

Meaningful Predictive Modeling

August 24, 2020

0.0.1 Dataset Description

Our dataset is from a study of heart disease that has been open to the public for many years. The study collects various measurements on patient health and cardiovascular statistics, and of course makes patient identities anonymous.

There are 14 columns in the dataset, where the `patient_id` column is a unique and random identifier. The remaining 13 features are described in the section below.

1. age
2. sex
3. chest pain type (4 values)
4. resting blood pressure
5. serum cholestoral in mg/dl
6. fasting blood sugar > 120 mg/dl
7. resting electrocardiographic results (values 0,1,2)
8. maximum heart rate achieved
9. exercise induced angina
10. oldpeak = ST depression induced by exercise relative to rest
11. the slope of the peak exercise ST segment
12. number of major vessels (0-3) colored by flourosopy
13. thal: 3 = normal; 6 = fixed defect; 7 = reversable defect

0.0.2 Attributes types

Real: 1,4,5,8,10,12

Ordered:11,

Binary: 2,6,9

Nominal:7,3,13

Data is provided courtesy of the Cleveland Heart Disease Database via the UCI Machine Learning repository.

Aha, D., and Dennis Kibler. "Instance-based prediction of heart-disease presence with the Cleveland database." University of California 3.1 (1988): 3-2.

0.0.3 Import Libraries

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels.api as sm
import datetime
import sklearn

from sklearn.linear_model import LinearRegression, LogisticRegression

from sklearn.model_selection import cross_val_score, train_test_split, \
    GridSearchCV, RandomizedSearchCV
from sklearn.preprocessing import LabelEncoder, StandardScaler, MinMaxScaler, \
    OneHotEncoder
from sklearn.metrics import confusion_matrix, classification_report, \
    mean_absolute_error, mean_squared_error, r2_score
from sklearn.metrics import plot_confusion_matrix, plot_precision_recall_curve, \
    plot_roc_curve, accuracy_score
from sklearn.metrics import auc, f1_score, precision_score, recall_score, \
    roc_auc_score

%matplotlib inline
sns.set_style('dark')
sns.set(font_scale=1.2)

import warnings
warnings.filterwarnings('ignore')

pd.set_option('display.max_columns',100)
#pd.set_option('display.max_rows',None)
```

```
[2]: df = pd.read_csv("heart.csv")
```

```
[3]: df
```

```
[3]:
```

	age	sex	chestpain	bloodpressure	serum	bloodsugar	\
0	70	1	4	130	322	0	
1	67	0	3	115	564	0	

2	57	1	2	124	261	0
3	64	1	4	128	263	0
4	74	0	2	120	269	0
..
265	52	1	3	172	199	1
266	44	1	2	120	263	0
267	56	0	2	140	294	0
268	57	1	4	140	192	0
269	67	1	4	160	286	0

	electrocardiographic	heartrate	angina	depression	slope	vessels	\
0	2	109	0	2.4	2	3	
1	2	160	0	1.6	2	0	
2	0	141	0	0.3	1	0	
3	0	105	1	0.2	2	1	
4	2	121	1	0.2	1	1	
..	
265	0	162	0	0.5	1	0	
266	0	173	0	0.0	1	0	
267	2	153	0	1.3	2	0	
268	0	148	0	0.4	2	0	
269	2	108	1	1.5	2	3	

	thal	disease
0	3	1
1	7	0
2	7	1
3	7	0
4	3	0
..
265	7	0
266	7	0
267	3	0
268	6	0
269	3	1

[270 rows x 14 columns]

0.0.4 Exploratory Data Analysis

```
[4]: 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   chestpain          270 non-null    int64
3   bloodpressure      270 non-null    int64
4   serum              270 non-null    int64
5   bloodsugar         270 non-null    int64
6   electrocardiographic 270 non-null    int64
7   heartrate          270 non-null    int64
8   angina             270 non-null    int64
9   depression         270 non-null    float64
10  slope              270 non-null    int64
11  vessels            270 non-null    int64
12  thal               270 non-null    int64
13  disease            270 non-null    int64
dtypes: float64(1), int64(13)
memory usage: 29.7 KB

```

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

```

[5]:
count    age                sex    chestpain    bloodpressure    serum  \
mean     54.433333    0.677778    3.174074    131.344444    249.659259
std       9.109067    0.468195    0.950090    17.861608    51.686237
min      29.000000    0.000000    1.000000    94.000000    126.000000
25%      48.000000    0.000000    3.000000    120.000000    213.000000
50%      55.000000    1.000000    3.000000    130.000000    245.000000
75%      61.000000    1.000000    4.000000    140.000000    280.000000
max      77.000000    1.000000    4.000000    200.000000    564.000000

count    bloodsugar    electrocardiographic    heartrate    angina    depression  \
mean      0.148148                1.022222    149.677778    0.329630    1.05000
std       0.355906                0.997891    23.165717    0.470952    1.14521
min       0.000000                0.000000    71.000000    0.000000    0.00000
25%       0.000000                0.000000    133.000000    0.000000    0.00000
50%       0.000000                2.000000    153.500000    0.000000    0.80000
75%       0.000000                2.000000    166.000000    1.000000    1.60000
max       1.000000                2.000000    202.000000    1.000000    6.20000

count    slope    vessels    thal    disease
mean     1.585185    0.670370    4.696296    0.444444
std      0.614390    0.943896    1.940659    0.497827
min      1.000000    0.000000    3.000000    0.000000
25%      1.000000    0.000000    3.000000    0.000000
50%      2.000000    0.000000    3.000000    0.000000
75%      2.000000    1.000000    7.000000    1.000000

```

```
max      3.000000    3.000000    7.000000    1.000000
```

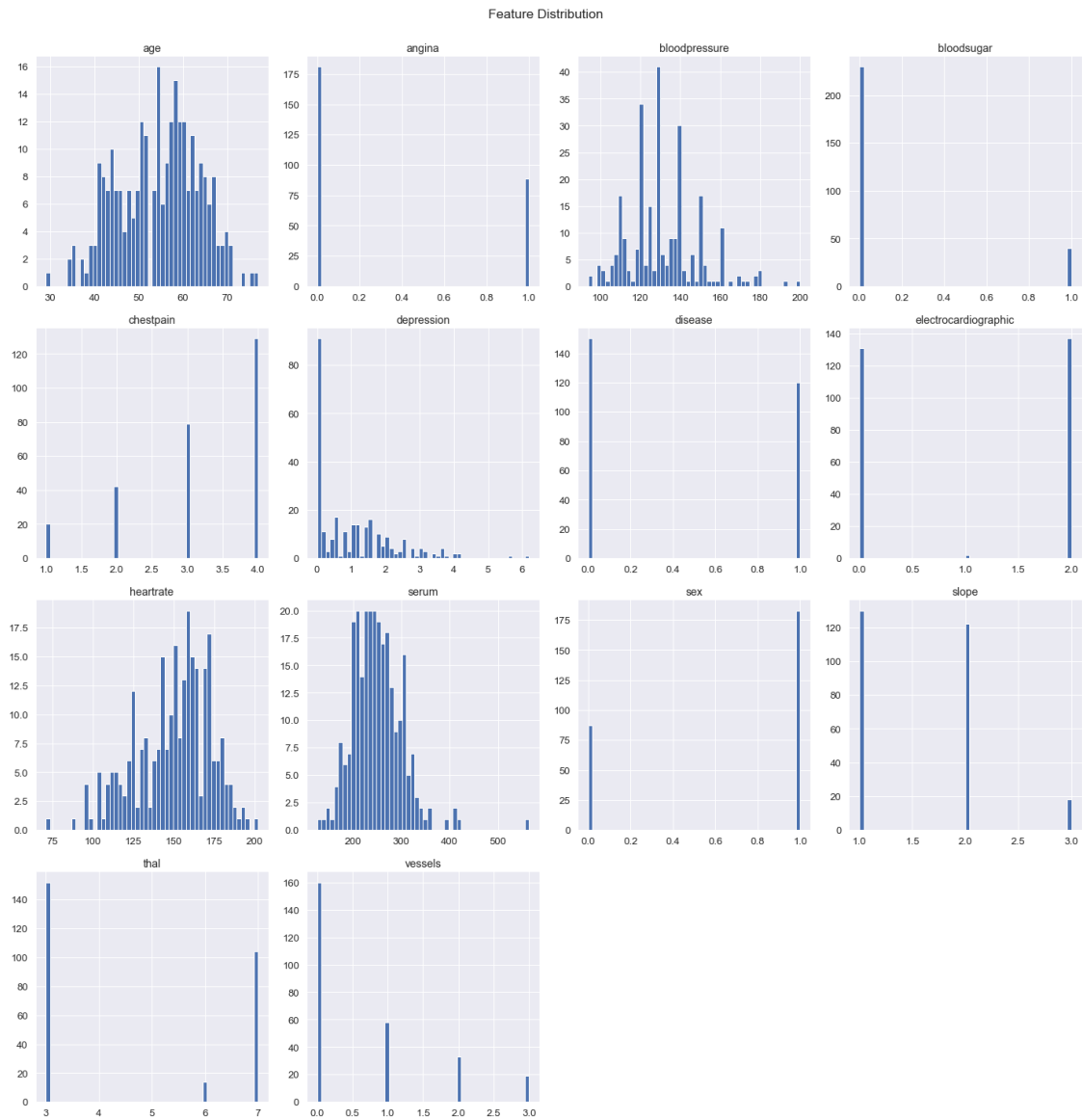
```
[6]: df.columns
```

