|---------|-------|----|------|--------|
| 0 | 63 | 1 | 3 | 145 | 233 | 1 | 0 | 150 | 0 | 2.3 | 0 | 0 | 1 | 1 |
| 1 | 37 | 1 | 2 | 130 | 250 | 0 | 1 | 187 | 0 | 3.5 | 0 | 0 | 2 | 1 |
| 2 | 41 | 0 | 1 | 130 | 204 | 0 | 0 | 172 | 0 | 1.4 | 2 | 0 | 2 | 1 |
| 3 | 56 | 1 | 1 | 120 | 236 | 0 | 1 | 178 | 0 | 8.0 | 2 | 0 | 2 | 1 |
| 4 | 57 | 0 | 0 | 120 | 354 | 0 | 1 | 163 | 1 | 0.6 | 2 | 0 | 2 | 1 |
| | | | | | | | | | | | | | | |
| 298 | 57 | 0 | 0 | 140 | 241 | 0 | 1 | 123 | 1 | 0.2 | 1 | 0 | 3 | 0 |
| 299 | 45 | 1 | 3 | 110 | 264 | 0 | 1 | 132 | 0 | 1.2 | 1 | 0 | 3 | 0 |
| 300 | 68 | 1 | 0 | 144 | 193 | 1 | 1 | 141 | 0 | 3.4 | 1 | 2 | 3 | 0 |
| 301 | 57 | 1 | 0 | 130 | 131 | 0 | 1 | 115 | 1 | 1.2 | 1 | 1 | 3 | 0 |
| 302 | 57 | 0 | 1 | 130 | 236 | 0 | 0 | 174 | 0 | 0.0 | 1 | 1 | 2 | 0 |

303 rows × 14 columns

heart_dis.info()

```
303 non-null
1
                              int64
    sex
              303 non-null
2
                              int64
   ср
   trestbps 303 non-null
                              int64
    chol
              303 non-null
                              int64
   fbs
              303 non-null
                              int64
             303 non-null
                              int64
   restecg
7
   thalach
             303 non-null
                              int64
              303 non-null
                              int64
8
    exang
   oldpeak
             303 non-null
                              float64
9
10 slope
              303 non-null
                              int64
11 ca
              303 non-null
                              int64
12 thal
              303 non-null
                              int64
13 target
              303 non-null
                              int64
```

dtypes: float64(1), int64(13)

memory usage: 33.3 KB

heart dis.describe()

| | age | sex | ср | trestbps | chol | fbs | restecg | thalach | |
|-------|------------|------------|------------|------------|------------|------------|------------|------------|-----|
| count | 303.000000 | 303.000000 | 303.000000 | 303.000000 | 303.000000 | 303.000000 | 303.000000 | 303.000000 | 303 |
| mean | 54.366337 | 0.683168 | 0.966997 | 131.623762 | 246.264026 | 0.148515 | 0.528053 | 149.646865 | 0 |
| std | 9.082101 | 0.466011 | 1.032052 | 17.538143 | 51.830751 | 0.356198 | 0.525860 | 22.905161 | 0 |
| min | 29.000000 | 0.000000 | 0.000000 | 94.000000 | 126.000000 | 0.000000 | 0.000000 | 71.000000 | 0 |
| 25% | 47.500000 | 0.000000 | 0.000000 | 120.000000 | 211.000000 | 0.000000 | 0.000000 | 133.500000 | 0 |
| 50% | 55.000000 | 1.000000 | 1.000000 | 130.000000 | 240.000000 | 0.000000 | 1.000000 | 153.000000 | 0 |
| 75% | 61.000000 | 1.000000 | 2.000000 | 140.000000 | 274.500000 | 0.000000 | 1.000000 | 166.000000 | 1 |
| max | 77.000000 | 1.000000 | 3.000000 | 200.000000 | 564.000000 | 1.000000 | 2.000000 | 202.000000 | 1 |

y_dep=heart_dis.target

x_ind=heart_dis.drop("target",axis=1)

