

Load and explore the dataset

```
In [1]: import random
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```

```
In [2]: data = pd.read_csv("C:/Users/Administrator/Desktop/UST campus training/Dataset/Heart.csv")
data.head(10)
```

Out[2]:

| | age | sex | cp | trestbps | chol | fbs | restecg | thalach | exang | oldpeak | slope | ca | thal |
|---|-----|-----|----|----------|------|-----|---------|---------|-------|---------|-------|----|------|
| 0 | 69 | 1 | 0 | 160 | 234 | 1 | 2 | 131 | 0 | 0.1 | 1 | 1 | 0 |
| 1 | 69 | 0 | 0 | 140 | 239 | 0 | 0 | 151 | 0 | 1.8 | 0 | 2 | 0 |
| 2 | 66 | 0 | 0 | 150 | 226 | 0 | 0 | 114 | 0 | 2.6 | 2 | 0 | 0 |
| 3 | 65 | 1 | 0 | 138 | 282 | 1 | 2 | 174 | 0 | 1.4 | 1 | 1 | 0 |
| 4 | 64 | 1 | 0 | 110 | 211 | 0 | 2 | 144 | 1 | 1.8 | 1 | 0 | 0 |
| 5 | 64 | 1 | 0 | 170 | 227 | 0 | 2 | 155 | 0 | 0.6 | 1 | 0 | 2 |
| 6 | 63 | 1 | 0 | 145 | 233 | 1 | 2 | 150 | 0 | 2.3 | 2 | 0 | 1 |
| 7 | 61 | 1 | 0 | 134 | 234 | 0 | 0 | 145 | 0 | 2.6 | 1 | 2 | 0 |
| 8 | 60 | 0 | 0 | 150 | 240 | 0 | 0 | 171 | 0 | 0.9 | 0 | 0 | 0 |
| 9 | 59 | 1 | 0 | 178 | 270 | 0 | 2 | 145 | 0 | 4.2 | 2 | 0 | 2 |

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```
In [14]: data.columns
```

```
Out[14]: Index(['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thalach',
       'exang', 'oldpeak', 'slope', 'ca', 'thal', 'condition'],
      dtype='object')
```

```
In [3]: data.dtypes
```

```
Out[3]: age          int64
sex          int64
cp           int64
trestbps    int64
chol         int64
fbs          int64
restecg     int64
thalach     int64
exang        int64
oldpeak     float64
slope        int64
ca           int64
thal         int64
condition    int64
dtype: object
```

```
In [6]: data['condition'].value_counts()
```

```
Out[6]: condition
0    160
1    137
Name: count, dtype: int64
```

Calculating the number of missing values in the dataset

```
In [9]: data.isna().sum()
```

```
Out[9]: 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
condition 0
dtype: int64
```

Summary statistics

```
In [15]: data.describe()
```

| | age | sex | cp | trestbps | chol | fbs | restecg |
|--------------|------------|------------|------------|------------|------------|------------|------------|
| count | 297.000000 | 297.000000 | 297.000000 | 297.000000 | 297.000000 | 297.000000 | 297.000000 |
| mean | 54.542088 | 0.676768 | 2.158249 | 131.693603 | 247.350168 | 0.144781 | 0.996667 |
| std | 9.049736 | 0.468500 | 0.964859 | 17.762806 | 51.997583 | 0.352474 | 0.994900 |
| min | 29.000000 | 0.000000 | 0.000000 | 94.000000 | 126.000000 | 0.000000 | 0.000000 |
| 25% | 48.000000 | 0.000000 | 2.000000 | 120.000000 | 211.000000 | 0.000000 | 0.000000 |
| 50% | 56.000000 | 1.000000 | 2.000000 | 130.000000 | 243.000000 | 0.000000 | 1.000000 |
| 75% | 61.000000 | 1.000000 | 3.000000 | 140.000000 | 276.000000 | 0.000000 | 2.000000 |
| max | 77.000000 | 1.000000 | 3.000000 | 200.000000 | 564.000000 | 1.000000 | 2.000000 |

Gender Distribution Analysis

Count the number of males and females

```
In [26]: gender = data['sex'].value_counts()
gender.index = ['Male', 'Female']
gender
```

```
Out[26]: Male      201
          Female    96
          Name: count, dtype: int64
```

In [25]:

```
Out[25]: sex
1      201
0      96
Name: count, dtype: int64
```

In [42]:

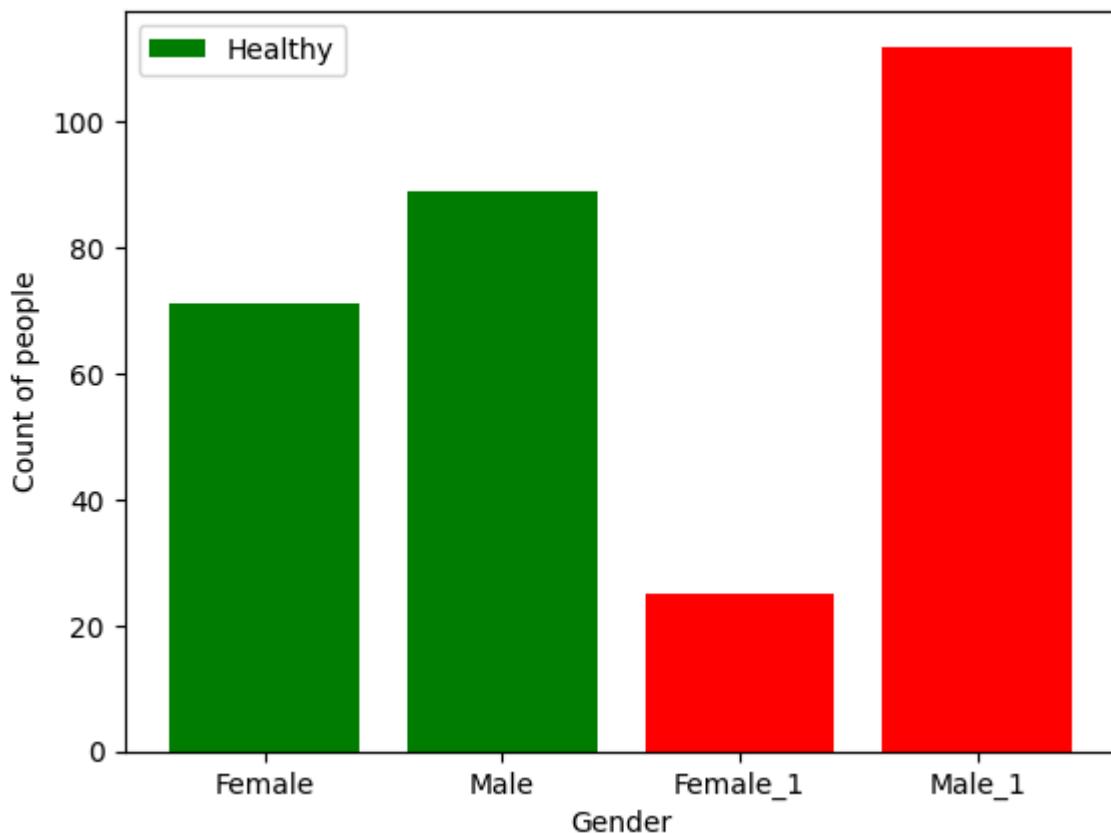
```
# Percentage distribution using numpy
gender_dev = data.groupby(['condition', 'sex'])['sex'].value_counts()
gender_dev
```

```
Out[42]: condition  sex
0            0      71
                 1      89
1            0      25
                 1     112
Name: count, dtype: int64
```

In [177]:

```
categories = ['Female', 'Male', 'Female_1', 'Male_1']
plt.bar(categories, gender_dev.values, color = ['green', 'green', 'red', 'red'])
plt.title('Gender wise distribution')
plt.xlabel('Gender')
plt.legend(['Healthy', 'Has disease'])
plt.ylabel('Count of people')
plt.show()
```

Gender wise distribution



Age analysis

```
In [50]: min_age = data['age'].min()
max_age = data['age'].max()
avg_age = data['age'].mean()
median_age = data['age'].median()
print(f'Minimum age in dataset is : {min_age}')
print(f'Maximum age in dataset is : {max_age}')
print(f'Average age in dataset is : {round(avg_age)}')
print(f'Median age in dataset is : {median_age}' )
```

Minimum age in dataset is : 29
 Maximum age in dataset is : 77
 Average age in dataset is : 55
 Median age in dataset is : 56.0

