# In [1]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

# In [2]:

```
df=pd.read_csv('Downloads/Heart_Disease_Prediction.csv')
```

# In [3]:

```
df.head()
```

# Out[3]:

	Age	Sex	Chest pain type	ВР	Cholesterol	FBS over 120	EKG results	Max HR	Exercise angina	ST depression	Slope of ST	Numb vesse flu
0	70	1	4	130	322	0	2	109	0	2.4	2	
1	67	0	3	115	564	0	2	160	0	1.6	2	
2	57	1	2	124	261	0	0	141	0	0.3	1	
3	64	1	4	128	263	0	0	105	1	0.2	2	
4	74	0	2	120	269	0	2	121	1	0.2	1	
4												<b>&gt;</b>

# In [4]:

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

# Out[4]:

Age	0
Sex	0
Chest pain type	0
BP	0
Cholesterol	0
FBS over 120	0
EKG results	0
Max HR	0
Exercise angina	0
ST depression	0
Slope of ST	0
Number of vessels fluro	0
Thallium	0
Heart Disease	0
dtype: int64	

# In [5]:

# print(df.info())

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 270 entries, 0 to 269
Data columns (total 14 columns):

#	Column	Non-Null Count	Dtype
0	Age	270 non-null	int64
1	Sex	270 non-null	int64
2	Chest pain type	270 non-null	int64
3	BP	270 non-null	int64
4	Cholesterol	270 non-null	int64
5	FBS over 120	270 non-null	int64
6	EKG results	270 non-null	int64
7	Max HR	270 non-null	int64
8	Exercise angina	270 non-null	int64
9	ST depression	270 non-null	float64
10	Slope of ST	270 non-null	int64
11	Number of vessels fluro	270 non-null	int64
12	Thallium	270 non-null	int64
13	Heart Disease	270 non-null	object
			-

dtypes: float64(1), int64(12), object(1)

memory usage: 29.7+ KB

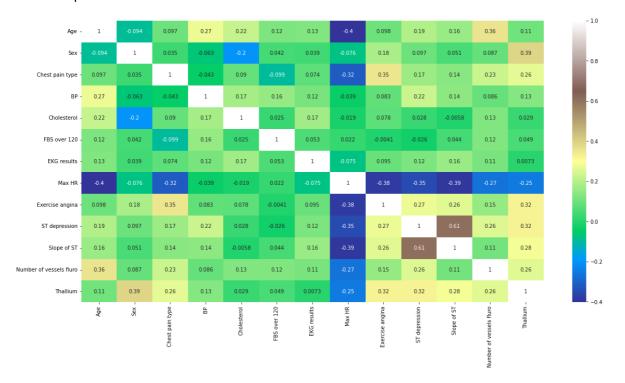
None

# In [6]:

```
plt.figure(figsize=(20,10))
sns.heatmap(df.corr(), annot=True, cmap='terrain')
```

# Out[6]:

# <AxesSubplot:>

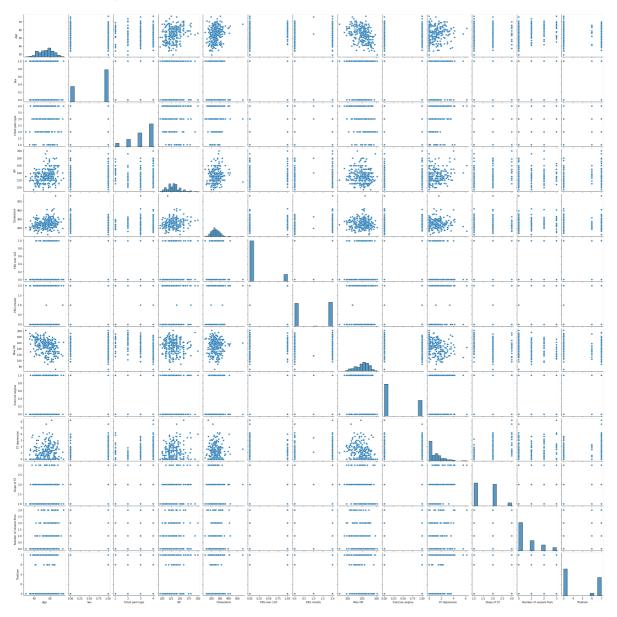


# In [7]:

sns.pairplot(data=df)