```
[6]: Index(['age', 'sex', 'chestpain', 'bloodpressure', 'serum', 'bloodsugar',  
        'electrocardiographic', 'heartrate', 'angina', 'depression', 'slope',  
        'vessels', 'thal', 'disease'],  
        dtype='object')
```

0.0.5 Data Visualization

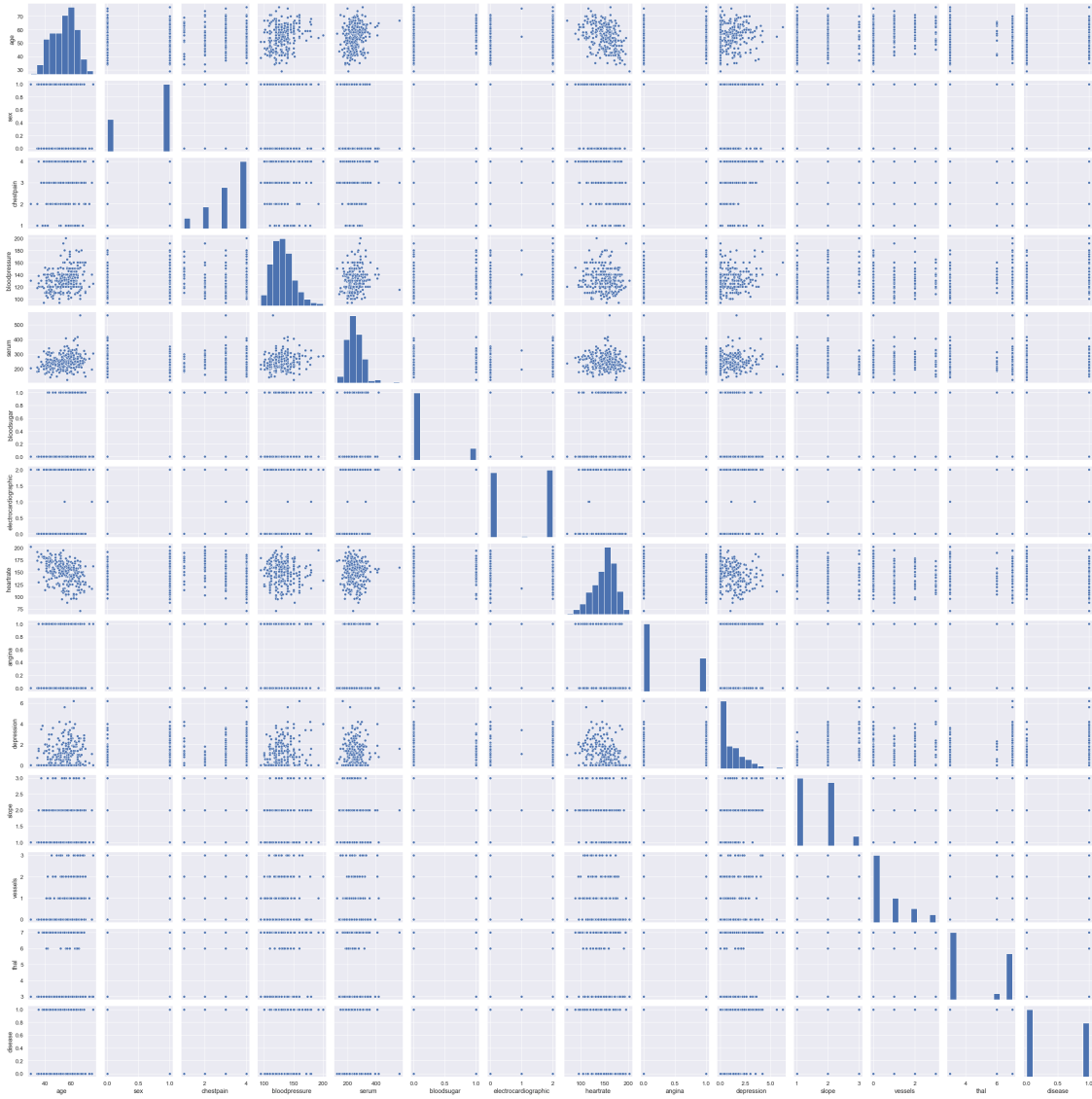
0.0.6 Univariate Data Exploration

```
[7]: df.hist(bins=50, figsize=(20,20))  
  
plt.suptitle('Feature Distribution', x=0.5, y=1.02, ha='center',  
            ↳fontsize='large')  
  
plt.tight_layout()  
  
plt.show();
```



```
[8]: plt.figure(figsize=(20,20))
plt.suptitle('Pairplots of features', x=0.5, y=1.02, ha='center',
            ↳fontsize='large')
sns.pairplot(df)
plt.show()
```

<Figure size 1440x1440 with 0 Axes>



0.0.7 Correlation

