

Heart Stroke Prediction

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib.pyplot import rcParams
from matplotlib.cm import rainbow
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
```

Here we will be experimenting with 2 algorithms

1.KNeighborsClassifier 2.RandomForestClassifier

```
In [2]: from sklearn.neighbors import KNeighborsClassifier
from sklearn.ensemble import RandomForestClassifier
```

```
In [3]: df = pd.read_csv('datasetheart.csv')
```

```
In [4]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 303 entries, 0 to 302
Data columns (total 14 columns):
 #   column      Non-Null Count  Dtype
---  -
 0   age         303 non-null     int64
 1   sex         303 non-null     int64
 2   cp          303 non-null     int64
 3   trestbps    303 non-null     int64
 4   chol        303 non-null     int64
 5   fbs         303 non-null     int64
 6   restecg     303 non-null     int64
 7   thalach     303 non-null     int64
 8   exang       303 non-null     int64
 9   oldpeak     303 non-null     float64
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
```

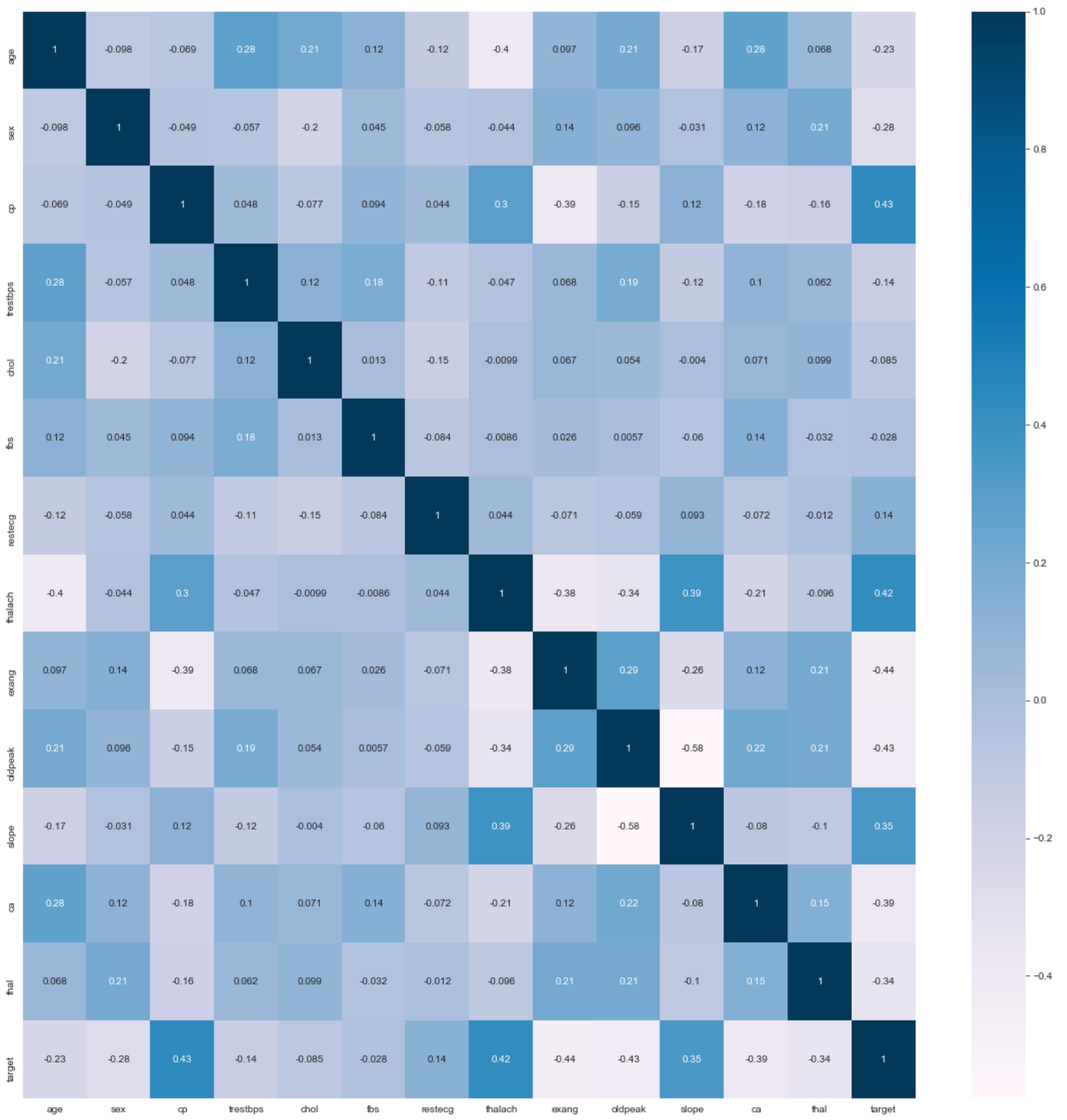
```
In [5]: df.describe()
```

Out[5]:

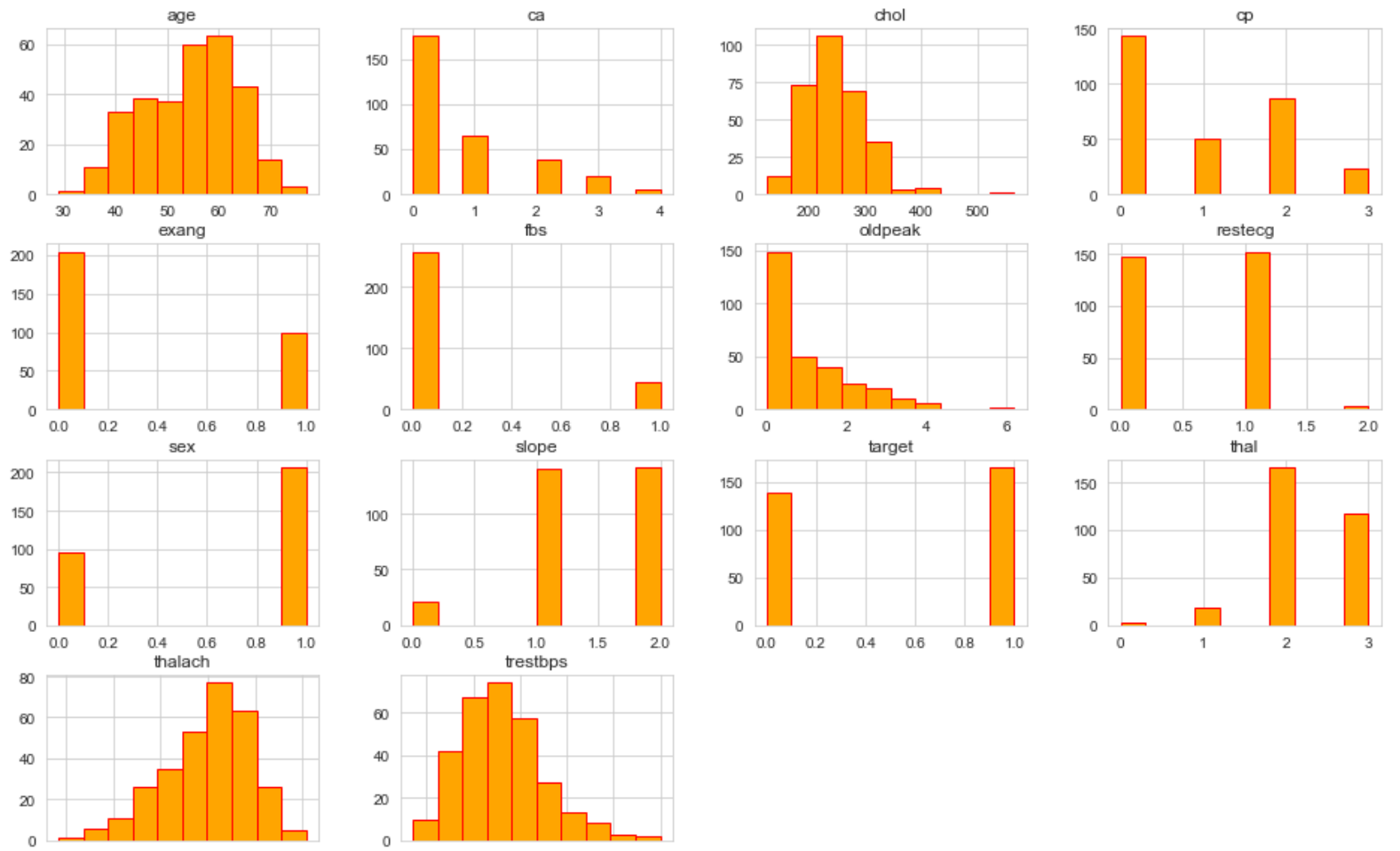
	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpe
count	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000
mean	54.366337	0.683168	0.966997	131.623762	246.264026	0.148515	0.528053	149.646865	0.326733	1.03967
std	9.082101	0.466011	1.032052	17.538143	51.830751	0.356198	0.525860	22.905161	0.469794	1.16107