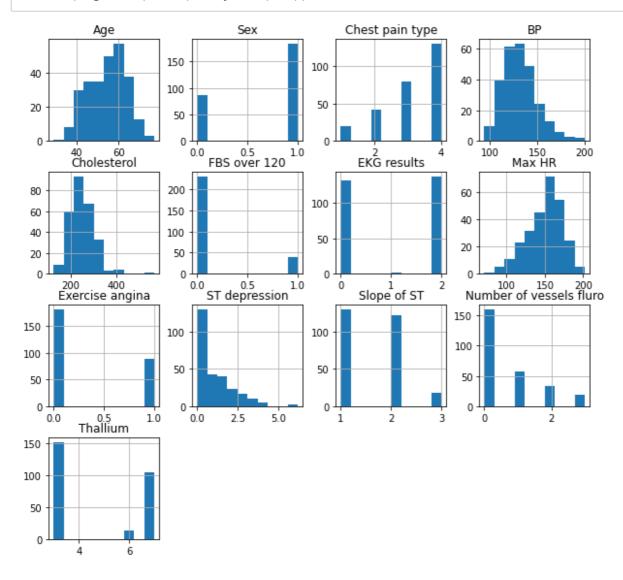
# Out[7]:

<seaborn.axisgrid.PairGrid at 0x1bb678b6308>



# In [8]:

# df.hist(figsize=(10,12), layout=(5,4));



# In [9]:

```
df.plot(kind='box', subplots=True, layout=(6,3), figsize=(10,10))
plt.show()
                                  1.0
  60
                                  0.5
                                                                    2
  40
                                  0.0
                                                                             Chest pain type
 200
                                                                  1.0
                                 400
 150
                                                                  0.5
                                 200
 100
                                                                  0.0
                  ВP
                                               Cholestero
                                                                              FBS over 120
                                                                  1.0
   2
                                 200
                                 150
                                                                  0.5
   1
                                 100
                                                                  0.0
   0
              EKG results
                                                Max HR
                                                                             Exercise angina
                                    3
  5.0
                                                                    2
                                    2
```

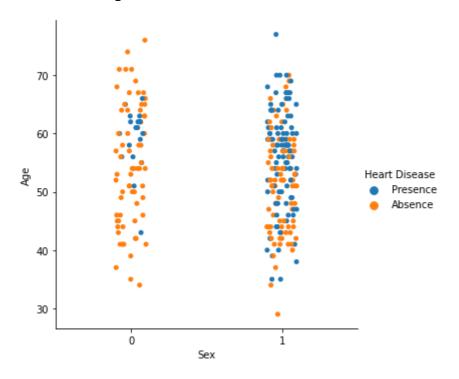
# In [10]:

2.5

```
sns.catplot(data=df, x='Sex', y='Age', hue='Heart Disease', palette='tab10')
```

# Out[10]:

<seaborn.axisgrid.FacetGrid at 0x1bb71a93fc8>

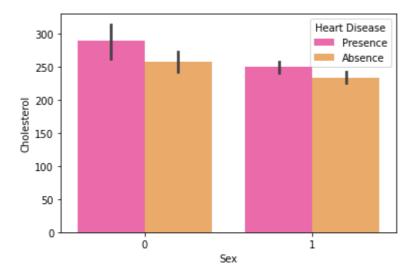


# In [11]:

```
sns.barplot(data=df, \ x='Sex', \ y='Cholesterol', \ hue='Heart \ Disease', \ palette='spring')
```

# Out[11]:

<AxesSubplot:xlabel='Sex', ylabel='Cholesterol'>



# In [12]:

```
df['Sex'].value_counts()
```

# Out[12]:

183
 87

Name: Sex, dtype: int64

# In [13]:

```
df['Chest pain type'].value_counts()
```

# Out[13]:

4 129

3 79

2 42

1 20

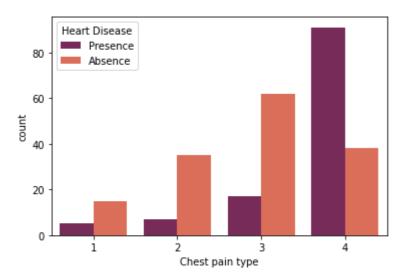
Name: Chest pain type, dtype: int64

# In [14]:

```
sns.countplot(x='Chest pain type', hue='Heart Disease' , data=df, palette='rocket')
```

# Out[14]:

<AxesSubplot:xlabel='Chest pain type', ylabel='count'>



# In [15]:

gen = pd.crosstab(df['Sex'], df['Heart Disease'])
print(gen)