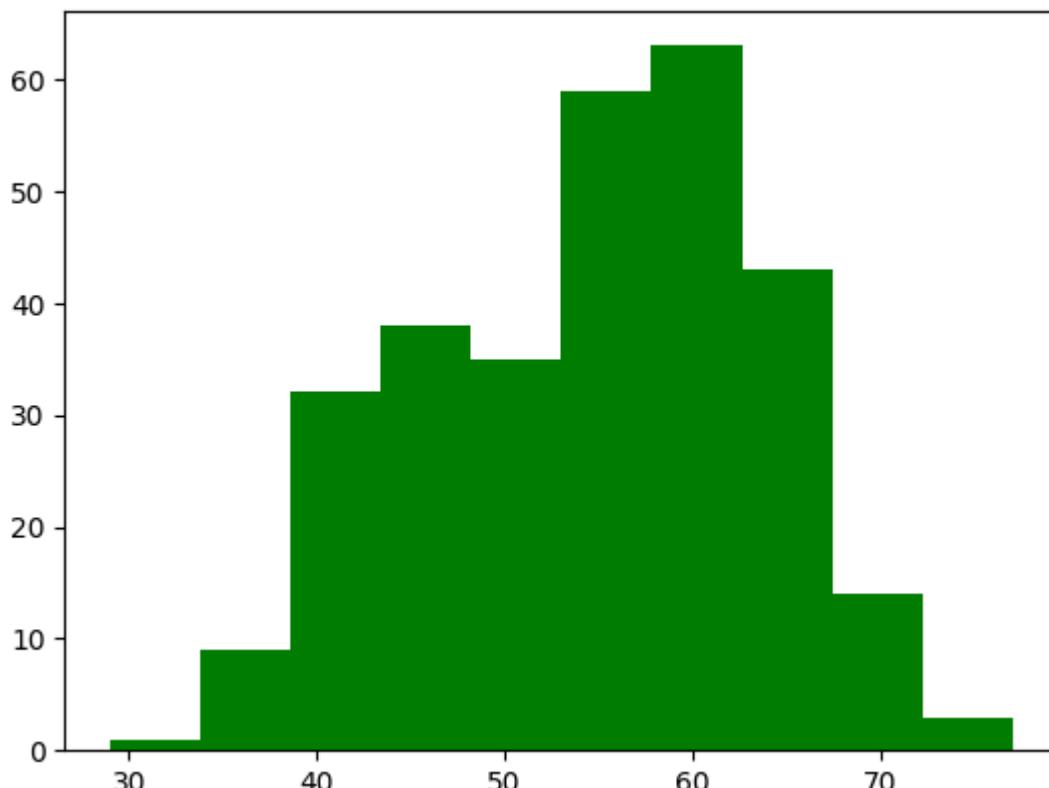
```
In [59]: min_age = data[data['condition']==1]['age'].min()
max_age = data[data['condition']==1]['age'].max()
avg_age = data[data['condition']==1]['age'].mean()
median_age = data[data['condition']==1]['age'].median()

print(f" Min age of patient with disease is {min_age}")
print(f" Maximum age of patient with disease is {max_age}")
print(f" Average age of patient with disease is {round(avg_age)}")
print(f" Median age of patient with disease is {median_age}")
```

Min age of patient with disease is 35
 Maximum age of patient with disease is 77
 Average age of patient with disease is 57
 Median age of patient with disease is 58.0

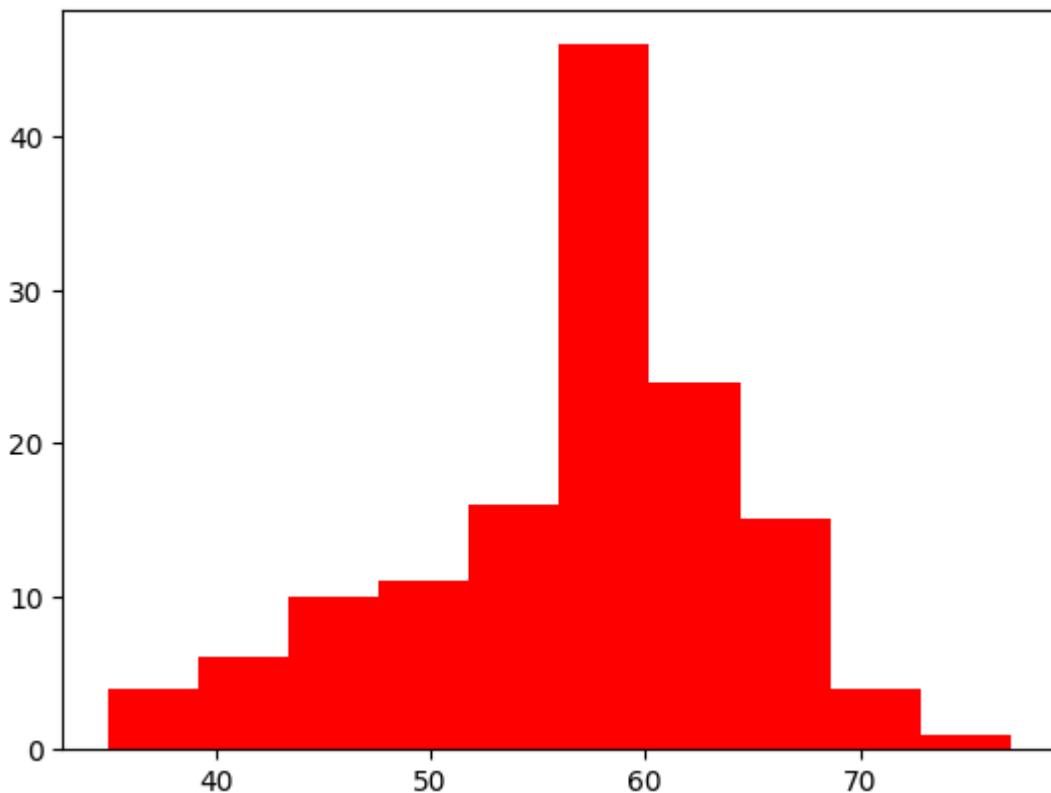
```
In [61]: plt.hist(data['age'], color='green', bins=10)
```

```
Out[61]: (array([ 1.,  9., 32., 38., 35., 59., 63., 43., 14.,  3.]),
 array([29. , 33.8, 38.6, 43.4, 48.2, 53. , 57.8, 62.6, 67.4, 72.2, 77. ]),
 <BarContainer object of 10 artists>)
```



```
In [62]: plt.hist(data[data['condition']==1]['age'], bins = 10, color = 'red')

Out[62]: (array([ 4.,  6., 10., 11., 16., 46., 24., 15.,  4.,  1.]),
array([35. , 39.2, 43.4, 47.6, 51.8, 56. , 60.2, 64.4, 68.6, 72.8, 77. ]),
<BarContainer object of 10 artists>)
```



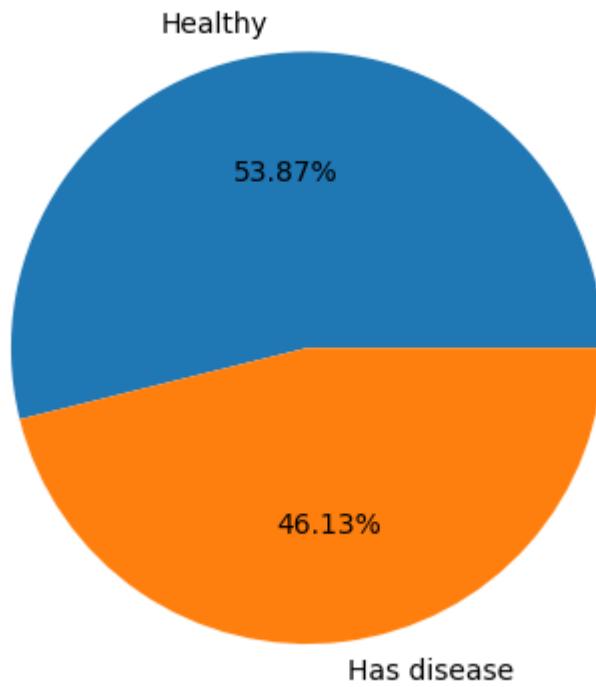
4. Target Variable Analysis

```
In [65]: patients_with_without = data['condition'].value_counts()
patients_with_without
```

```
Out[65]: condition
0    160
1    137
Name: count, dtype: int64
```

```
In [71]: plt.pie(patients_with_without, labels=['Healthy', 'Has disease'], autopct='%2.2f%%')
```

```
Out[71]: ([<matplotlib.patches.Wedge at 0x2da4cd87050>,
<matplotlib.patches.Wedge at 0x2da4cd86bd0>],
[Text(-0.13347885143430296, 1.0918715108563732, 'Healthy'),
Text(0.13347885143430332, -1.0918715108563732, 'Has disease')],
[Text(-0.07280664623689252, 0.5955662786489309, '53.87%'),
Text(0.07280664623689272, -0.5955662786489309, '46.13%')])
```



```
In [13]: disease_percentage = (len(data[data['condition']==1])/len(data))*100
disease_percentage
```

```
Out[13]: 46.12794612794613
```

```
In [79]: data.corr()
```

| | age | sex | cp | trestbps | chol | fbs | restecg |
|------------------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|
| age | 1.000000 | -0.092399 | 0.110471 | 0.290476 | 0.202644 | 0.132062 | 0.149917 |
| sex | -0.092399 | 1.000000 | 0.008908 | -0.066340 | -0.198089 | 0.038850 | 0.033897 |
| cp | 0.110471 | 0.008908 | 1.000000 | -0.036980 | 0.072088 | -0.057663 | 0.063905 |
| trestbps | 0.290476 | -0.066340 | -0.036980 | 1.000000 | 0.131536 | 0.180860 | 0.149242 |
| chol | 0.202644 | -0.198089 | 0.072088 | 0.131536 | 1.000000 | 0.012708 | 0.165046 |
| fbs | 0.132062 | 0.038850 | -0.057663 | 0.180860 | 0.012708 | 1.000000 | 0.068831 |
| restecg | 0.149917 | 0.033897 | 0.063905 | 0.149242 | 0.165046 | 0.068831 | 1.000000 |
| thalach | -0.394563 | -0.060496 | -0.339308 | -0.049108 | -0.000075 | -0.007842 | -0.072290 |
| exang | 0.096489 | 0.143581 | 0.377525 | 0.066691 | 0.059339 | -0.000893 | 0.081874 |
| oldpeak | 0.197123 | 0.106567 | 0.203244 | 0.191243 | 0.038596 | 0.008311 | 0.113726 |
| slope | 0.159405 | 0.033345 | 0.151079 | 0.121172 | -0.009215 | 0.047819 | 0.135141 |
| ca | 0.362210 | 0.091925 | 0.235644 | 0.097954 | 0.115945 | 0.152086 | 0.129021 |
| thal | 0.120795 | 0.370556 | 0.266275 | 0.130612 | 0.023441 | 0.051038 | 0.013612 |
| condition | 0.227075 | 0.278467 | 0.408945 | 0.153490 | 0.080285 | 0.003167 | 0.166343 |