```
[9]: df.corr()
```

```
[9]:
```

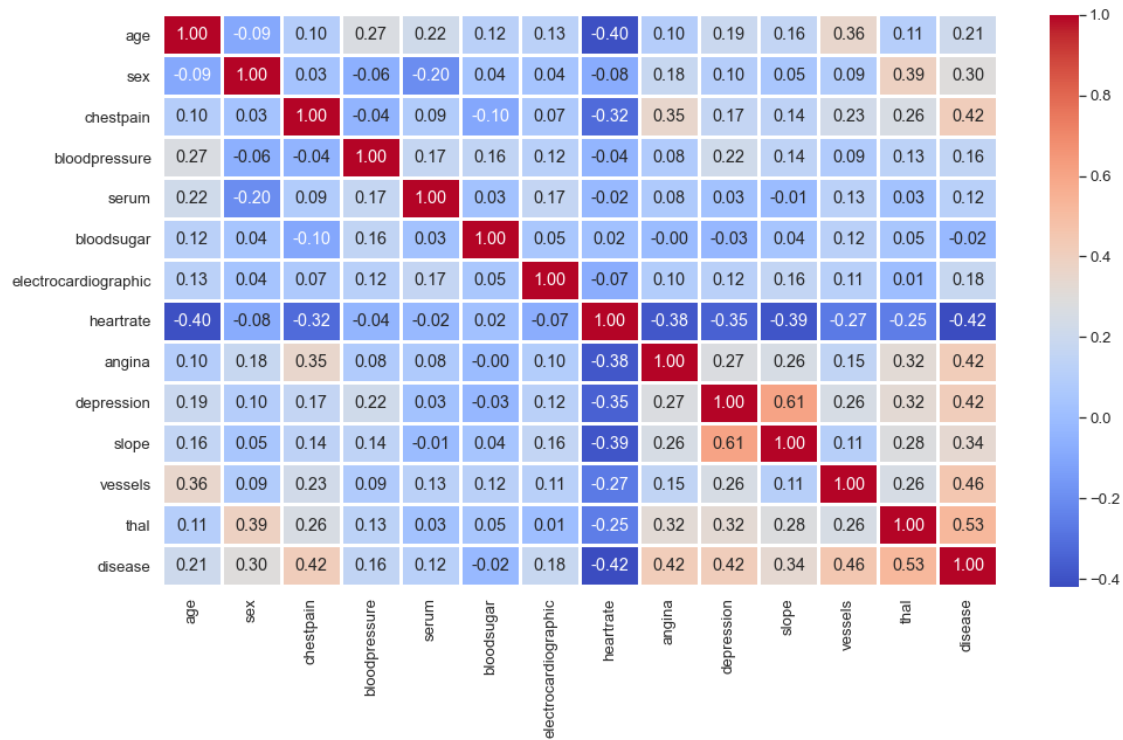
	age	sex	chestpain	bloodpressure	serum	\
age	1.000000	-0.094401	0.096920	0.273053	0.220056	
sex	-0.094401	1.000000	0.034636	-0.062693	-0.201647	
chestpain	0.096920	0.034636	1.000000	-0.043196	0.090465	
bloodpressure	0.273053	-0.062693	-0.043196	1.000000	0.173019	
serum	0.220056	-0.201647	0.090465	0.173019	1.000000	
bloodsugar	0.123458	0.042140	-0.098537	0.155681	0.025186	
electrocardiographic	0.128171	0.039253	0.074325	0.116157	0.167652	
heartrate	-0.402215	-0.076101	-0.317682	-0.039136	-0.018739	

angina	0.098297	0.180022	0.353160	0.082793	0.078243
depression	0.194234	0.097412	0.167244	0.222800	0.027709
slope	0.159774	0.050545	0.136900	0.142472	-0.005755
vessels	0.356081	0.086830	0.225890	0.085697	0.126541
thal	0.106100	0.391046	0.262659	0.132045	0.028836
disease	0.212322	0.297721	0.417436	0.155383	0.118021

	bloodsugar	electrocardiographic	heartrate	angina	\
age	0.123458		0.128171	-0.402215	0.098297
sex	0.042140		0.039253	-0.076101	0.180022
chestpain	-0.098537		0.074325	-0.317682	0.353160
bloodpressure	0.155681		0.116157	-0.039136	0.082793
serum	0.025186		0.167652	-0.018739	0.078243
bloodsugar	1.000000		0.053499	0.022494	-0.004107
electrocardiographic	0.053499		1.000000	-0.074628	0.095098
heartrate	0.022494		-0.074628	1.000000	-0.380719
angina	-0.004107		0.095098	-0.380719	1.000000
depression	-0.025538		0.120034	-0.349045	0.274672
slope	0.044076		0.160614	-0.386847	0.255908
vessels	0.123774		0.114368	-0.265333	0.153347
thal	0.049237		0.007337	-0.253397	0.321449
disease	-0.016319		0.182091	-0.418514	0.419303

	depression	slope	vessels	thal	disease
age	0.194234	0.159774	0.356081	0.106100	0.212322
sex	0.097412	0.050545	0.086830	0.391046	0.297721
chestpain	0.167244	0.136900	0.225890	0.262659	0.417436
bloodpressure	0.222800	0.142472	0.085697	0.132045	0.155383
serum	0.027709	-0.005755	0.126541	0.028836	0.118021
bloodsugar	-0.025538	0.044076	0.123774	0.049237	-0.016319
electrocardiographic	0.120034	0.160614	0.114368	0.007337	0.182091
heartrate	-0.349045	-0.386847	-0.265333	-0.253397	-0.418514
angina	0.274672	0.255908	0.153347	0.321449	0.419303
depression	1.000000	0.609712	0.255005	0.324333	0.417967
slope	0.609712	1.000000	0.109498	0.283678	0.337616
vessels	0.255005	0.109498	1.000000	0.255648	0.455336
thal	0.324333	0.283678	0.255648	1.000000	0.525020
disease	0.417967	0.337616	0.455336	0.525020	1.000000

```
[10]: plt.figure(figsize=(16,9))
sns.heatmap(df.corr(),cmap="coolwarm",annot=True,fmt='.2f',linewidths=2)
plt.show()
```

```
[ ]:
```

0.0.8 Treat Data Types

```
[11]: df.head()
```

```
[11]:
```

	age	sex	chestpain	bloodpressure	serum	bloodsugar	\
0	70	1	4	130	322	0	
1	67	0	3	115	564	0	
2	57	1	2	124	261	0	
3	64	1	4	128	263	0	
4	74	0	2	120	269	0	

	electrocardiographic	heartrate	angina	depression	slope	vessels	thal	\
0	2	109	0	2.4	2	3	3	
1	2	160	0	1.6	2	0	7	
2	0	141	0	0.3	1	0	7	
3	0	105	1	0.2	2	1	7	
4	2	121	1	0.2	1	1	3	

	disease
0	1
1	0

```
2      1
3      0
4      0
```

```
[12]: df.columns
```

```
[12]: Index(['age', 'sex', 'chestpain', 'bloodpressure', 'serum', 'bloodsugar',
            'electrocardiographic', 'heartrate', 'angina', 'depression', 'slope',
            'vessels', 'thal', 'disease'],
            dtype='object')
```

```
[13]: df[['sex', 'chestpain', 'bloodsugar', 'electrocardiographic', 'angina', 'slope',
            'thal']] =_
      ↪df[['sex', 'chestpain', 'bloodsugar', 'electrocardiographic', 'angina', 'slope',
            'thal']].astype("object")
```

```
[14]: 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   object
2   chestpain             270 non-null   object
3   bloodpressure          270 non-null   int64
4   serum                  270 non-null   int64
5   bloodsugar             270 non-null   object
6   electrocardiographic  270 non-null   object
7   heartrate             270 non-null   int64
8   angina                 270 non-null   object
9   depression            270 non-null   float64
10  slope                  270 non-null   object
11  vessels                270 non-null   int64
12  thal                   270 non-null   object
13  disease                270 non-null   int64
dtypes: float64(1), int64(6), object(7)
memory usage: 29.7+ KB
```

0.0.9 Feature Scaling/Transform

```
[15]: df2 = pd.get_dummies(df, drop_first=True)
```

```
[16]: df2
```

```
[16]:      age  bloodpressure  serum  heartrate  depression  vessels  disease  \
0      70             130    322      109        2.4        3        1
1      67             115    564      160        1.6        0        0
2      57             124    261      141        0.3        0        1
3      64             128    263      105        0.2        1        0
4      74             120    269      121        0.2        1        0
..      ...             ...    ...      ...        ...        ...
265    52             172    199      162        0.5        0        0
266    44             120    263      173        0.0        0        0
267    56             140    294      153        1.3        0        0
268    57             140    192      148        0.4        0        0
269    67             160    286      108        1.5        3        1
```

```
      sex_1  chestpain_2  chestpain_3  chestpain_4  bloodsugar_1  \
0         1           0           0           1           0
1         0           0           1           0           0
2         1           1           0           0           0
3         1           0           0           1           0
4         0           1           0           0           0
..      ...             ...             ...             ...
265      1           0           1           0           1
266      1           1           0           0           0
267      0           1           0           0           0
268      1           0           0           1           0
269      1           0           0           1           0
```