min	29.000000	0.000000	0.000000	94.000000	126.000000	0.000000	0.000000	71.000000	0.000000	0.00000
25%	47.500000	0.000000	0.000000	120.000000	211.000000	0.000000	0.000000	133.500000	0.000000	0.00000
50%	55.000000	1.000000	1.000000	130.000000	240.000000	0.000000	1.000000	153.000000	0.000000	0.80000
75%	61.000000	1.000000	2.000000	140.000000	274.500000	0.000000	1.000000	166.000000	1.000000	1.60000
max	77.000000	1.000000	3.000000	200.000000	564.000000	1.000000	2.000000	202.000000	1.000000	6.20000

Feature Selection

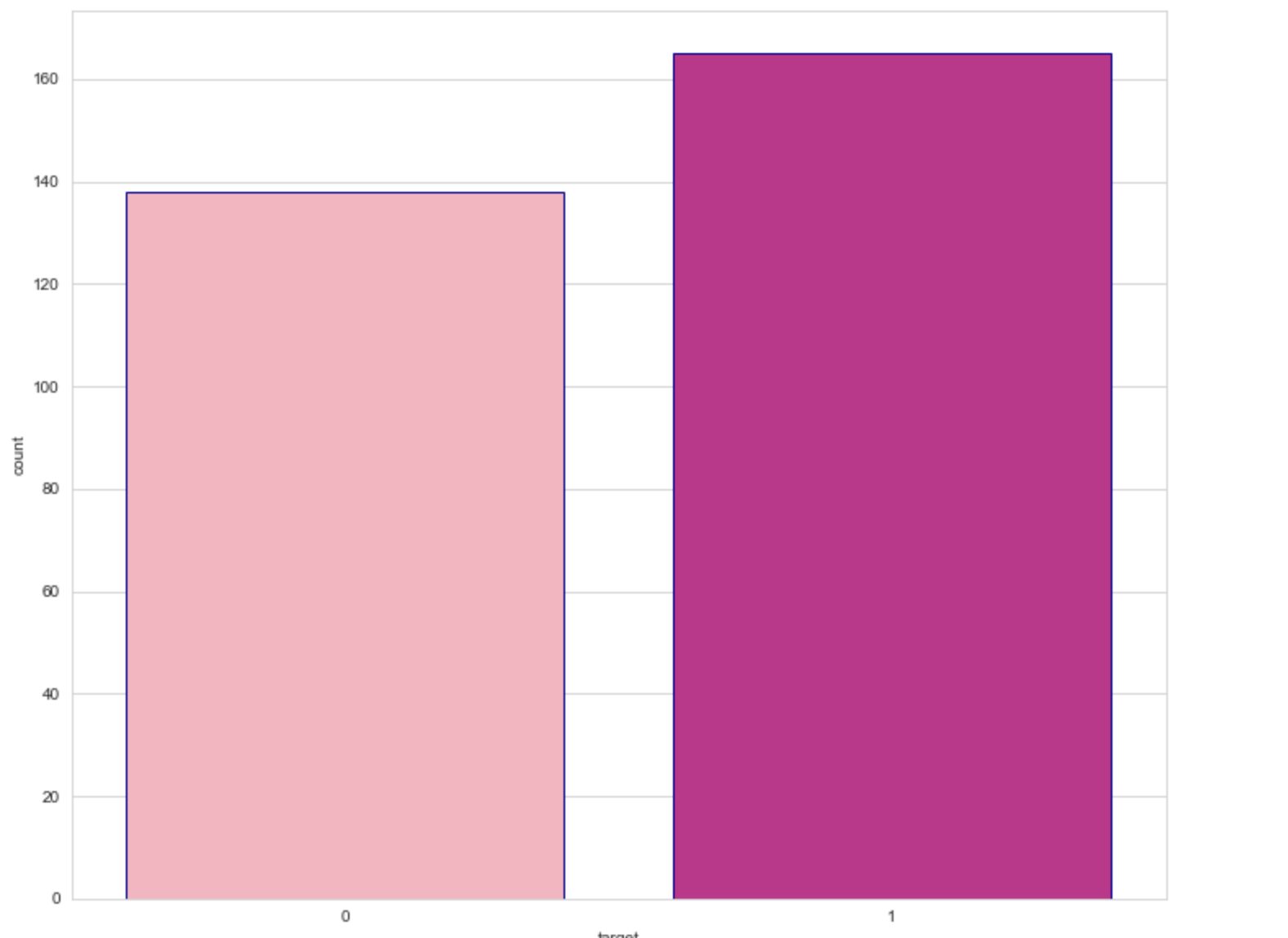
```
In [27]: import seaborn as sns
# get correlations of each features in dataset
corrmat = df.corr()
top_corr_features = corrmat.index
plt.figure(figsize=(20,20))
# plot heat map
g=sns.heatmap(df[top_corr_features].corr(),annot=True,cmap="PuBu")
```