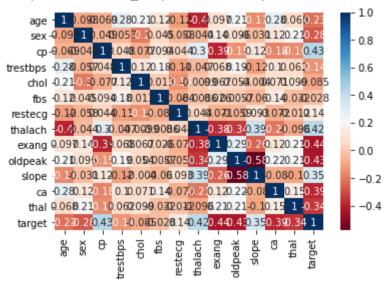
h_corr=heart_dis.corr()
h_corr

| | age | sex | ср | trestbps | chol | fbs | restecg | thalach | exang | |
|----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|----|
| age | 1.000000 | -0.098447 | -0.068653 | 0.279351 | 0.213678 | 0.121308 | -0.116211 | -0.398522 | 0.096801 | (|
| sex | -0.098447 | 1.000000 | -0.049353 | -0.056769 | -0.197912 | 0.045032 | -0.058196 | -0.044020 | 0.141664 | (|
| ср | -0.068653 | -0.049353 | 1.000000 | 0.047608 | -0.076904 | 0.094444 | 0.044421 | 0.295762 | -0.394280 | -(|
| trestbps | 0.279351 | -0.056769 | 0.047608 | 1.000000 | 0.123174 | 0.177531 | -0.114103 | -0.046698 | 0.067616 | (|
| chol | 0.213678 | -0.197912 | -0.076904 | 0.123174 | 1.000000 | 0.013294 | -0.151040 | -0.009940 | 0.067023 | (|
| fbs | 0.121308 | 0.045032 | 0.094444 | 0.177531 | 0.013294 | 1.000000 | -0.084189 | -0.008567 | 0.025665 | (|
| restecg | -0.116211 | -0.058196 | 0.044421 | -0.114103 | -0.151040 | -0.084189 | 1.000000 | 0.044123 | -0.070733 | -(|
| thalach | -0.398522 | -0.044020 | 0.295762 | -0.046698 | -0.009940 | -0.008567 | 0.044123 | 1.000000 | -0.378812 | -(|
| exang | 0.096801 | 0.141664 | -0.394280 | 0.067616 | 0.067023 | 0.025665 | -0.070733 | -0.378812 | 1.000000 | (|
| oldpeak | 0.210013 | 0.096093 | -0.149230 | 0.193216 | 0.053952 | 0.005747 | -0.058770 | -0.344187 | 0.288223 | 1 |
| slope | -0.168814 | -0.030711 | 0.119717 | -0.121475 | -0.004038 | -0.059894 | 0.093045 | 0.386784 | -0.257748 | -(|
| са | 0.276326 | 0.118261 | -0.181053 | 0.101389 | 0.070511 | 0.137979 | -0.072042 | -0.213177 | 0.115739 | (|
| thal | 0.068001 | 0.210041 | -0.161736 | 0.062210 | 0.098803 | -0.032019 | -0.011981 | -0.096439 | 0.206754 | (|
| target | -0.225439 | -0.280937 | 0.433798 | -0.144931 | -0.085239 | -0.028046 | 0.137230 | 0.421741 | -0.436757 | -(|

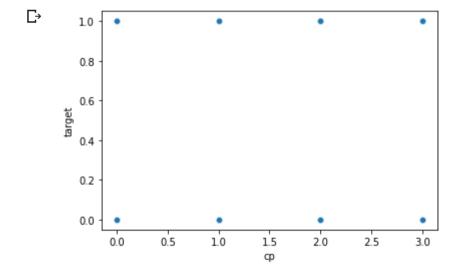
import seaborn as sns
import matplotlib.pyplot as plt

sns.heatmap(h_corr,annot=True,cmap="RdBu")

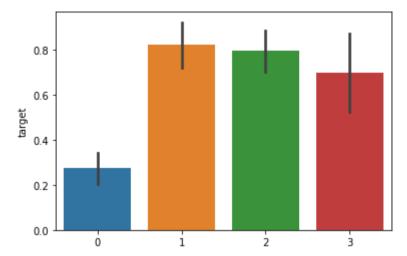
<matplotlib.axes._subplots.AxesSubplot at 0x7fdb2f373250>



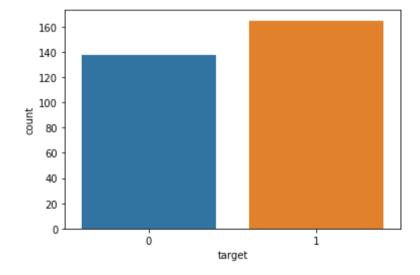
Psns.scatterplot(x="cp",y="target",data=heart dis);



sns.barplot(x="cp",y="target",data=heart_dis);



sns.countplot(x="target",data=heart_dis);