```
      electrocardiographic_1  electrocardiographic_2  angina_1  slope_2  \
0                           0                        1        0        1
1                           0                        1        0        1
2                           0                        0        0        0
3                           0                        0        1        1
4                           0                        1        1        0
..                          ...                      ...        ...
265                         0                        0        0        0
266                         0                        0        0        0
267                         0                        1        0        1
268                         0                        0        0        1
269                         0                        1        1        1
```

```
      slope_3  thal_6  thal_7
0           0       0       0
1           0       0       1
2           0       0       1
3           0       0       1
4           0       0       0
..          ...       ...       ...
265         0       0       1
```

266	0	0	1
267	0	0	0
268	0	1	0
269	0	0	0

[270 rows x 19 columns]

```
[17]: df2.columns
```

```
[17]: Index(['age', 'bloodpressure', 'serum', 'heartrate', 'depression', 'vessels',
          'disease', 'sex_1', 'chestpain_2', 'chestpain_3', 'chestpain_4',
          'bloodsugar_1', 'electrocardiographic_1', 'electrocardiographic_2',
          'angina_1', 'slope_2', 'slope_3', 'thal_6', 'thal_7'],
          dtype='object')
```

```
[18]: df2 = df2[['age', 'bloodpressure', 'serum', 'heartrate', 'depression',
                'vessels',
                'sex_1', 'chestpain_2', 'chestpain_3', 'chestpain_4',
                'bloodsugar_1', 'electrocardiographic_1', 'electrocardiographic_2',
                'angina_1', 'slope_2', 'slope_3', 'thal_6', 'thal_7', 'disease']]
```

```
[19]: df2
```

```
[19]:
```

	age	bloodpressure	serum	heartrate	depression	vessels	sex_1	\
0	70	130	322	109	2.4	3	1	
1	67	115	564	160	1.6	0	0	
2	57	124	261	141	0.3	0	1	
3	64	128	263	105	0.2	1	1	
4	74	120	269	121	0.2	1	0	
..	
265	52	172	199	162	0.5	0	1	
266	44	120	263	173	0.0	0	1	
267	56	140	294	153	1.3	0	0	
268	57	140	192	148	0.4	0	1	
269	67	160	286	108	1.5	3	1	

	chestpain_2	chestpain_3	chestpain_4	bloodsugar_1	\
0	0	0	1	0	
1	0	1	0	0	
2	1	0	0	0	
3	0	0	1	0	
4	1	0	0	0	
..	
265	0	1	0	1	
266	1	0	0	0	
267	1	0	0	0	
268	0	0	1	0	

269	0	0	1	0	
	electrocardiographic_1	electrocardiographic_2	angina_1	slope_2	\
0	0	1	0	1	
1	0	1	0	1	
2	0	0	0	0	
3	0	0	1	1	
4	0	1	1	0	
..	
265	0	0	0	0	
266	0	0	0	0	
267	0	1	0	1	
268	0	0	0	1	
269	0	1	1	1	

	slope_3	thal_6	thal_7	disease
0	0	0	0	1
1	0	0	1	0
2	0	0	1	1
3	0	0	1	0
4	0	0	0	0
..
265	0	0	1	0
266	0	0	1	0
267	0	0	0	0
268	0	1	0	0
269	0	0	0	1

[270 rows x 19 columns]

0.0.10 Create and save processed dataset

```
[20]: #df2.to_csv("heartrain.csv",index=False)
```

0.0.11 Train Test Split

```
[21]: df = pd.read_csv("heartrain.csv")
```

```
[22]: df.shape
```

```
[22]: (270, 19)
```

```
[23]: df.columns
```

```
[23]: Index(['age', 'bloodpressure', 'serum', 'heartrate', 'depression', 'vessels',  
        'sex_1', 'chestpain_2', 'chestpain_3', 'chestpain_4', 'bloodsugar_1',
```

```

    'electrocardiographic_1', 'electrocardiographic_2', 'angina_1',
    'slope_2', 'slope_3', 'thal_6', 'thal_7', 'disease'],
    dtype='object')

```

```

[24]: df = df[['age', 'serum', 'heartrate', 'depression', 'vessels',
    'sex_1', 'chestpain_2', 'chestpain_3', 'chestpain_4', 'bloodsugar_1',
    'electrocardiographic_1', 'electrocardiographic_2', 'angina_1',
    'slope_2', 'slope_3', 'thal_6', 'thal_7', 'disease', 'bloodpressure']]

```

```

[25]: df

```

```

[25]:
    age  serum  heartrate  depression  vessels  sex_1  chestpain_2 \
0    70    322      109        2.4        3      1          0
1    67    564      160        1.6        0      0          0
2    57    261      141        0.3        0      1          1
3    64    263      105        0.2        1      1          0
4    74    269      121        0.2        1      0          1
..    ..    ..      ..      ..      ..      ..      ..
265   52    199      162        0.5        0      1          0
266   44    263      173        0.0        0      1          1
267   56    294      153        1.3        0      0          1
268   57    192      148        0.4        0      1          0
269   67    286      108        1.5        3      1          0

    chestpain_3  chestpain_4  bloodsugar_1  electrocardiographic_1 \
0              0            1             0                      0
1              1            0             0                      0
2              0            0             0                      0
3              0            1             0                      0
4              0            0             0                      0
..            ...          ...          ...                    ...
265            1            0             1                      0
266            0            0             0                      0
267            0            0             0                      0
268            0            1             0                      0
269            0            1             0                      0

    electrocardiographic_2  angina_1  slope_2  slope_3  thal_6  thal_7 \
0                        1          0        1         0        0        0
1                        1          0        1         0        0        1
2                        0          0        0         0        0        1
3                        0          1        1         0        0        1
4                        1          1        0         0        0        0
..                      ...        ...      ...      ...      ...
265                      0          0        0         0        0        1
266                      0          0        0         0        0        1
267                      1          0        1         0        0        0

```

```

268          0          0          1          0          1          0
269          1          1          1          0          0          0

```

```

      disease  bloodpressure
0          1          130
1          0          115
2          1          124
3          0          128
4          0          120
..      ...      ...
265         0          172
266         0          120
267         0          140
268         0          140
269         1          160

```

```
[270 rows x 19 columns]
```

```
[26]: X = df.iloc[:,0:18]
      y = df.iloc[:,18]
```

```
[27]: X.values, y.values
```

```

[27]: (array([[ 70., 322., 109., ...,  0.,  0.,  1.],
               [ 67., 564., 160., ...,  0.,  1.,  0.],
               [ 57., 261., 141., ...,  0.,  1.,  1.],
               ...,
               [ 56., 294., 153., ...,  0.,  0.,  0.],
               [ 57., 192., 148., ...,  1.,  0.,  0.],
               [ 67., 286., 108., ...,  0.,  0.,  1.]]),
      array([130, 115, 124, 128, 120, 120, 130, 110, 140, 150, 135, 142, 140,
              134, 128, 112, 140, 140, 110, 140, 120, 130, 115, 112, 132, 130,
              138, 120, 112, 110, 128, 160, 120, 170, 144, 130, 140, 160, 130,
              122, 152, 124, 130, 101, 126, 140, 118, 110, 160, 150, 136, 128,
              140, 140, 130, 105, 138, 120, 174, 120, 150, 130, 120, 150, 145,
              150, 140, 136, 118, 108, 120, 120, 156, 140, 106, 142, 104,  94,
              120, 120, 146, 120, 150, 130, 110, 148, 128, 178, 126, 150, 140,
              130, 124, 110, 125, 110, 120, 100, 140, 120, 108, 120, 130, 165,
              130, 124, 100, 150, 140, 112, 180, 110, 158, 135, 120, 134, 120,
              200, 150, 130, 120, 122, 152, 160, 125, 160, 120, 136, 134, 117,
              108, 112, 140, 120, 150, 142, 152, 125, 118, 132, 145, 138, 140,
              125, 192, 123, 112, 110, 132, 112, 112, 120, 108, 130, 130, 105,
              140, 128, 120, 178, 120, 150, 130, 128, 110, 180, 110, 130, 138,
              138, 160, 140, 100, 120, 118, 138, 140, 150, 125, 129, 120, 134,
              110, 102, 130, 130, 132, 108, 140, 160, 140, 145, 108, 126, 124,
              135, 100, 110, 140, 125, 118, 125, 125, 140, 160, 152, 102, 105,
              125, 130, 170, 125, 122, 128, 130, 130, 135,  94, 120, 120, 110,

```