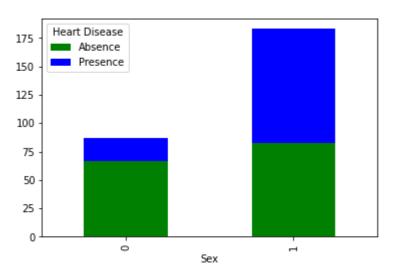
Heart	Disease	Absence	Presence
Sex			
0		67	20
1		83	100

### In [16]:

```
gen.plot(kind='bar', stacked='True', color=['green','blue'],grid=False)
```

# Out[16]:

<AxesSubplot:xlabel='Sex'>



# In [17]:

```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
StandardScaler = StandardScaler()
columns_to_scale=['Age', 'EKG results', 'Cholesterol', 'Thallium', 'Number of vessels fluro
df[columns_to_scale] = StandardScaler.fit_transform(df[columns_to_scale])
```

# In [18]:

df.head()

# Out[18]:

	Age	Sex	Chest pain type	ВР	Cholesterol	FBS over 120	EKG results	Max HR	Exercise angina	ST depression	Slope of ST
0	1.712094	1	4	130	1.402212	0	0.981664	109	0	2.4	2
1	1.382140	0	3	115	6.093004	0	0.981664	160	0	1.6	2
2	0.282294	1	2	124	0.219823	0	-1.026285	141	0	0.3	1
3	1.052186	1	4	128	0.258589	0	-1.026285	105	1	0.2	2
4	2.152032	0	2	120	0.374890	0	0.981664	121	1	0.2	1
4											•

### In [19]:

```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
StandardScaler = StandardScaler()
columns_to_scale=['Age', 'EKG results', 'Cholesterol', 'Thallium', 'Number of vessels fluro
df[columns_to_scale] = StandardScaler.fit_transform(df[columns_to_scale])
```

# In [20]:

```
df.head()
```

#### Out[20]:

	Age	Sex	Chest pain type	ВР	Cholesterol	FBS over 120	EKG results	Max HR	Exercise angina	ST depression	Slope of ST
0	1.712094	1	4	130	1.402212	0	0.981664	109	0	2.4	2
1	1.382140	0	3	115	6.093004	0	0.981664	160	0	1.6	2
2	0.282294	1	2	124	0.219823	0	-1.026285	141	0	0.3	1
3	1.052186	1	4	128	0.258589	0	-1.026285	105	1	0.2	2
4	2.152032	0	2	120	0.374890	0	0.981664	121	1	0.2	1
4											•

### In [21]:

```
x=df.drop(['Heart Disease'], axis=1)
y=df['Heart Disease']
```

#### In [22]:

```
x_train, x_test, y_train, y_test=train_test_split(x,y,test_size=0.3, random_state=40)
```

#### In [23]:

```
print('x_train-', x_train.size)
print('x_test-', x_test.size)
print('y_train-', y_train.size)
print('x_test-', x_test.size)
```

```
x_train- 2457
x_test- 1053
y_train- 189
x_test- 1053
```

### In [24]:

```
from sklearn.linear_model import LogisticRegression
lr=LogisticRegression()
model1=lr.fit(x_train,y_train)
prediction1=model1.predict(x_test)
```

# In [25]:

```
from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,prediction1)
cm
```

#### Out[25]:

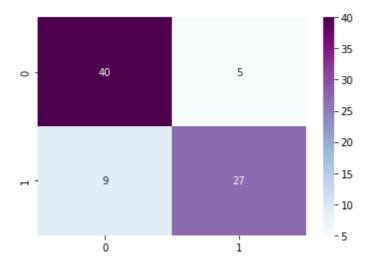
```
array([[40, 5], [ 9, 27]], dtype=int64)
```