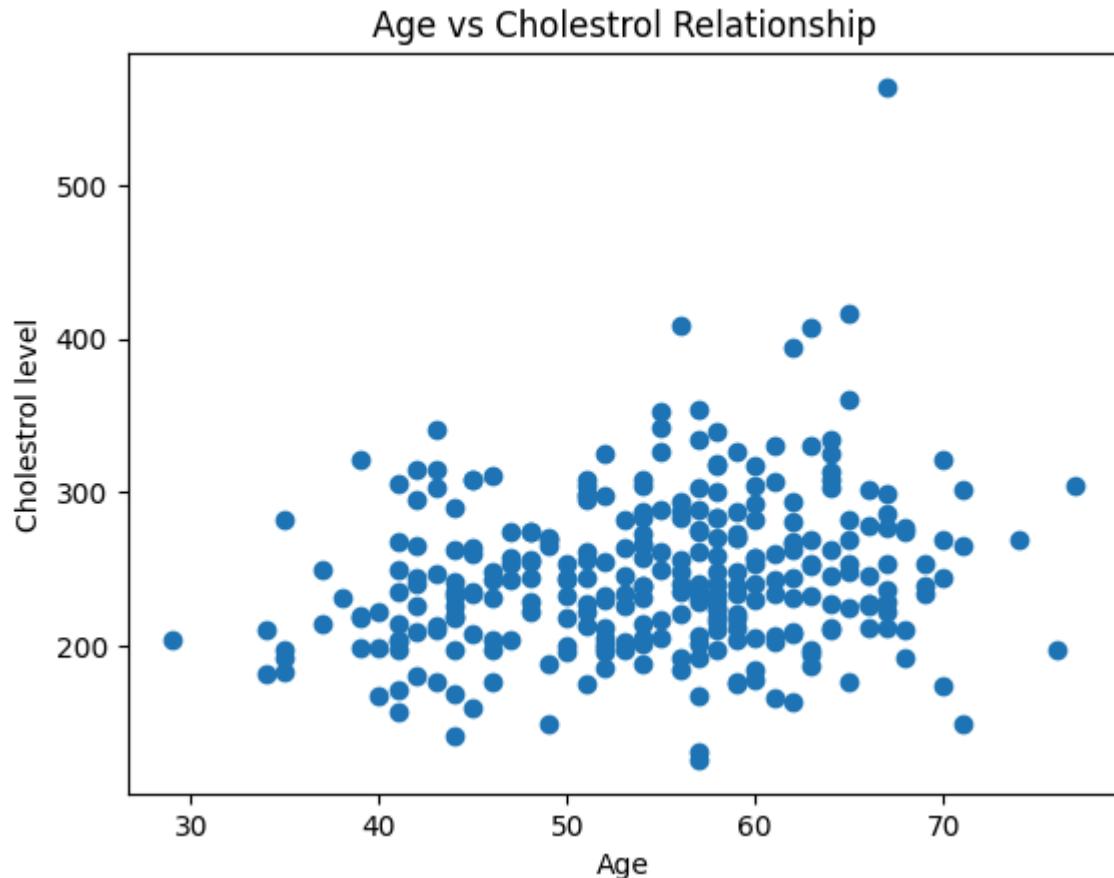


```
In [14]: print(data['age'].corr(data['chol']))
```

0.20264354584662683

```
In [82]: plt.scatter(data['age'],data['chol'])
plt.xlabel('Age')
plt.ylabel('Cholestrol level')
plt.title('Age vs Cholestrol Relationship')
```

Out[82]: Text(0.5, 1.0, 'Age vs Cholestrol Relationship')



6. Chest Pain type vs disease

```
In [83]: cp_group = data.groupby('cp')['condition'].value_counts().unstack()
```

Out[83]: condition 0 1

| | cp | |
|---|----|-----|
| | 0 | 1 |
| 0 | 16 | 7 |
| 1 | 40 | 9 |
| 2 | 65 | 18 |
| 3 | 39 | 103 |

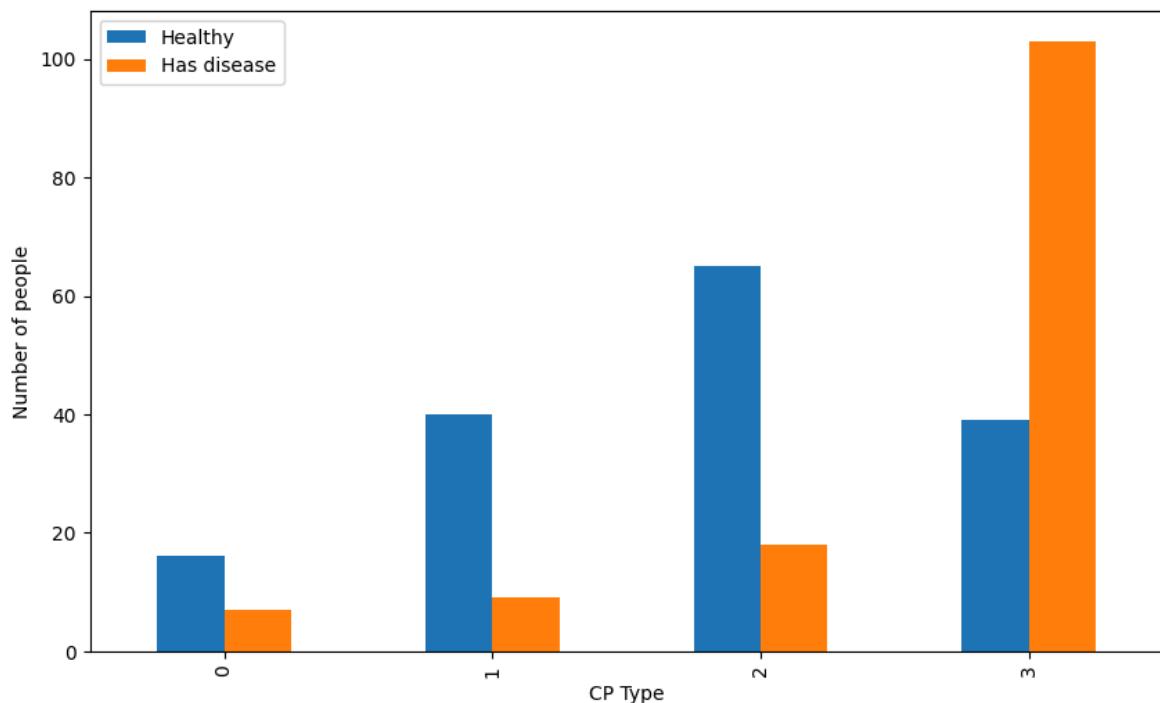
```
In [9]: cp_group_values = data.groupby('cp')['condition'].mean()*100
cp_group_values
```

```
Out[9]: cp
0    30.434783
1    18.367347
2    21.686747
3    72.535211
Name: condition, dtype: float64
```

```
In [89]: #plt.bar(cp_group.index, cp_group.values)

cp_group.plot(kind='bar', figsize=(10, 6))
plt.legend(['Healthy', 'Has disease'])
plt.xlabel('CP Type')
plt.ylabel('Number of people')
```

```
Out[89]: Text(0, 0.5, 'Number of people')
```



From the above graph we can interpret that the type 3 cp is more dangerous

```
In [11]: # Assuming 'summary' is the DataFrame from the previous step
riskiest_cp = cp_group_values.idxmax()
highest_rate = cp_group_values.max()

print(f"The most risky chest pain type is Type {riskiest_cp} with a rate of {highest_rate}")
```