```
135, 150, 130, 138, 135, 130, 132, 150, 118, 145, 118, 115, 128,
130, 160, 138, 120, 138, 120, 180, 140, 130, 140, 140, 130, 110,
155, 140, 145, 120, 130, 112, 110, 150, 160, 150, 132, 140, 150,
120, 130, 120, 130, 110, 172, 120, 140, 140, 160], dtype=int64))
```

```
[28]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
↳ random_state=0)
```

```
[29]: X_train.shape, X_test.shape, y_train.shape, y_test.shape
```

```
[29]: ((216, 18), (54, 18), (216,), (54,))
```

0.0.12 Feature Scaling

```
[30]: X_train
```

```
[30]:
```

	age	serum	heartrate	depression	vessels	sex_1	chestpain_2	\
5	65	177	140	0.4	0	1	0	
22	43	303	181	1.2	0	1	0	
196	58	248	122	1.0	0	0	0	
110	55	327	117	3.4	0	0	0	
12	44	235	180	0.0	0	1	0	
..	
251	44	219	188	0.0	0	1	1	
192	54	309	156	0.0	0	1	1	
117	56	288	133	4.0	2	0	0	
47	44	197	177	0.0	1	1	0	
172	67	299	125	0.9	2	1	0	

	chestpain_3	chestpain_4	bloodsugar_1	electrocardiographic_1	\
5	0	1	0	0	
22	0	1	0	0	
196	0	1	0	0	
110	0	1	0	1	
12	1	0	0	0	
..	
251	0	0	0	0	
192	0	0	0	0	
117	0	1	1	0	
47	0	1	0	0	
172	0	1	0	0	

	electrocardiographic_2	angina_1	slope_2	slope_3	thal_6	thal_7	\
5	0	0	0	0	0	1	
22	0	0	1	0	0	0	
196	1	0	1	0	0	0	
110	0	1	1	0	0	0	


```

12          1          0          0          0          0          0
..          ...          ...          ...          ...          ...
251         1          0          0          0          0          0
192         0          0          0          0          0          1
117         1          1          0          1          0          1
47          1          0          0          0          0          0
172         1          1          1          0          0          0

```

```

disease
5          0
22         0
196        0
110        1
12         0
..         ...
251        0
192        0
117        1
47         1
172        1

```

[216 rows x 18 columns]

```
[31]: minmax = MinMaxScaler()
```

```
[32]: X_train_scaled = minmax.fit_transform(X_train)
```

```
[33]: X_test_scaled = minmax.transform(X_test)
```

```
[34]: X_train_scaled
```

```

[34]: array([[0.76595745, 0.11643836, 0.45614035, ..., 0.          , 1.          ,
              0.          ],
              [0.29787234, 0.40410959, 0.81578947, ..., 0.          , 0.          ,
              0.          ],
              [0.61702128, 0.27853881, 0.29824561, ..., 0.          , 0.          ,
              0.          ],
              ...,
              [0.57446809, 0.36986301, 0.39473684, ..., 0.          , 1.          ,
              1.          ],
              [0.31914894, 0.16210046, 0.78070175, ..., 0.          , 0.          ,
              1.          ],
              [0.80851064, 0.39497717, 0.3245614 , ..., 0.          , 0.          ,
              1.          ]])

```

```
[35]: X_test_scaled
```

```

[35]: array([[ 0.72340426,  0.24429224,  0.54385965,  0.37096774,  0.          ,
                1.          ,  0.          ,  0.          ,  0.          ,  1.          ,
                0.          ,  1.          ,  0.          ,  0.          ,  1.          ,
                1.          ,  0.          ,  0.          ],
 [ 0.36170213,  0.11643836,  0.63157895,  0.22580645,  0.          ,
                0.          ,  0.          ,  1.          ,  0.          ,  0.          ,
                0.          ,  1.          ,  1.          ,  0.          ,  1.          ,
                0.          ,  0.          ,  0.          ],
 [ 0.74468085,  0.40410959,  0.29824561,  0.32258065,  0.66666667,
                0.          ,  0.          ,  0.          ,  1.          ,  0.          ,
                0.          ,  0.          ,  0.          ,  1.          ,  0.          ,
                0.          ,  0.          ,  0.          ],
 [ 0.87234043,  0.32648402,  0.21052632,  0.46774194,  0.33333333,
                1.          ,  0.          ,  1.          ,  0.          ,  0.          ,
                0.          ,  0.          ,  1.          ,  1.          ,  0.          ,
                0.          ,  1.          ,  1.          ],
 [ 0.46808511,  0.11187215,  0.30701754,  0.09677419,  0.          ,
                1.          ,  0.          ,  1.          ,  0.          ,  0.          ,
                0.          ,  0.          ,  0.          ,  0.          ,  0.          ,
                0.          ,  0.          ,  0.          ],
 [ 0.85106383,  0.25799087,  0.55263158,  0.29032258,  0.66666667,
                0.          ,  0.          ,  0.          ,  0.          ,  0.          ,
                0.          ,  0.          ,  0.          ,  0.          ,  0.          ,
                0.          ,  0.          ,  0.          ],
 [ 0.53191489,  0.35844749,  0.93859649,  0.          ,  0.33333333,
                1.          ,  1.          ,  0.          ,  0.          ,  0.          ,
                0.          ,  1.          ,  0.          ,  0.          ,  0.          ,
                0.          ,  1.          ,  1.          ],
 [ 0.27659574,  0.26027397,  0.92982456,  0.12903226,  0.          ,
                1.          ,  0.          ,  1.          ,  0.          ,  1.          ,
                0.          ,  0.          ,  0.          ,  0.          ,  1.          ,
                0.          ,  1.          ,  0.          ],
 [ 0.68085106,  0.26712329,  0.42982456,  0.16129032,  0.          ,
                1.          ,  0.          ,  1.          ,  0.          ,  1.          ,
                0.          ,  0.          ,  1.          ,  1.          ,  0.          ,
                0.          ,  0.          ,  0.          ],
 [ 0.89361702,  0.05251142,  0.3245614 ,  0.25806452,  0.          ,
                0.          ,  0.          ,  0.          ,  1.          ,  0.          ,
                0.          ,  0.          ,  0.          ,  1.          ,  0.          ,
                0.          ,  0.          ,  0.          ],
 [ 0.46808511,  0.41552511,  0.47368421,  0.24193548,  0.33333333,
                0.          ,  0.          ,  1.          ,  0.          ,  0.          ,
                0.          ,  1.          ,  0.          ,  0.          ,  0.          ,
                0.          ,  0.          ,  0.          ],
 [ 0.59574468,  0.18493151,  0.70175439,  0.          ,  0.          ,
                1.          ,  0.          ,  0.          ,  1.          ,  0.          ,
                0.          ,  0.          ,  1.          ,  0.          ,  0.          ,
                0.          ,  0.          ,  1.          ]])

```