```
In [25]: df.hist(color="orange",edgecolor="red")
plt.rcParams['figure.figsize']=(12,10)
```



```
In [28]: sns.set_style('whitegrid')
sns.countplot(x='target',data=df,palette='RdPu',edgecolor="darkblue")
plt.rcParams['figure.figsize']=(5,3)
```



Data Processing

```
In [12]: dataset = pd.get_dummies(df, columns = ['sex', 'cp', 'fbs', 'restecg', 'exang', 'slope', 'ca', 'thal'])
```

```
In [13]: from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
standardScaler = StandardScaler()
columns_to_scale = ['age', 'trestbps', 'chol', 'thalach', 'oldpeak']
dataset[columns_to_scale] = standardScaler.fit_transform(dataset[columns_to_scale])
```

```
In [14]: dataset.head()
```

Out[14]:

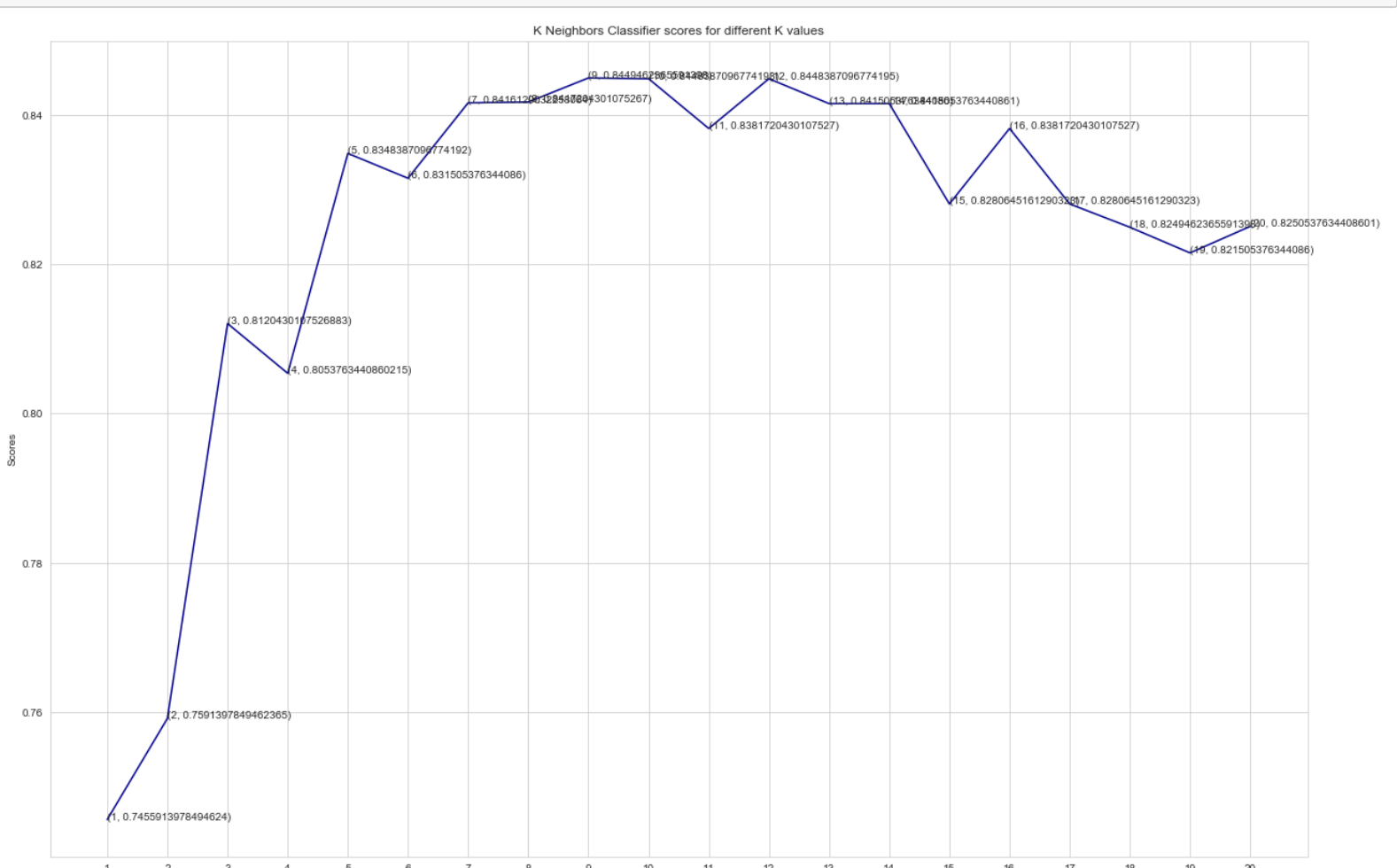
	age	trestbps	chol	thalach	oldpeak	target	sex_0	sex_1	cp_0	cp_1	...	slope_0	ca_0	ca_1	ca_2	ca
0	0.952197	0.763956	-0.256334	0.015443	1.087338	1	0	1	0	0	...	0	1	0	0	0
1	-1.915313	-0.092738	0.072199	1.633471	2.122573	1	0	1	0	0	...	0	1	0	0	0
2	-1.474158	-0.092738	-0.816773	0.977514	0.310912	1	1	0	0	1	...	1	1	0	0	0
3	0.180175	-0.663867	-0.198357	1.239897	-0.206705	1	0	1	0	1	...	1	1	0	0	0
4	0.290464	-0.663867	2.082050	0.583939	-0.379244	1	1	0	1	0	...	1	1	0	0	0