The most risky chest pain type is Type 3 with a rate of 72.54

7. Average cholesterol by gender

```
In [95]: gender = data.groupby('sex')['chol'].mean()
gender
```

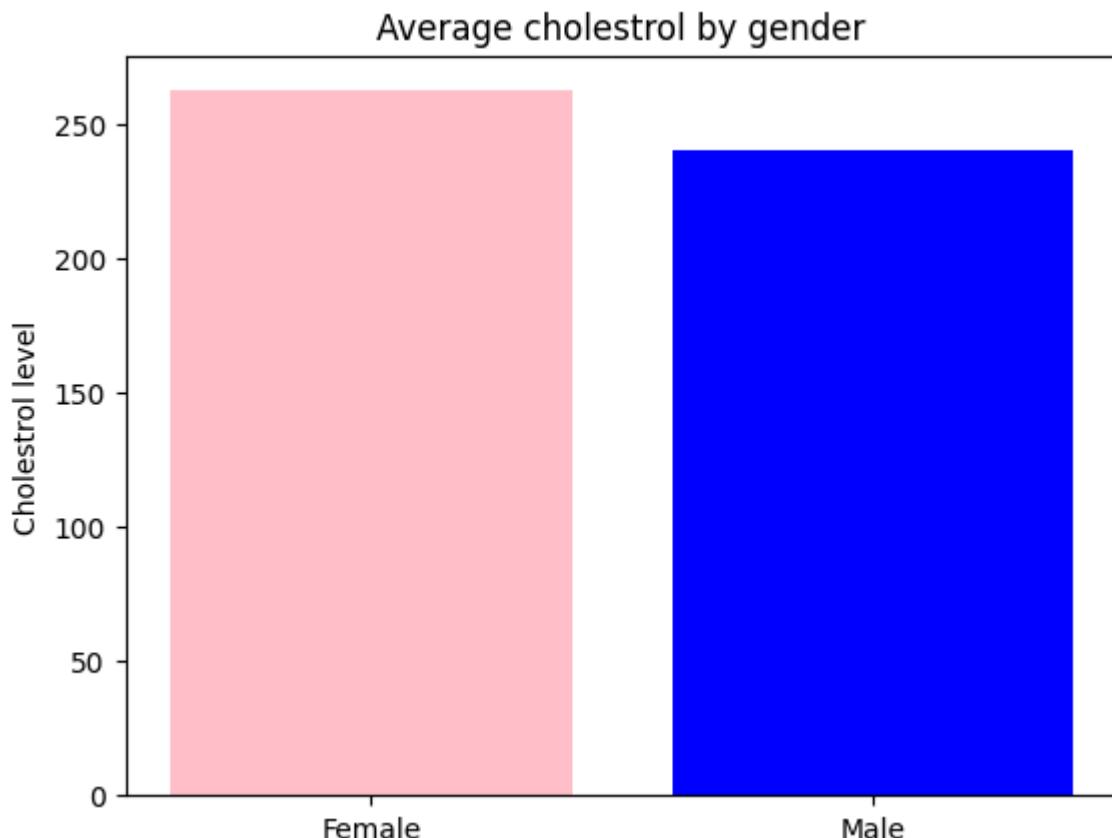
```
Out[95]: sex
0    262.229167
1    240.243781
Name: chol, dtype: float64
```

```
In [15]: print(data['chol'].mean())
```

247.35016835016836

```
In [97]: plt.bar(['Female','Male'],gender.values, color=['Pink','Blue'])
plt.ylabel('Cholestrol level')
plt.title('Average cholestrol by gender')
```

Out[97]: Text(0.5, 1.0, 'Average cholestrol by gender')



8. Resting blood pressure analysis

```
In [101]: avg_bp = data['trestbps'].mean()
print('Average blood pressure of dataset is ',round(avg_bp,2))
```

Average blood pressure of dataset is 131.69

```
In [105]: bpgt140 = data[data['trestbps']>140]
print(f"Number of people with bp greater than 140 is {len(bpgt140)}")
bpgt140.head(10)
```

Number of people with bp greater than 140 is 66

| | age | sex | cp | trestbps | chol | fb | restecg | thalach | exang | oldpeak | slope | ca | th |
|----|-----|-----|----|----------|------|----|---------|---------|-------|---------|-------|----|----|
| 0 | 69 | 1 | 0 | 160 | 234 | 1 | 2 | 131 | 0 | 0.1 | 1 | 1 | 1 |
| 2 | 66 | 0 | 0 | 150 | 226 | 0 | 0 | 114 | 0 | 2.6 | 2 | 0 | 0 |
| 5 | 64 | 1 | 0 | 170 | 227 | 0 | 2 | 155 | 0 | 0.6 | 1 | 0 | 0 |
| 6 | 63 | 1 | 0 | 145 | 233 | 1 | 2 | 150 | 0 | 2.3 | 2 | 0 | 0 |
| 8 | 60 | 0 | 0 | 150 | 240 | 0 | 0 | 171 | 0 | 0.9 | 0 | 0 | 0 |
| 9 | 59 | 1 | 0 | 178 | 270 | 0 | 2 | 145 | 0 | 4.2 | 2 | 0 | 0 |
| 10 | 59 | 1 | 0 | 170 | 288 | 0 | 2 | 159 | 0 | 0.2 | 1 | 0 | 0 |
| 11 | 59 | 1 | 0 | 160 | 273 | 0 | 2 | 125 | 0 | 0.0 | 0 | 0 | 0 |
| 13 | 58 | 0 | 0 | 150 | 283 | 1 | 2 | 162 | 0 | 1.0 | 0 | 0 | 0 |
| 16 | 52 | 1 | 0 | 152 | 298 | 1 | 0 | 178 | 0 | 1.2 | 1 | 0 | 0 |

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```
In [106]: highest_bp = data['trestbps'].max()
print(highest_bp)
```

200

```
In [113]: bp180 = data[data['trestbps']>180]['condition'].value_counts()
bp180
```

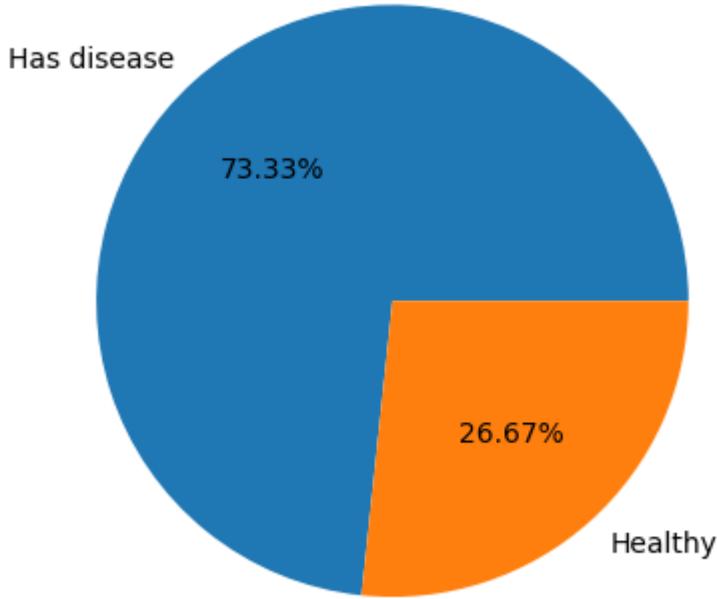
```
Out[113]: condition
1    2
Name: count, dtype: int64
```

```
In [114]: bp160 = data[data['trestbps']>160]['condition'].value_counts()
bp160
```

```
Out[114]: condition
1    11
0     4
Name: count, dtype: int64
```

```
In [117]: plt.pie(bp160.values, labels=['Has disease', 'Healthy'], autopct='%2.2f%%')
```

```
Out[117]: ([<matplotlib.patches.Wedge at 0x2da4f17cce0>,
<matplotlib.patches.Wedge at 0x2da4f1785f0>],
[Text(-0.7360437078139774, 0.817459271271329, 'Has disease'),
Text(0.7360437843500347, -0.8174592023579401, 'Healthy')],
[Text(-0.4014783860803513, 0.4458868752389067, '73.33%'),
Text(0.4014784278272916, -0.4458868376497855, '26.67%')])
```



9. Maximum heartrate vs disease

```
In [119]: avgHR_1 = data[data['condition']==1]['thalach'].mean()
print(avgHR_1)
```

139.1094890510949