```

0.      , 1.      , 0.      ],
[ 0.53191489, 0.33561644, 0.56140351, 0.08064516, 0.33333333,
1.      , 0.      , 1.      , 0.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 1.      ,
0.      , 0.      , 0.      ],
[ 0.59574468, 0.37214612, 0.31578947, 0.16129032, 1.      ,
1.      , 0.      , 0.      , 1.      , 1.      ,
0.      , 1.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.65957447, 0.38127854, 0.71929825, 0.19354839, 0.66666667,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.44680851, 0.29223744, 0.62280702, 0.      , 0.      ,
0.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.61702128, 0.19406393, 0.6754386 , 0.      , 0.      ,
1.      , 0.      , 1.      , 0.      , 1.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 1.0212766 , 0.40639269, 0.64912281, 0.      , 1.      ,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 1.      , 0.      , 0.      ,
0.      , 0.      , 1.      ],
[ 0.70212766, 0.35388128, 0.13157895, 0.22580645, 0.33333333,
1.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 1.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.40425532, 0.23515982, 0.70175439, 0.16129032, 0.      ,
1.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 1.      ,
0.      , 1.      , 1.      ],
[ 0.46808511, 0.39269406, 0.29824561, 0.67741935, 1.      ,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 0.      , 1.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.25531915, 0.16438356, 0.70175439, 0.      , 0.33333333,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.63829787, 0.25799087, 0.47368421, 0.19354839, 0.33333333,
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0.      , 1.      , 1.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.34042553, 0.18721461, 0.52631579, 0.48387097, 0.      ,
1.      , 0.      , 0.      , 1.      , 0.      ,

```

```

0.      , 1.      , 1.      , 1.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.31914894, 0.37442922, 0.57017544, 0.      , 0.33333333,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 1.      ],
[ 0.25531915, 0.20091324, 0.70175439, 0.32258065, 0.      ,
1.      , 0.      , 1.      , 0.      , 0.      ,
0.      , 1.      , 0.      , 1.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.53191489, 0.25799087, 0.63157895, 0.19354839, 0.      ,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.80851064, 0.34474886, 0.73684211, 0.      , 0.33333333,
0.      , 0.      , 1.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.70212766, 0.61187215, 0.60526316, 0.19354839, 0.      ,
0.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 0.      , 1.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.80851064, 0.25342466, -0.14912281, 0.16129032, 0.      ,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 0.      , 1.      ],
[ 0.78723404, 0.40182648, 0.55263158, 0.06451613, 0.      ,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 0.      , 1.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.80851064, 0.22146119, 0.47368421, 0.0483871 , 0.66666667,
0.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.65957447, 0.1826484 , 0.38596491, 0.38709677, 0.66666667,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 1.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.25531915, 0.17808219, 0.73684211, 0.22580645, 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.29787234, 0.49086758, 0.42105263, 0.48387097, 0.      ,
0.      , 0.      , 0.      , 1.      , 1.      ,
0.      , 1.      , 1.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.70212766, 0.31278539, 0.07894737, 0.19354839, 0.33333333,

```

```

0.      , 0.      , 1.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.53191489, 0.3196347 , 0.18421053, 0.35483871, 0.33333333,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 1.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.78723404, 0.23287671, 0.6754386 , 0.16129032, 0.66666667,
0.      , 0.      , 0.      , 1.      , 1.      ,
0.      , 0.      , 1.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.31914894, 0.21461187, 0.71929825, 0.      , 0.      ,
1.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.25531915, 0.10502283, 0.61403509, 0.      , 0.      ,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.36170213, 0.28082192, 0.49122807, 0.12903226, 0.      ,
1.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.46808511, 0.38584475, 0.60526316, 0.09677419, 0.      ,
0.      , 0.      , 1.      , 0.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.46808511, 0.29680365, 0.53508772, 0.08064516, 0.      ,
0.      , 0.      , 1.      , 0.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.68085106, 0.41324201, 0.50877193, 0.16129032, 0.      ,
0.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 1.      , 1.      , 0.      ,
0.      , 1.      , 1.      ],
[ 0.5106383 , 0.24657534, 0.63157895, 0.      , 0.      ,
0.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[ 0.63829787, 0.33561644, 0.3245614 , 0.      , 0.      ,
1.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 1.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 1.      ],
[ 0.4893617 , 0.24429224, 0.51754386, 0.01612903, 1.      ,
1.      , 0.      , 0.      , 1.      , 1.      ,
0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 1.      , 0.      ],

```

```
[ 0.53191489, 0.24200913, 0.6754386 , 0.25806452, 0.      ,
 1.      , 0.      , 1.      , 0.      , 0.      ,
 0.      , 1.      , 0.      , 0.      , 0.      ,
 0.      , 1.      , 0.      ],
[ 0.80851064, 0.23515982, 0.35964912, 0.41935484, 0.66666667,
 1.      , 0.      , 0.      , 1.      , 0.      ,
 0.      , 1.      , 1.      , 1.      , 0.      ,
 0.      , 1.      , 1.      ],
[ 0.82978723, 0.33789954, 0.54385965, 0.25806452, 0.      ,
 1.      , 0.      , 1.      , 0.      , 1.      ,
 0.      , 1.      , 1.      , 1.      , 0.      ,
 0.      , 1.      , 1.      ],
[ 0.57446809, 0.64611872, 0.54385965, 0.30645161, 0.66666667,
 0.      , 0.      , 0.      , 1.      , 0.      ,
 0.      , 1.      , 1.      , 1.      , 0.      ,
 0.      , 1.      , 1.      ],
[ 0.61702128, 0.26027397, 0.57894737, 0.09677419, 0.      ,
 1.      , 0.      , 1.      , 0.      , 0.      ,
 0.      , 1.      , 1.      , 1.      , 0.      ,
 0.      , 1.      , 0.      ],
[ 0.65957447, 0.43835616, 0.63157895, 0.      , 0.33333333,
 0.      , 0.      , 1.      , 0.      , 0.      ,
 0.      , 0.      , 0.      , 0.      , 0.      ,
 0.      , 0.      , 0.      ],
[ 0.4893617 , 0.23744292, 0.63157895, 0.      , 0.33333333,
 1.      , 0.      , 0.      , 1.      , 0.      ,
 0.      , 0.      , 0.      , 0.      , 0.      ,
 0.      , 0.      , 1.      ]])
```

0.0.13 Section 1 : Regression (Predict Blood Pressure)

```
[36]: lr = LinearRegression()
```

```
[37]: model = lr.fit(X_train_scaled,y_train)
```

```
[38]: y_pred = model.predict(X_test_scaled)
```

```
[39]: y_pred[:10]
```

```
[39]: array([162.48588562, 136.12316588, 133.4370176 , 132.48577756,
 119.31611785, 145.88854269, 135.07170234, 138.83091643,
 125.09789119, 131.10441131])
```

```
[40]: y_test[:10]
```

```
[40]: 64      145
      135     142
      153     130
      189     160
      253     110
      198     140
      144     192
      180     120
      256     150
      15      112
      Name: bloodpressure, dtype: int64
```

0.0.14 Model Evaluation

```
[41]: mse = mean_squared_error(y_test,y_pred)
      mse
```

```
[41]: 355.08887178634967
```

```
[42]: rmse = np.sqrt(mse)
      rmse
```

```
[42]: 18.843801946166533
```

```
[43]: r2score = r2_score(y_test,y_pred)
      r2score
```

```
[43]: 0.12561622123336702
```

```
[44]: fig, ax = plt.subplots(figsize=(10,8))
      sns.regplot(x=y_test, y=y_pred, ax=ax)
      plt.title("Plot to compare actual vs predicted")
      plt.ylabel("Predicted")
      plt.xlabel("Actual")
      plt.show()
```



0.0.15 Cross-Validation

```
[45]: cv = cross_val_score(lr,X,y,cv=5,verbose=1,scoring='r2')
```

```
[Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.
[Parallel(n_jobs=1)]: Done 5 out of 5 | elapsed: 0.0s finished
```