# In [26]:

```
sns.heatmap(cm, annot=True,cmap='BuPu')
```

#### Out[26]:

#### <AxesSubplot:>



#### In [27]:

```
TP=cm[0][0]
TN=cm[1][1]
FN=cm[1][0]
FP=cm[0][1]
print('Testing Accuracy:', (TP+TN+FN)/(TP+TN+FP))
```

Testing Accuracy: 0.9382716049382716

# In [28]:

```
from sklearn.metrics import accuracy_score
accuracy_score(y_test,prediction1)
l=accuracy_score(y_test,prediction1)
```

#### In [29]:

```
from sklearn.metrics import classification_report
print(classification_report(y_test, prediction1))
```

	precision	recall	f1-score	support
Absence	0.82	0.89	0.85	45
Presence	0.84	0.75	0.79	36
accuracy			0.83	81
macro avg	0.83	0.82	0.82	81
weighted avg	0.83	0.83	0.83	81

# In [30]:

```
import pandas as pd
from sklearn import neighbors,metrics
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
import numpy as np
import pickle
from sklearn.ensemble import RandomForestClassifier
import pandas as pd
from sklearn.tree import DecisionTreeClassifier
import seaborn as sns
import matplotlib.pyplot as plt
```

# In [31]:

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

# In [32]:

```
dataset = pd.read_csv("Downloads/Heart_Disease_Prediction.csv")
```

# In [33]:

```
KX = dataset[['Age','Sex','Chest pain type','BP','Cholesterol','FBS over 120','EKG results'
```

# In [34]:

```
KY = dataset[['Heart Disease']].values
```

```
In [35]:
```

ΚX

```
Out[35]:
```

### In [36]:

```
KY = KY.flatten()
print(KY)
```

```
['Presence' 'Absence' 'Presence' 'Absence' 'Absence' 'Presence'
 'Presence' 'Presence' 'Absence' 'Absence' 'Absence' 'Presence'
 'Absence' 'Absence' 'Presence' 'Absence' 'Absence' 'Presence'
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 'Absence' 'Absence' 'Presence' 'Presence' 'Absence' 'Presence' 'Absence'
 'Absence' 'Absence' 'Presence']
```

```
In [37]:
```

```
KX_train , KX_test , KY_train , KY_test = train_test_split(KX,KY,test_size=0.2,random_state
```

```
In [38]:
```

```
knn = KNeighborsClassifier(n_neighbors = 20)
knn.fit(KX_train, KY_train)
print(knn.score(KX_test, KY_test))
```

#### 0.6111111111111111

### In [39]:

```
pickle.dump(knn,open('heart_knn_model.sav','wb'))
```

## In [40]:

```
predict_knn = knn.predict(KX_test)
accuracy_knn = metrics.accuracy_score(KY_test,predict_knn)
```

#### In [41]:

```
predict_knn
```

## Out[41]:

# In [42]:

```
accuracy_knn
```

# Out[42]:

#### 0.6111111111111112

#### In [43]:

```
k=accuracy knn
```

```
In [45]:
```

```
import csv
import pandas as pd
import numpy as np
from sklearn.naive_bayes import GaussianNB
from sklearn.model_selection import train_test_split
from sklearn import metrics
from sklearn.metrics import confusion_matrix, f1_score, roc_curve, auc
import matplotlib.pyplot as plt
from itertools import cycle
from scipy import interp
```

# In [46]:

```
df = pd.read_csv('Downloads/Heart_Disease_Prediction.csv', header = None)
```

#### In [47]:

```
training_x=df.iloc[1:df.shape[0],0:13]
```

#### In [48]:

```
training_y=df.iloc[1:df.shape[0],13:14]
```

#### In [49]:

```
nx=np.array(training_x)
ny=np.array(training_y)
```

#### In [52]:

```
for z in range(5):
    print("\nTest Train Split no. ",z+1,"\n")
    nx_train,nx_test,ny_train,ny_test = train_test_split(nx,ny,test_size=0.25,random_state=
# Gaussian function of sklearn
gnb = GaussianNB()
gnb.fit(nx_train, ny_train.ravel())
ny_pred = gnb.predict(nx_test)
```

```
Test Train Split no. 1
```