```
In [120]: avgHR_0 = data[data['condition']==0]['thalach'].mean()
print(avgHR_0)
```

158.58125

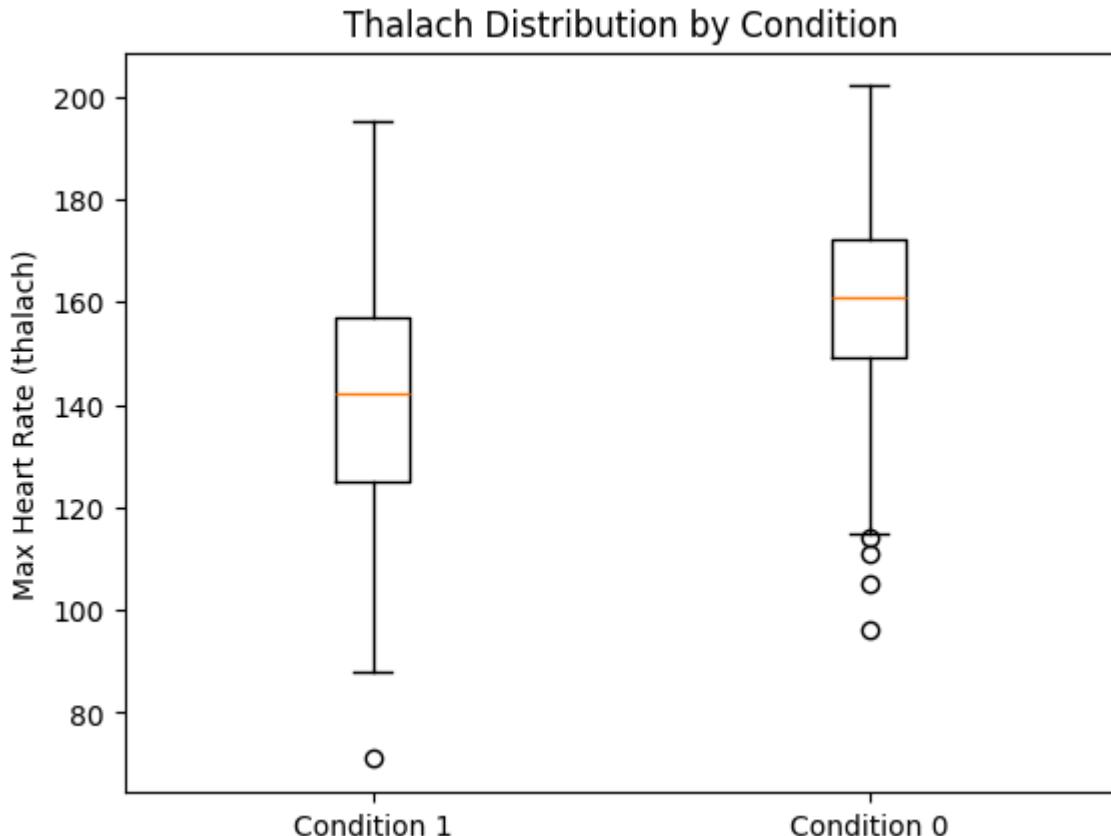
```
In [130]: boxplotdata = [data[data['condition']==1]['thalach'],
                     data[data['condition']==0]['thalach']]

plt.boxplot(boxplotdata, labels=['Condition 1', 'Condition 0'])

plt.title('Thalach Distribution by Condition')
plt.ylabel('Max Heart Rate (thalach)')
```

C:\Users\Administrator\AppData\Local\Temp\ipykernel_8504\3969326452.py:5: MatplotlibDeprecationWarning: The 'labels' parameter of boxplot() has been renamed 'tick_labels' since Matplotlib 3.9; support for the old name will be dropped in 3.11.
 plt.boxplot(boxplotdata, labels=['Condition 1', 'Condition 0'])

```
Out[130]: Text(0, 0.5, 'Max Heart Rate (thalach)')
```



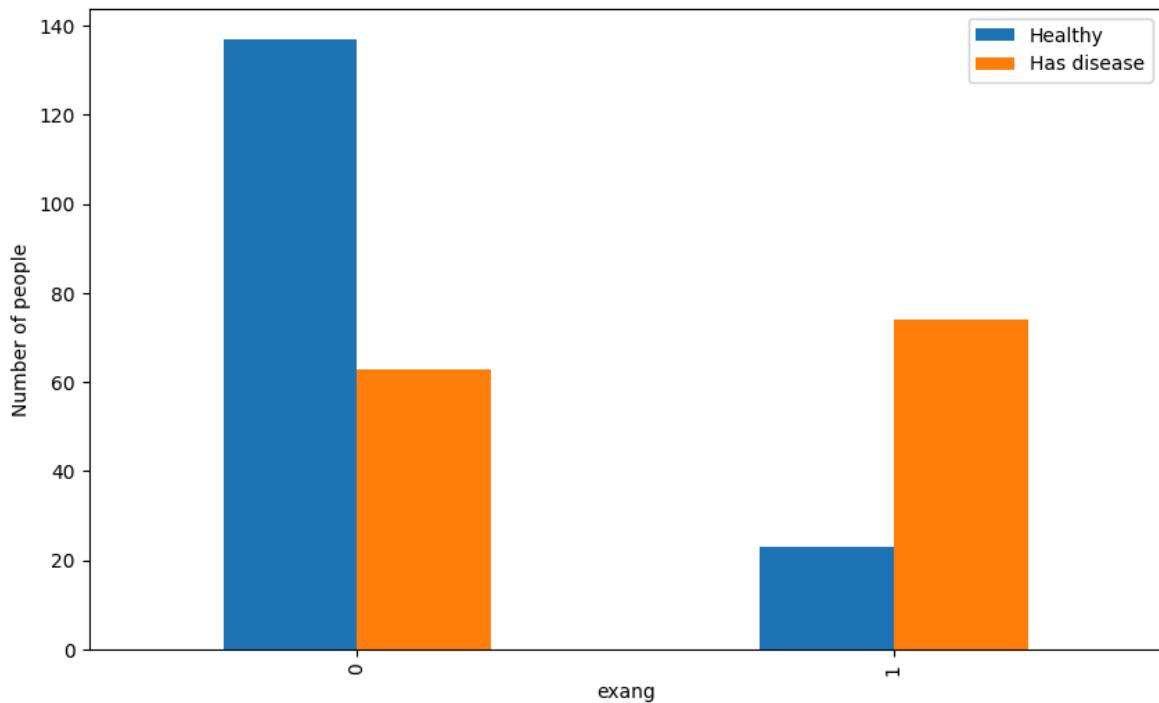
10. Exercise Induced Angina Impact

```
In [136]: exang_data = data.groupby('exang')['condition'].value_counts().unstack()
exang_data
```

```
Out[136]: condition      0      1
           exang
           0    137   63
           1     23   74
```

```
In [138]: exang_data.plot(kind='bar', figsize=(10, 6))
plt.legend(['Healthy', 'Has disease'])
plt.xlabel('exang')
plt.ylabel('Number of people')
```

```
Out[138]: Text(0, 0.5, 'Number of people')
```



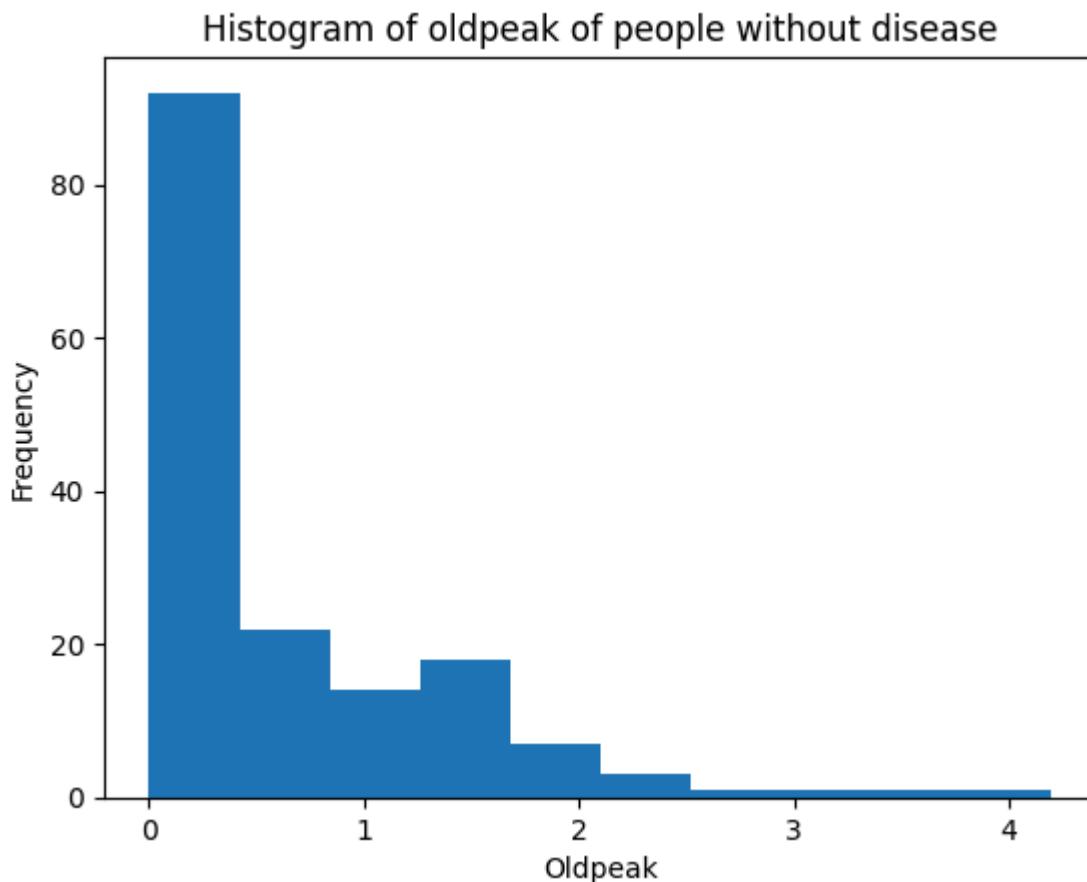
11. ST Depression (oldpeak) Analysis

```
In [141]: oldpeak = data.groupby('condition')['oldpeak'].mean()  
oldpeak
```

```
Out[141]: condition  
0    0.598750  
1    1.589051  
Name: oldpeak, dtype: float64
```

