

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

df = pd.read_csv('heart.csv')

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

df.describe()

      age      sex      cp      trestbps      chol
fbs \
count  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
std    9.082101  0.466011  1.032052  17.538143  51.830751
0.356198
min    29.000000  0.000000  0.000000  94.000000  126.000000
0.000000
25%    47.500000  0.000000  0.000000  120.000000  211.000000
0.000000
50%    55.000000  1.000000  1.000000  130.000000  240.000000
0.000000
75%    61.000000  1.000000  2.000000  140.000000  274.500000
0.000000
max    77.000000  1.000000  3.000000  200.000000  564.000000

```

```
1.000000
```

```
      restecg    thalach     exang   oldpeak      slope
ca \
count 303.000000 303.000000 303.000000 303.000000 303.000000
303.000000
mean 0.528053 149.646865 0.326733 1.039604 1.399340
0.729373
std 0.525860 22.905161 0.469794 1.161075 0.616226
1.022606
min 0.000000 71.000000 0.000000 0.000000 0.000000
0.000000
25% 0.000000 133.500000 0.000000 0.000000 1.000000
0.000000
50% 1.000000 153.000000 0.000000 0.800000 1.000000
0.000000
75% 1.000000 166.000000 1.000000 1.600000 2.000000
1.000000
max 2.000000 202.000000 1.000000 6.200000 2.000000
4.000000
```

```
      thal      target
count 303.000000 303.000000
mean 2.313531 0.544554
std 0.612277 0.498835
min 0.000000 0.000000
25% 2.000000 0.000000
50% 2.000000 1.000000
75% 3.000000 1.000000
max 3.000000 1.000000
```

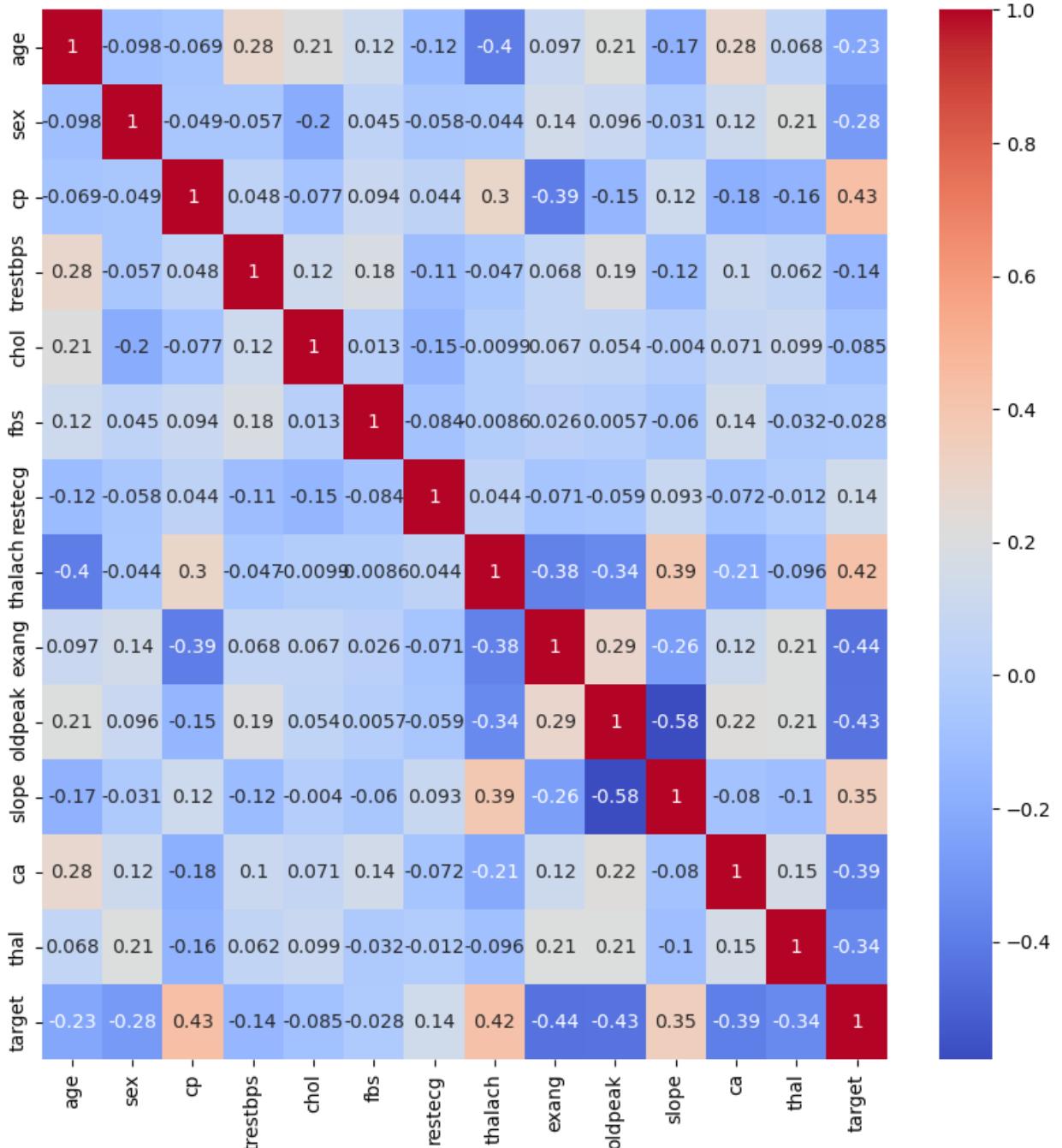
```
df.isnull().sum()
```

```
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
```

```
print(df.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

core = df.corr()
plt.figure(figsize =(10,10))
sns.heatmap(core,annot=True,cmap='coolwarm')
plt.show()
```



```

y = df["target"]
X = df.drop("target", axis=1)

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42
)

```

```

from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)

from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score

lr = LogisticRegression()
lr.fit(X_train, y_train)

y_pred = lr.predict(X_test)

print("Logistic Regression Accuracy:", accuracy_score(y_test, y_pred))
Logistic Regression Accuracy: 0.8524590163934426

from sklearn.ensemble import RandomForestClassifier
rf = RandomForestClassifier()
rf.fit(X_train, y_train)

y_pred_rf = rf.predict(X_test)

print("Random Forest Accuracy:", accuracy_score(y_test, y_pred_rf))
Random Forest Accuracy: 0.8524590163934426

from sklearn.metrics import classification_report, confusion_matrix

print(confusion_matrix(y_test, y_pred_rf))
print(classification_report(y_test, y_pred_rf))

[[24  5]
 [ 4 28]]
      precision    recall  f1-score   support
          0       0.86      0.83      0.84      29
          1       0.85      0.88      0.86      32
  accuracy                           0.85      61
  macro avg       0.85      0.85      0.85      61
weighted avg       0.85      0.85      0.85      61

```