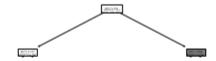


import sklearn
from sklearn import model_selection
from sklearn.model_selection import train_test_split

x_train, x_test, y_train, y_test = train_test_split(x_ind, y_dep, test_size=0.3, random_state=2)

```
from sklearn import tree
model1=tree.DecisionTreeClassifier()
model1.fit(x train,y train)
y pred=model1.predict(x test)
y pred
     array([1, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 1,
            0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0,
            1, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1,
            1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1,
            1, 0, 0])
from sklearn.metrics import accuracy score, confusion matrix
confusion matrix(y test,y pred)
     array([[32, 9],
            [ 3, 47]])
accuracy score(y test,y pred)
     0.8681318681318682
tree.plot tree(model1,max depth=3)
```

```
[\text{Text}(167.4, 195.696, 'X[2] <= 0.5 \mid = 0.496 \mid = 212 \mid = [97, 115]'),
 Text(83.7, 152.208, 'X[11] \le 0.5 \cdot ini = 0.411 \cdot insamples = 104 \cdot invalue = [74, 30]'),
 Text(41.85, 108.72, X[12] \le 2.5 = 0.498 = 47 = 47 = [22, 25]
 Text(20.925, 65.232, 'X[6] \le 0.5 \le 0.5 \le 0.366 \le 29 \le 29 \le 0.366 \le 0
 Text(10.4625, 21.744, '\n (...) \n'),
 Text(31.387500000000003, 21.744, '\n (...) \n'),
 Text(62.77500000000006, 65.232, 'X[9] \le 0.45 \ngini = 0.278\nsamples = 18\nvalue = [15, 3]'),
 Text(52.3125, 21.744, '\n (...) \n'),
 Text(73.2375, 21.744, '\n (...) \n'),
 Text(104.625, 65.232, 'X[1] \le 0.5 \text{ ngini} = 0.391 \text{ nsamples} = 15 \text{ nvalue} = [11, 4]'),
 Text(94.16250000000001, 21.744, '\n (...) \n'),
 Text(115.0875, 21.744, '\n (...) \n'),
 Text(146.475, 65.232, X[4] \le 301.0  ngini = 0.046 \ nsamples = 42 \ nvalue = [41, 1]'),
 Text(136.01250000000002, 21.744, '\n (...) \n'),
 Text(156.9375, 21.744, '\n (...) \n'),
 Text(251.100000000000000, 152.208, X[0] <= 56.5 \ngini = 0.335 \nsamples = 108 \nvalue = [23, 85]'),
 Text(209.25, 108.72, X[12] \le 2.5  = 0.172\nsamples = 63\nvalue = [6, 57]'),
 Text(188.3250000000000, 65.232, X[9] <= 3.55  ngini = 0.075 \nsamples = 51 \nvalue = [2, 49]'),
 Text(177.8625, 21.744, '\n (...) \n'),
 Text(198.7875, 21.744, '\n (...) \n'),
 Text(230.175, 65.232, X[11] \le 0.5 = 0.444 = 12 = 12 = 14, 8]'),
 Text(219.7125, 21.744, '\n (...) \n'),
 Text(240.63750000000002, 21.744, '\n (...) \n'),
 Text(292.95, 108.72, X[1] <= 0.5 \cdot ini = 0.47 \cdot ini = 45 \cdot ini = 17, 28]'
 Text(272.02500000000003, 65.232, 'X[0] <= 57.5 \mid = 0.208 \mid = 17 \mid = [2, 15]'),
 Text(261.5625, 21.744, '\n (...) \n'),
 Text(282.4875, 21.744, '\n (...) \n'),
 Text(313.875, 65.232, 'X[4] \le 245.5 \cdot = 0.497 \cdot = 28 \cdot = [15, 13]'),
 Text(303.4125, 21.744, '\n (...) \n'),
 Text(324.33750000000003, 21.744, '\n (...) \n')]
```