```
In [143]: plt.hist(data[data['condition']==0]['oldpeak'])  
plt.xlabel('Oldpeak')  
plt.ylabel('Frequency')  
plt.title('Histogram of oldpeak of people without disease')
```

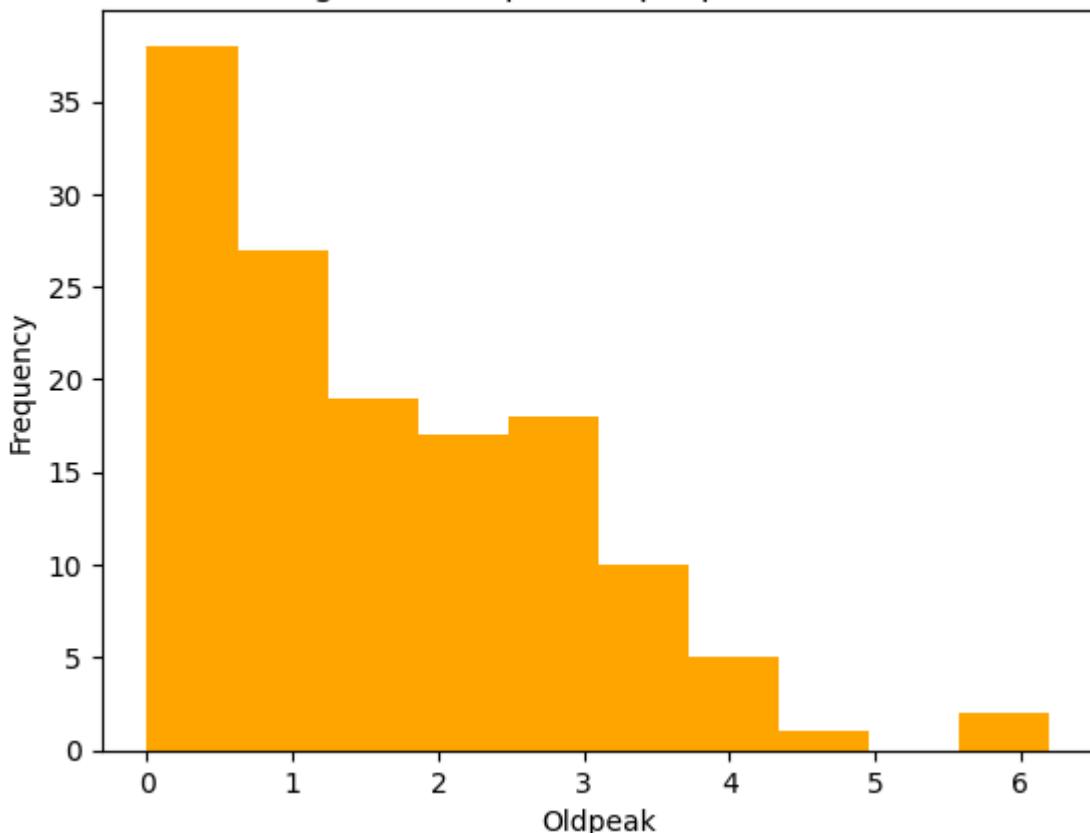
```
Out[143]: Text(0.5, 1.0, 'Histogram of oldpeak of people without disease')
```



```
In [144]: plt.hist(data[data['condition']==1]['oldpeak'], color='orange')
plt.xlabel('Oldpeak')
plt.ylabel('Frequency')
plt.title('Histogram of oldpeak of people with disease')
```

```
Out[144]: Text(0.5, 1.0, 'Histogram of oldpeak of people with disease')
```

Histogram of oldpeak of people with disease



People who has heart disease has a higher value of ST depression

12. Number of Major Vessels (ca) Impact

```
In [145]: ca_group = data.groupby('ca')['condition'].value_counts()
```

```
Out[145]: ca    condition
          0      0            129
                  1            45
          1      1            44
                  0            21
          2      1            31
                  0             7
          3      1            17
                  0             3
Name: count, dtype: int64
```

```
In [167]: #x = (len(ca_group[ca_group['condition']==1] / ca_group['ca']==0))/len(ca_group['ca'])
x = data[data['ca'] == 0]['condition'].mean() * 100
x
```

```
Out[167]: 25.862068965517242
```

```
In [168]: y = data[data['ca'] == 1]['condition'].mean() * 100
y
```

```
Out[168]: 67.6923076923077
```

```
In [169]: z = data[data['ca'] == 2]['condition'].mean() * 100
z
```

```
Out[169]: 81.57894736842105
```

```
In [171]: a = data[data['ca'] == 3]['condition'].mean() * 100
a
```

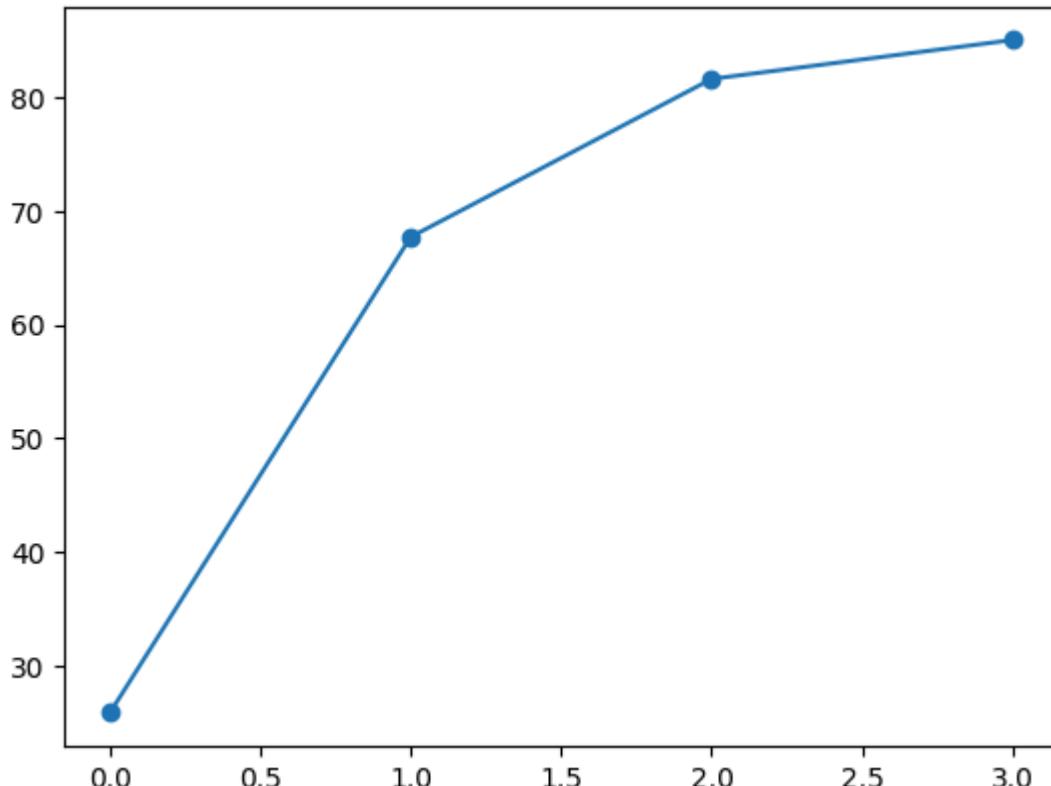
```
Out[171]: 85.0
```

```
In [147]: disease_prob = ca_group.groupby('ca').mean()
disease_prob
```

```
Out[147]: ca
0    87.0
1    32.5
2    19.0
3    10.0
Name: count, dtype: float64
```

```
In [173]: plt.plot([0,1,2,3],[x,y,z,a], linestyle='-', marker='o')
```

```
Out[173]: []
```

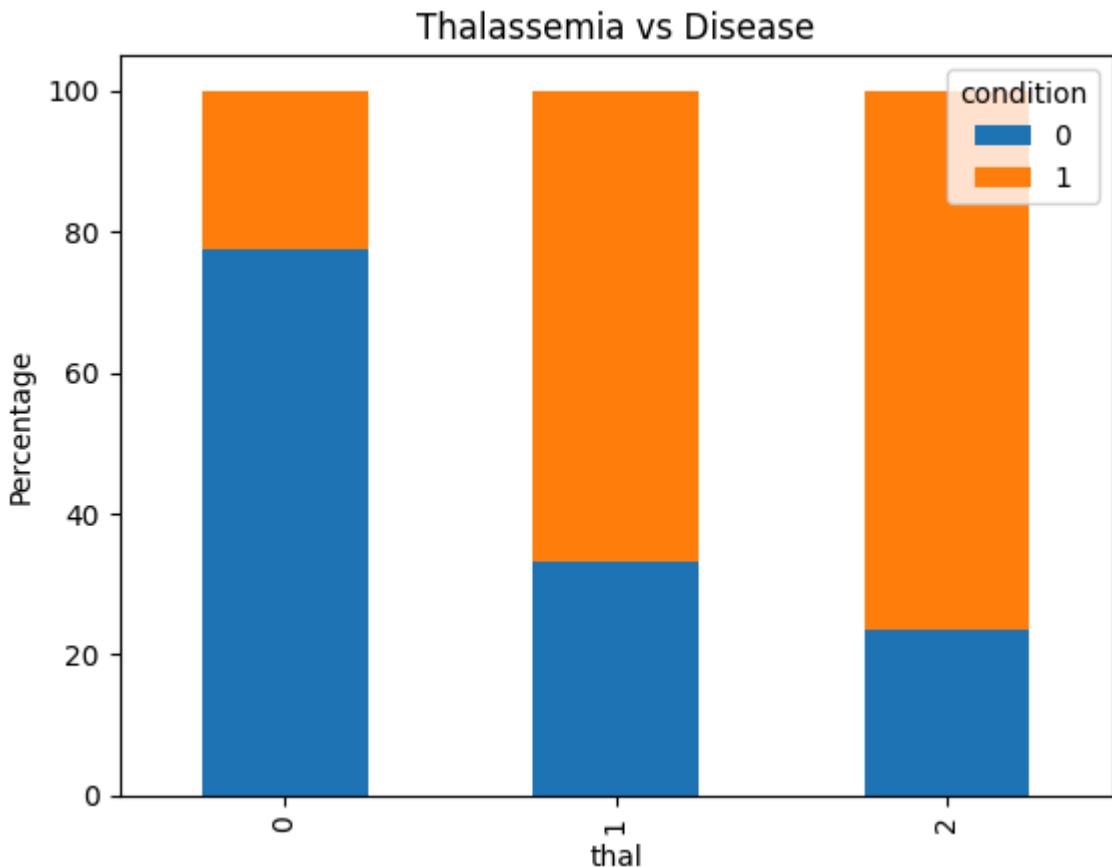


13. Thalassemia vs Disease