5 rows x 31 columns

```
In [15]: y = dataset['target']
X = dataset.drop(['target'], axis = 1)
```

```
In [16]: from sklearn.model_selection import cross_val_score
knn_scores = []
for k in range(1,21):
    knn_classifier = KNeighborsClassifier(n_neighbors = k)
    score=cross_val_score(knn_classifier,X,y,cv=10)
    knn_scores.append(score.mean())
```

```
In [18]: plt.plot([k for k in range(1,21)], knn_scores, color = 'darkblue')
for i in range(1,21):
    plt.text(i, knn_scores[i-1], (i, knn_scores[i-1]))
plt.xticks([i for i in range(1,21)])
plt.xlabel('Number of Neighbors (K)')
plt.ylabel('Score')
plt.title('K Neighbors Classifier scores for different K values')
plt.rcParams['figure.figsize']=(21,14)
```



```
In [19]: knn_classifier = KNeighborsClassifier(n_neighbors = 12)
score=cross_val_score(knn_classifier,X,y,cv=10)
```

```
In [20]: score.mean()
```

Out[20]: 0.8448387096774195

Random Forest Classifier

```
In [21]: from sklearn.ensemble import RandomForestClassifier
```

```
In [22]: randomforest_classifier= RandomForestClassifier(n_estimators=10)
score=cross_val_score(randomforest_classifier,X,y,cv=10)
```

```
In [23]: score.mean()
```

Out [23]: 0.811505376344086

```
In [ ]:
```