import graphviz
from sklearn.tree import export_graphviz
from six import StringIO
import IPython
from IPython.display import Image

```
import pydotplus

my_graph=StringIO()

export_graphviz(model1,out_file=my_graph,filled=True)

my_graph=pydotplus.graph_from_dot_data(my_graph.getvalue())

my_graph.write_jpg("DT.jpg")

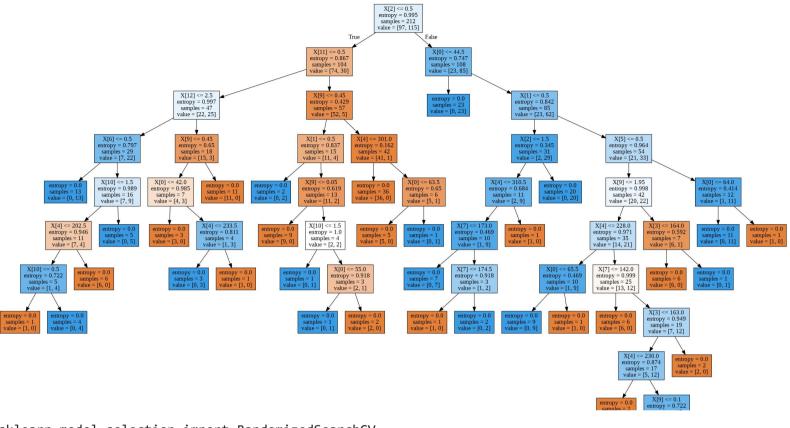
True

Image(my_graph.create_jpg())
```

```
model e=tree.DecisionTreeClassifier(criterion="entropy",random state=2)
     model e.fit(x train,y train)
    DecisionTreeClassifier(ccp alpha=0.0, class weight=None, criterion='entropy',
                          max depth=None, max features=None, max leaf nodes=None,
                          min impurity decrease=0.0, min impurity split=None,
                          min_samples_leaf=1, min_samples_split=2,
                          min weight fraction leaf=0.0, presort='deprecated',
                          random state=2, splitter='best')
y pred e=model e.predict(x test)
confusion_matrix(y_test,y_pred_e)
    array([[28, 13],
           [5, 45]])
accuracy_score(y_test,y_pred_e)
     0.8021978021978022
my_graph=StringIO()
```

my_graph=pydotplus.graph_from_dot_data(my_graph.getvalue())

Image(my_graph.create_jpg())



from sklearn.model_selection import RandomizedSearchCV



samples = 7

DT hp=RandomizedSearchCV(tree.DecisionTreeClassifier(),param distributions=parameters,cv=5)

```
DT_hp.fit(x_train,y_train)
```

DT hp.best estimator

```
max depth=None,
                                                          max features=None,
                                                         max_leaf_nodes=None,
                                                         min_impurity_decrease=0.0,
                                                         min impurity split=None,
                                                         min samples leaf=1,
                                                         min samples split=2,
                                                         min weight fraction leaf=0.0,
                                                          presort='deprecated',
                                                          random state=None,
                                                          splitter='best'),
                        iid='deprecated', n iter=10, n jobs=None,
                        param distributions={'criterion': ('gini', 'entropy'),
                                              'max depth': (10, 20, 30, 40, 50, 60,
                                                            70, 100),
                                              'max features': ('log2', 'auto',
                                                               'sart').
                                              'min samples split': (2, 4, 6)},
                        pre dispatch='2*n jobs', random state=None, refit=True,
                        return train score=False, scoring=None, verbose=0)
     DecisionTreeClassifier(ccp alpha=0.0, class weight=None, criterion='entropy',
                            max depth=40, max features='sqrt', max leaf nodes=None,
                            min impurity decrease=0.0, min impurity split=None,
                            min samples leaf=1, min samples split=2,
                            min weight fraction leaf=0.0, presort='deprecated',
                            random state=None, splitter='best')
model after Ht=tree.DecisionTreeClassifier(criterion='entropy', max depth=40, max features='sqrt')
model after Ht=model after Ht.fit(x train,y train)
pred after hp=model after Ht.predict(x test)
confusion_matrix(y_test,pred_after_hp)
```

finally got the good fit accuracy of 76%