Test Train Split no. 2

Test Train Split no. 3

Test Train Split no. 4

Test Train Split no. 5

```
In [61]:
```

```
print("\n Naive Bayes model accuracy(in %):", metrics.accuracy_score(ny_test, ny_pred))
```

Naive Bayes model accuracy(in %): 0.7794117647058824

### In [62]:

```
n=metrics.accuracy_score(ny_test, ny_pred)
```

### In [64]:

```
import pandas as pd
from sklearn import neighbors,metrics
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
import numpy as np
import pickle
from sklearn.ensemble import RandomForestClassifier
import pandas as pd
from sklearn.tree import DecisionTreeClassifier
import seaborn as sns
import matplotlib.pyplot as plt
```

### In [65]:

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

#### In [67]:

```
dataset = pd.read_csv("Downloads/Heart_Disease_Prediction.csv")
```

### In [69]:

```
DX = dataset[['Age','Sex','Chest pain type','BP','Cholesterol','FBS over 120','EKG results
```

# In [70]:

```
dy = dataset[['Heart Disease']].values
```

# In [71]:

DX

#### Out[71]:

```
2.,
array([[70., 1., 4., ...,
                                 3., 3.],
             0.,
                           2.,
      [67.,
                  3., ...,
                                0.,
                                     7.],
             1.,
                           1.,
      [57.,
                 2., ...,
                                0.,
                                     7.],
      [56.,
             0., 2., ...,
                            2.,
                                0., 3.],
            1.,
                            2.,
      [57.,
                 4., ...,
                                0., 6.],
                           2.,
                                3.,
            1., 4., ...,
                                    3.]])
      [67.,
```

```
In [72]:
```

```
dy = dy.flatten()
print(dy)
```

```
['Presence' 'Absence' 'Presence' 'Absence' 'Absence' 'Presence'
 'Presence' 'Presence' 'Presence' 'Absence' 'Absence' 'Absence' 'Presence'
 'Absence' 'Absence' 'Presence' 'Presence' 'Absence' 'Absence' 'Presence'
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'Absence' 'Absence' 'Presence' 'Absence' 'Presence' 'Absence'
 'Absence' 'Absence' 'Presence']
```

#### In [73]:

```
DX_train , DX_test , dy_train , dy_test = train_test_split(DX,dy,test_size=0.2,random_state
```

# In [74]:

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

```
In [75]:
```

```
for x in range(200):
    dt = DecisionTreeClassifier(random_state=x)
    dt.fit(DX_train,dy_train)
    dy_pred_dt = dt.predict(DX_test)
    current_accuracy = round(accuracy_score(dy_pred_dt,dy_test)*100,2)
    if(current_accuracy>max_accuracy):
        max_accuracy = current_accuracy
        best_x = x
```

#### In [85]:

```
dt = DecisionTreeClassifier(random_state=best_x)
dt.fit(DX_train,dy_train)
dy_pred_dt = dt.predict(DX_test)
```

### In [88]:

```
score_dt = (accuracy_score(dy_pred_dt,dy_test))
```

#### In [89]:

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

The accuracy score achieved using Decision Tree is: 0.7962962962963

#### In [90]:

```
d=(accuracy_score(dy_pred_dt,dy_test))
```

#### In [91]:

```
print('Logistic Regression :',1)
print('KNN :',k)
print('Naive Bayes :',n)
print('Decision Tree :' ,d)
```

Logistic Regression : 0.8271604938271605

KNN: 0.6111111111111112

Naive Bayes : 0.7794117647058824 Decision Tree : 0.7962962962963

## In [93]:

```
print('Logistic Regression :',l*100,'%')
print('KNN :',k*100,'%')
print('Naive Bayes :',n*100,'%')
print('Decision Tree :' ,d*100,'%')
```

Logistic Regression: 82.71604938271605 %

KNN : 61.11111111111114 %

Naive Bayes : 77.94117647058823 % Decision Tree : 79.62962962963 %

In [ ]:		