```
In [174]: cross= pd.crosstab(data['thal'], data['condition'])

cross = cross.div(cross.sum(axis=1), axis=0) * 100

cross.plot(kind='bar', stacked=True, title="Thalassemia vs Disease")
plt.ylabel('Percentage')
plt.show()
```



In []:

In []:

14. Multi-Factor Risk Analysis

```
In [154]: agegt50 = data[data['age'] > 50]  
agegt50
```

Out[154]:

| | age | sex | cp | trestbps | chol | fbp | restecg | thalach | exang | oldpeak | slope | ca | tl |
|-----|-----|-----|-----|----------|------|-----|---------|---------|-------|---------|-------|-----|-----|
| 0 | 69 | 1 | 0 | 160 | 234 | 1 | 2 | 131 | 0 | 0.1 | 1 | 1 | |
| 1 | 69 | 0 | 0 | 140 | 239 | 0 | 0 | 151 | 0 | 1.8 | 0 | 2 | |
| 2 | 66 | 0 | 0 | 150 | 226 | 0 | 0 | 114 | 0 | 2.6 | 2 | 0 | |
| 3 | 65 | 1 | 0 | 138 | 282 | 1 | 2 | 174 | 0 | 1.4 | 1 | 1 | |
| 4 | 64 | 1 | 0 | 110 | 211 | 0 | 2 | 144 | 1 | 1.8 | 1 | 0 | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 258 | 52 | 1 | 3 | 125 | 212 | 0 | 0 | 168 | 0 | 1.0 | 0 | 2 | |
| 259 | 51 | 0 | 3 | 130 | 305 | 0 | 0 | 142 | 1 | 1.2 | 1 | 0 | |
| 260 | 51 | 1 | 3 | 140 | 298 | 0 | 0 | 122 | 1 | 4.2 | 1 | 3 | |
| 261 | 51 | 1 | 3 | 140 | 261 | 0 | 2 | 186 | 1 | 0.0 | 0 | 0 | |
| 262 | 51 | 1 | 3 | 140 | 299 | 0 | 0 | 173 | 1 | 1.6 | 0 | 0 | |

205 rows × 15 columns



In [157]:

```
cholegt240 = data[data['chol'] > 240]
cholegt240
```

Out[157]:

| | age | sex | cp | trestbps | chol | fbp | restecg | thalach | exang | oldpeak | slope | ca | tl |
|-----|-----|-----|-----|----------|------|-----|---------|---------|-------|---------|-------|-----|-----|
| 3 | 65 | 1 | 0 | 138 | 282 | 1 | 2 | 174 | 0 | 1.4 | 1 | 1 | |
| 9 | 59 | 1 | 0 | 178 | 270 | 0 | 2 | 145 | 0 | 4.2 | 2 | 0 | |
| 10 | 59 | 1 | 0 | 170 | 288 | 0 | 2 | 159 | 0 | 0.2 | 1 | 0 | |
| 11 | 59 | 1 | 0 | 160 | 273 | 0 | 2 | 125 | 0 | 0.0 | 0 | 0 | |
| 13 | 58 | 0 | 0 | 150 | 283 | 1 | 2 | 162 | 0 | 1.0 | 0 | 0 | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 284 | 43 | 1 | 3 | 115 | 303 | 0 | 0 | 181 | 0 | 1.2 | 1 | 0 | |
| 285 | 43 | 1 | 3 | 150 | 247 | 0 | 0 | 171 | 0 | 1.5 | 0 | 0 | |
| 287 | 42 | 0 | 3 | 102 | 265 | 0 | 2 | 122 | 0 | 0.6 | 1 | 0 | |
| 288 | 42 | 1 | 3 | 136 | 315 | 0 | 0 | 125 | 1 | 1.8 | 1 | 0 | |
| 296 | 35 | 1 | 3 | 126 | 282 | 0 | 2 | 156 | 1 | 0.0 | 0 | 0 | |

151 rows × 15 columns



In [158]:

```
bpgt140 = data[data['trestbps'] > 140]
bpgt140
```

Out[158]:

| | age | sex | cp | trestbps | chol | fbp | restecg | thalach | exang | oldpeak | slope | ca | tl |
|-----|-----|-----|-----|----------|------|-----|---------|---------|-------|---------|-------|-----|----|
| 0 | 69 | 1 | 0 | 160 | 234 | 1 | 2 | 131 | 0 | 0.1 | 1 | 1 | |
| 2 | 66 | 0 | 0 | 150 | 226 | 0 | 0 | 114 | 0 | 2.6 | 2 | 0 | |
| 5 | 64 | 1 | 0 | 170 | 227 | 0 | 2 | 155 | 0 | 0.6 | 1 | 0 | |
| 6 | 63 | 1 | 0 | 145 | 233 | 1 | 2 | 150 | 0 | 2.3 | 2 | 0 | |
| 8 | 60 | 0 | 0 | 150 | 240 | 0 | 0 | 171 | 0 | 0.9 | 0 | 0 | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | |
| 263 | 50 | 1 | 3 | 150 | 243 | 0 | 2 | 128 | 0 | 2.6 | 1 | 0 | |
| 264 | 50 | 1 | 3 | 144 | 200 | 0 | 2 | 126 | 1 | 0.9 | 1 | 0 | |
| 277 | 45 | 1 | 3 | 142 | 309 | 0 | 2 | 147 | 1 | 0.0 | 1 | 3 | |
| 285 | 43 | 1 | 3 | 150 | 247 | 0 | 0 | 171 | 0 | 1.5 | 0 | 0 | |
| 292 | 40 | 1 | 3 | 152 | 223 | 0 | 0 | 181 | 0 | 0.0 | 0 | 0 | |

66 rows × 15 columns



In [162]:

```
x = (len(agegt50[agegt50['condition']==1])/len(agegt50))*100
print(f"The percentage of people with disease having age > 50 is : {round(x,2)}")
```

The percentage of people with disease having age > 50 is : 53.17

In [164]:

```
x = (len(cholegt240[cholegt240['condition']==1])/len(cholegt240))*100
print(f"The percentage of people with disease having cholestrol > 240 is : {round(x,2)}")
```

The percentage of people with disease having cholestrol > 240 is : 52.32

In [165]:

```
bpgt140
x = (len(bpgt140[bpgt140['condition']==1])/len(bpgt140))*100
print(f"The percentage of people with disease having bp > 140 is : {round(x,2)}")
```

The percentage of people with disease having bp > 140 is : 59.09

15. Create Risk Score (Custom Analysis)

In [150]:

```
data['risk_score'] = (data['chol']/200) + (data['trestbps']/120) + (data['oldpeak'])
data['risk_score']
```

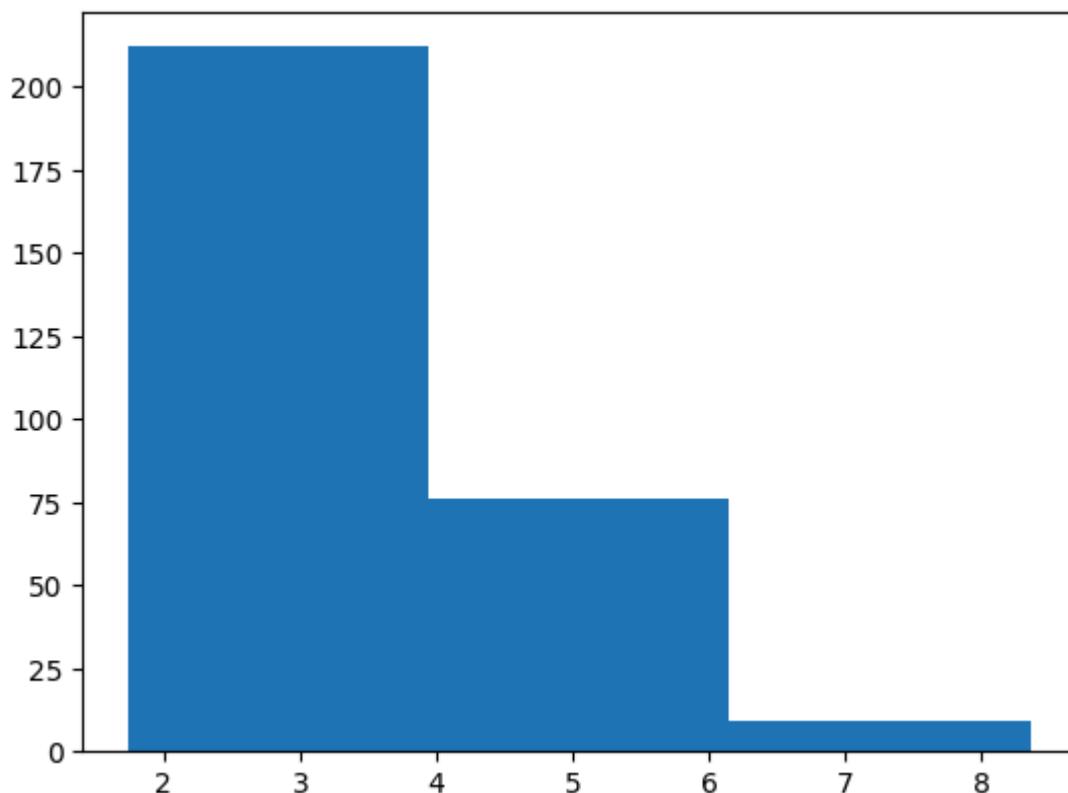
Out[150]:

| | |
|-----|----------|
| 0 | 2.603333 |
| 1 | 4.161667 |
| 2 | 4.980000 |
| 3 | 3.960000 |
| 4 | 3.771667 |
| | ... |
| 292 | 2.381667 |
| 293 | 3.278333 |
| 294 | 3.590000 |
| 295 | 3.465000 |
| 296 | 2.460000 |

Name: risk_score, Length: 297, dtype: float64

```
In [151]: plt.hist(data['risk_score'], bins=3)
```

```
Out[151]: (array([212.,  76.,   9.]),
 array([1.73333333, 3.94        , 6.14666667, 8.35333333]),
 <BarContainer object of 3 artists>)
```

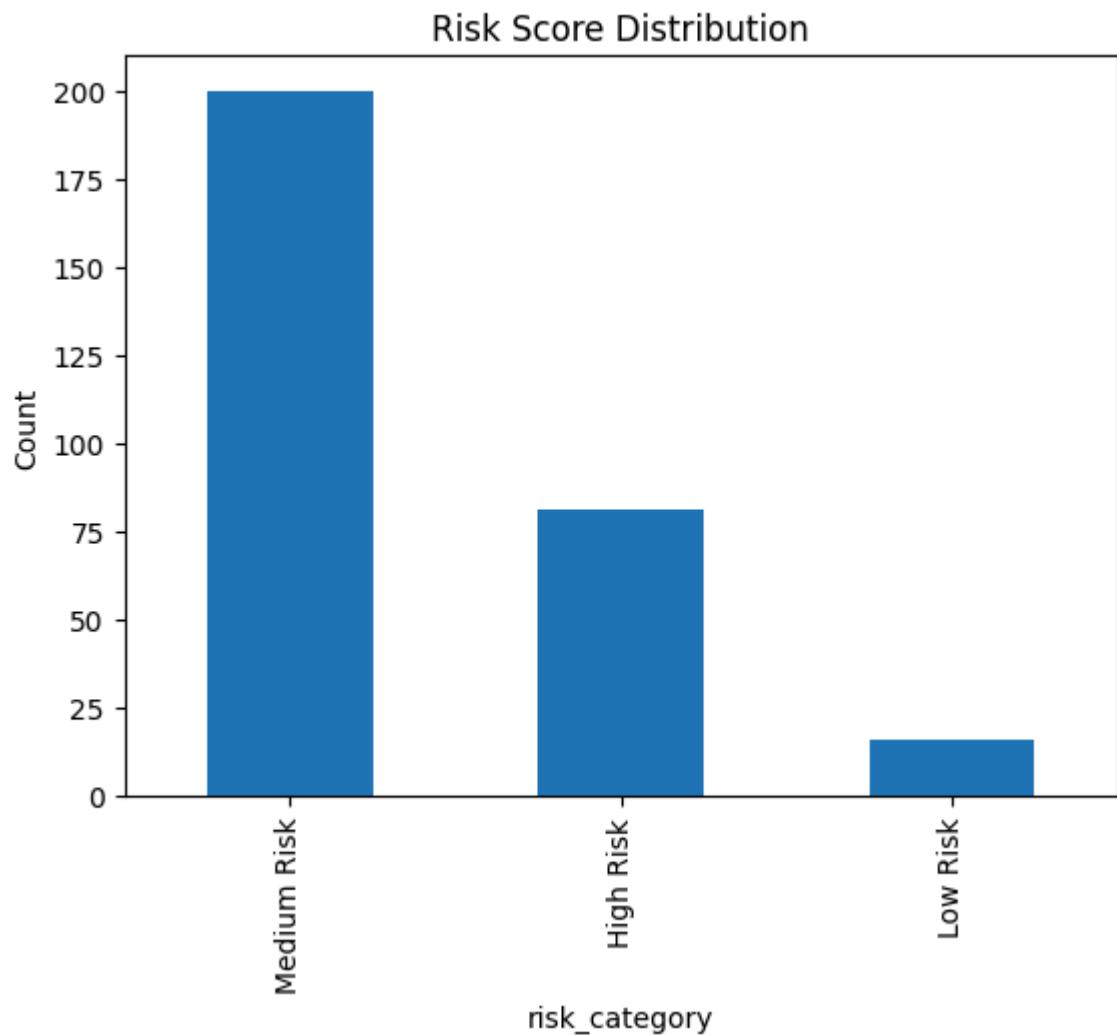


```
In [180]: #data['risk_score'] = (data['chol'] / 200) + (data['trestbps'] / 120) + data['oldf']

conditions = [
    (data['risk_score'] < 2),
    (data['risk_score'] >= 2) & (data['risk_score'] < 4),
    (data['risk_score'] >= 4)
]
choices = ['Low Risk', 'Medium Risk', 'High Risk']

data['risk_category'] = np.select(conditions, choices)

data['risk_category'].value_counts().plot(kind='bar', title="Risk Score Distribution")
plt.ylabel('Count')
plt.show()
```



```
In [16]: data.corr()
```

Out[16]:

| | age | sex | cp | trestbps | chol | fbs | restecg | |
|------------------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|----|
| age | 1.000000 | -0.092399 | 0.110471 | 0.290476 | 0.202644 | 0.132062 | 0.149917 | -0 |
| sex | -0.092399 | 1.000000 | 0.008908 | -0.066340 | -0.198089 | 0.038850 | 0.033897 | -0 |
| cp | 0.110471 | 0.008908 | 1.000000 | -0.036980 | 0.072088 | -0.057663 | 0.063905 | -0 |
| trestbps | 0.290476 | -0.066340 | -0.036980 | 1.000000 | 0.131536 | 0.180860 | 0.149242 | -0 |
| chol | 0.202644 | -0.198089 | 0.072088 | 0.131536 | 1.000000 | 0.012708 | 0.165046 | -0 |
| fbs | 0.132062 | 0.038850 | -0.057663 | 0.180860 | 0.012708 | 1.000000 | 0.068831 | -0 |
| restecg | 0.149917 | 0.033897 | 0.063905 | 0.149242 | 0.165046 | 0.068831 | 1.000000 | -0 |
| thalach | -0.394563 | -0.060496 | -0.339308 | -0.049108 | -0.000075 | -0.007842 | -0.072290 | -0 |
| exang | 0.096489 | 0.143581 | 0.377525 | 0.066691 | 0.059339 | -0.000893 | 0.081874 | -0 |
| oldpeak | 0.197123 | 0.106567 | 0.203244 | 0.191243 | 0.038596 | 0.008311 | 0.113726 | -0 |
| slope | 0.159405 | 0.033345 | 0.151079 | 0.121172 | -0.009215 | 0.047819 | 0.135141 | -0 |
| ca | 0.362210 | 0.091925 | 0.235644 | 0.097954 | 0.115945 | 0.152086 | 0.129021 | -0 |
| thal | 0.120795 | 0.370556 | 0.266275 | 0.130612 | 0.023441 | 0.051038 | 0.013612 | -0 |
| condition | 0.227075 | 0.278467 | 0.408945 | 0.153490 | 0.080285 | 0.003167 | 0.166343 | -0 |

Does cholesterol strongly impact heart disease?

It has a very small positive corelation, ie a person with high cholesterol has a small chance of having heart disease.

Is male population more vulnerable? Yes based on the data provided men have a higher chance of having heart disease.

Does exercise-induced angina significantly increase risk? Yes exercise-induced angina can significantly increase the risk of heart disease as they have a very strong positive corelation.

Which feature has strongest correlation with disease? thal has the strongest correlation with having a heart disease.