```
[46]: cv.mean()
```

```
[46]: -0.002751760815410864
```

0.0.16 Section 2: Classification (Predict Heart Disease)

```
[47]: df = pd.read_csv("hearttrain.csv")
```

```
[48]: df
```



```

[48]:      age  bloodpressure  serum  heartrate  depression  vessels  sex_1  \
0       70             130    322      109        2.4        3        1
1       67             115    564      160        1.6        0        0
2       57             124    261      141        0.3        0        1
3       64             128    263      105        0.2        1        1
4       74             120    269      121        0.2        1        0
..      ...
265     52             172    199      162        0.5        0        1
266     44             120    263      173        0.0        0        1
267     56             140    294      153        1.3        0        0
268     57             140    192      148        0.4        0        1
269     67             160    286      108        1.5        3        1

      chestpain_2  chestpain_3  chestpain_4  bloodsugar_1  \
0                0            0            1            0
1                0            1            0            0
2                1            0            0            0
3                0            0            1            0
4                1            0            0            0
..              ...
265              0            1            0            1
266              1            0            0            0
267              1            0            0            0
268              0            0            1            0
269              0            0            1            0

      electrocardiographic_1  electrocardiographic_2  angina_1  slope_2  \
0                          0                        1          0          1
1                          0                        1          0          1
2                          0                        0          0          0
3                          0                        0          1          1
4                          0                        1          1          0
..                          ...
265                         0                        0          0          0
266                         0                        0          0          0
267                         0                        1          0          1
268                         0                        0          0          1
269                         0                        1          1          1

      slope_3  thal_6  thal_7  disease
0           0       0       0         1
1           0       0       1         0
2           0       0       1         1
3           0       0       1         0
4           0       0       0         0
..          ...
265         0       0       1         0

```

```

266      0      0      1      0
267      0      0      0      0
268      0      1      0      0
269      0      0      0      1

```

[270 rows x 19 columns]

```
[49]: X = df.iloc[:,0:18]
      y = df.iloc[:,18]
```

```
[50]: X.values, y.values
```

```
[50]: (array([[ 70., 130., 322., ...,  0.,  0.,  0.],
              [ 67., 115., 564., ...,  0.,  0.,  1.],
              [ 57., 124., 261., ...,  0.,  0.,  1.],
              ...,
              [ 56., 140., 294., ...,  0.,  0.,  0.],
              [ 57., 140., 192., ...,  0.,  1.,  0.],
              [ 67., 160., 286., ...,  0.,  0.,  0.])),
      array([1, 0, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0,
            0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 0,
            1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1,
            0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0,
            0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0,
            1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1,
            0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0,
            0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 1,
            1, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0,
            0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0,
            1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0,
            0, 1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0,
            1, 0, 0, 0, 0, 1], dtype=int64))
```

```
[51]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
      ↪random_state=0, stratify=y)
```

```
[52]: X_train.shape, X_test.shape, y_train.shape, y_test.shape
```

```
[52]: ((216, 18), (54, 18), (216,), (54,))
```

0.0.17 Feature Scaling

```
[53]: X_train
```

```
[53]:      age  bloodpressure  serum  heartrate  depression  vessels  sex_1  \
222   57             150    168         174           1.6         0        1
10    59             135    234         161           0.5         0        1
```

131	66	112	212	132	0.1	1	1
176	51	140	298	122	4.2	3	1
180	42	120	240	194	0.8	0	1
..
171	69	140	254	146	2.0	3	1
11	53	142	226	111	0.0	0	1
23	47	112	204	143	0.1	0	1
60	57	150	126	173	0.2	1	1
216	63	135	252	172	0.0	0	0

	chestpain_2	chestpain_3	chestpain_4	bloodsugar_1	\
222	0	1	0	0	
10	0	0	1	0	
131	0	0	1	0	
176	0	0	1	0	
180	0	1	0	1	
..	
171	0	1	0	0	
11	0	0	1	0	
23	0	0	1	0	
60	0	1	0	1	
216	0	1	0	0	

	electrocardiographic_1	electrocardiographic_2	angina_1	slope_2	\
222	0	0	0	0	
10	0	0	0	1	
131	0	1	1	0	
176	0	0	1	1	
180	0	0	0	0	
..	
171	0	1	0	1	
11	0	1	1	0	
23	0	0	0	0	
60	0	0	0	0	
216	0	1	0	0	

	slope_3	thal_6	thal_7
222	0	0	0
10	0	0	1
131	0	0	0
176	0	0	1
180	1	0	1
..
171	0	0	1
11	0	0	1
23	0	0	0
60	0	0	1

```
216      0      0      0
```

```
[216 rows x 18 columns]
```

```
[54]: minmax = MinMaxScaler()
```

```
[55]: X_train_scaled = minmax.fit_transform(X_train)
```

```
[56]: X_test_scaled = minmax.transform(X_test)
```

```
[57]: X_train_scaled
```

```
[57]: array([[0.58333333, 0.57142857, 0.09589041, ..., 0.        , 0.        ,
              0.        ],
              [0.625      , 0.41836735, 0.24657534, ..., 0.        , 0.        ,
              1.        ],
              [0.77083333, 0.18367347, 0.19634703, ..., 0.        , 0.        ,
              0.        ],
              ...,
              [0.375      , 0.18367347, 0.17808219, ..., 0.        , 0.        ,
              0.        ],
              [0.58333333, 0.57142857, 0.        , ..., 0.        , 0.        ,
              1.        ],
              [0.70833333, 0.41836735, 0.28767123, ..., 0.        , 0.        ,
              0.        ]])
```

```
[58]: X_test_scaled
```

```
[58]: array([[0.1875      , 0.26530612, 0.23972603, 0.84732824, 0.61290323,
              0.        , 1.        , 0.        , 0.        , 0.        ,
              0.        , 0.        , 0.        , 1.        , 1.        ,
              0.        , 0.        , 1.        ],
              [0.47916667, 0.40816327, 0.17123288, 0.66412214, 0.12903226,
              0.33333333, 1.        , 1.        , 0.        , 0.        ,
              0.        , 0.        , 0.        , 0.        , 0.        ,
              0.        , 0.        , 0.        ],
              [0.29166667, 0.28571429, 0.19863014, 0.71755725, 0.03225806,
              0.        , 0.        , 0.        , 1.        , 0.        ,
              0.        , 0.        , 0.        , 0.        , 1.        ,
              0.        , 0.        , 0.        ],
              [0.79166667, 0.06122449, 0.39497717, 0.41221374, 0.14516129,
              0.66666667, 1.        , 0.        , 0.        , 1.        ,
              0.        , 0.        , 1.        , 1.        , 1.        ,
              0.        , 0.        , 0.        ],
              [0.52083333, 0.26530612, 0.30136986, 0.58015267, 0.06451613,
              0.        , 1.        , 0.        , 1.        , 0.        ,
              0.        , 0.        , 1.        , 0.        , 1.        ,
              0.        , 0.        , 1.        ]])
```

```

0.      , 0.      , 1.      ],
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0.      , 0.      , 1.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
[0.41666667, 0.36734694, 0.3196347 , 0.76335878, 0.09677419,
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0.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
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1.      , 0.      , 0.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
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0.      , 0.      , 1.      , 0.      , 1.      ,
0.      , 0.      , 0.      ],
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1.      , 0.      , 0.      ],
[0.52083333, 0.14285714, 0.32191781, 0.73282443, 0.      ,
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0.      , 0.      , 1.      , 0.      , 0.      ,
0.      , 0.      , 0.      ],
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0.      , 0.      , 1.      , 0.      , 1.      ,
0.      , 0.      , 1.      ],
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0.33333333, 1.      , 0.      , 0.      , 1.      ,
0.      , 0.      , 0.      , 1.      , 0.      ,
0.      , 0.      , 1.      ],
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0.      , 0.      , 0.      ],
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1.      , 0.      , 1.      , 1.      , 0.      ,
0.      , 0.      , 1.      ],
[0.64583333, 0.46938776, 0.38127854, 0.75572519, 0.19354839,
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```

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```

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0.      , 0.      , 1.      ],
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0.      , 0.      , 0.      ],
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1.      , 0.      , 0.      ],
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0.      , 0.      , 1.      , 0.      , 1.      ,
0.      , 0.      , 0.      ],
[0.6875   , 0.36734694, 0.23972603, 0.57251908, 0.29032258,
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0.      , 0.      , 0.      , 0.      , 1.      ,
0.      , 0.      , 1.      ],

```

[0.66666667, 0.44897959, 0.0913242 , 0.41221374, 0.58064516,
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 0. , 0. , 0.],
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 1. , 0. , 1. , 0. , 0. ,
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 0. , 0. , 0. , 1. , 0. ,


```

1.          , 0.          , 1.          ],
[0.60416667, 0.34693878, 0.20547945, 0.45801527, 0.35483871,
1.          , 1.          , 0.          , 0.          , 1.          ,
0.          , 0.          , 1.          , 1.          , 1.          ,
0.          , 0.          , 1.          ],
[0.52083333, 0.26530612, 0.14155251, 0.32061069, 0.22580645,
0.33333333, 1.          , 0.          , 0.          , 1.          ,
0.          , 0.          , 0.          , 0.          , 1.          ,
0.          , 0.          , 1.          ]])

```

```
[59]: logic = LogisticRegression(random_state=0)
```

```
[60]: model2 = logic.fit(X_train_scaled,y_train)
```

```
[61]: y_pred2 = model2.predict(X_test_scaled)
```

You can do comparison with test data actual value with predicted data

```
[62]: y_pred2[:10]
```

```
[62]: array([1, 0, 0, 1, 1, 0, 0, 0, 0, 1], dtype=int64)
```

```
[63]: y_test[:10]
```

```
[63]: 160    1
      128    0
      212    0
      172    1
      218    0
       85    0
      263    0
      254    0
      183    0
      133    1
      Name: disease, dtype: int64
```

0.0.18 Cross-Validation

```
[64]: cv = cross_val_score(model2,X,y,cv=5,verbose=1,scoring='accuracy')
```

```
[Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.
[Parallel(n_jobs=1)]: Done   5 out of   5 | elapsed:   0.1s finished
```

```
[65]: cv.mean()
```

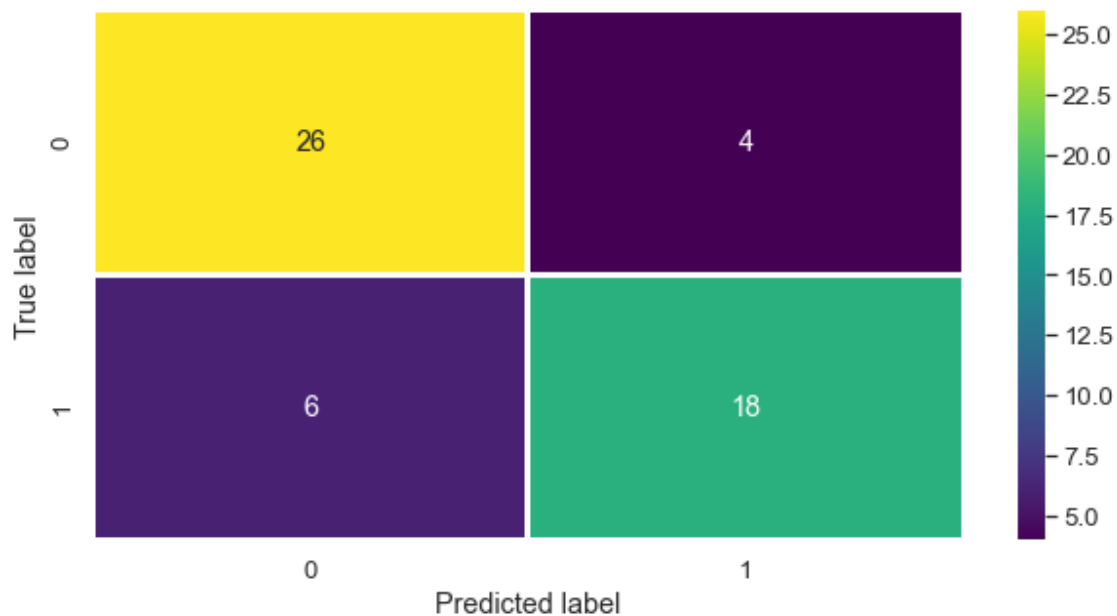
```
[65]: 0.8555555555555555
```

0.0.19 Model Evaluation

```
[66]: cm = confusion_matrix(y_test,y_pred2)
      cm
```

```
[66]: array([[26,  4],
      [ 6, 18]], dtype=int64)
```

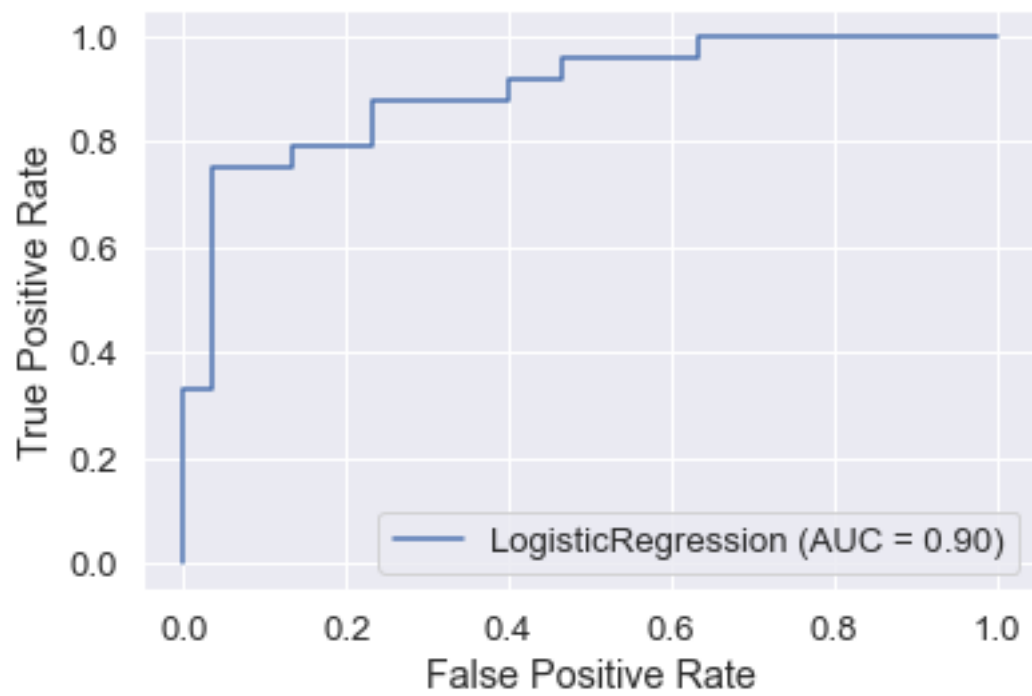
```
[67]: fig , ax = plt.subplots(figsize=(10,5))
      sns.heatmap(cm, annot=True,fmt='.4g',linewidths=2, cmap='viridis')
      plt.ylabel('True label')
      plt.xlabel('Predicted label')
      plt.show()
```



```
[68]: print(classification_report(y_test,y_pred2))
```

	precision	recall	f1-score	support
0	0.81	0.87	0.84	30
1	0.82	0.75	0.78	24
accuracy			0.81	54
macro avg	0.82	0.81	0.81	54
weighted avg	0.82	0.81	0.81	54

```
[69]: plot_roc_curve(model2,X_test_scaled,y_test)
plt.show()
```



```
[ ]:
```