

PROBLEM STATEMENT

To predict and analyze which gender has a high chance of survival at the time of disaster

In [2]:

```
#import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
```

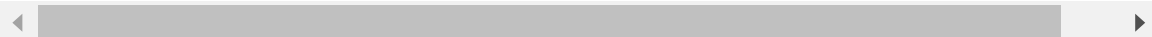
In [3]:

```
df=pd.read_csv(r"C:\Users\shaha\Downloads\heart_disease_data.csv")
df
```

Out[3]:

| | age | sex | cp | trestbps | chol | fbs | restecg | thalach | exang | oldpeak | slope | ca | thal | ta |
|-----|-----|-----|-----|----------|------|-----|---------|---------|-------|---------|-------|-----|------|-----|
| 0 | 63 | 1 | 3 | 145 | 233 | 1 | 0 | 150 | 0 | 2.3 | 0 | 0 | 1 | |
| 1 | 37 | 1 | 2 | 130 | 250 | 0 | 1 | 187 | 0 | 3.5 | 0 | 0 | 2 | |
| 2 | 41 | 0 | 1 | 130 | 204 | 0 | 0 | 172 | 0 | 1.4 | 2 | 0 | 2 | |
| 3 | 56 | 1 | 1 | 120 | 236 | 0 | 1 | 178 | 0 | 0.8 | 2 | 0 | 2 | |
| 4 | 57 | 0 | 0 | 120 | 354 | 0 | 1 | 163 | 1 | 0.6 | 2 | 0 | 2 | |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 298 | 57 | 0 | 0 | 140 | 241 | 0 | 1 | 123 | 1 | 0.2 | 1 | 0 | 3 | |
| 299 | 45 | 1 | 3 | 110 | 264 | 0 | 1 | 132 | 0 | 1.2 | 1 | 0 | 3 | |
| 300 | 68 | 1 | 0 | 144 | 193 | 1 | 1 | 141 | 0 | 3.4 | 1 | 2 | 3 | |
| 301 | 57 | 1 | 0 | 130 | 131 | 0 | 1 | 115 | 1 | 1.2 | 1 | 1 | 3 | |
| 302 | 57 | 0 | 1 | 130 | 236 | 0 | 0 | 174 | 0 | 0.0 | 1 | 1 | 2 | |

303 rows × 14 columns



In [4]:

```
df.head()
```

Out[4]:

| | age | sex | cp | trestbps | chol | fbs | restecg | thalach | exang | oldpeak | slope | ca | thal | targ |
|---|-----|-----|----|----------|------|-----|---------|---------|-------|---------|-------|----|------|------|
| 0 | 63 | 1 | 3 | 145 | 233 | 1 | 0 | 150 | 0 | 2.3 | 0 | 0 | 1 | |
| 1 | 37 | 1 | 2 | 130 | 250 | 0 | 1 | 187 | 0 | 3.5 | 0 | 0 | 2 | |
| 2 | 41 | 0 | 1 | 130 | 204 | 0 | 0 | 172 | 0 | 1.4 | 2 | 0 | 2 | |
| 3 | 56 | 1 | 1 | 120 | 236 | 0 | 1 | 178 | 0 | 0.8 | 2 | 0 | 2 | |
| 4 | 57 | 0 | 0 | 120 | 354 | 0 | 1 | 163 | 1 | 0.6 | 2 | 0 | 2 | |

In [5]:

```
df.tail()
```

Out[5]:

| | age | sex | cp | trestbps | chol | fbs | restecg | thalach | exang | oldpeak | slope | ca | thal | ta |
|-----|-----|-----|----|----------|------|-----|---------|---------|-------|---------|-------|----|------|----|
| 298 | 57 | 0 | 0 | 140 | 241 | 0 | 1 | 123 | 1 | 0.2 | 1 | 0 | 3 | |
| 299 | 45 | 1 | 3 | 110 | 264 | 0 | 1 | 132 | 0 | 1.2 | 1 | 0 | 3 | |
| 300 | 68 | 1 | 0 | 144 | 193 | 1 | 1 | 141 | 0 | 3.4 | 1 | 2 | 3 | |
| 301 | 57 | 1 | 0 | 130 | 131 | 0 | 1 | 115 | 1 | 1.2 | 1 | 1 | 3 | |
| 302 | 57 | 0 | 1 | 130 | 236 | 0 | 0 | 174 | 0 | 0.0 | 1 | 1 | 2 | |

In [6]:

```
df.shape
```

Out[6]:

```
(303, 14)
```

In [9]:

```
df.isnull().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
target       0
dtype: int64
```

In [7]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 303 entries, 0 to 302
Data columns (total 14 columns):
 #   Column      Non-Null Count  Dtype  
---  -
 0   age         303 non-null   int64  
 1   sex         303 non-null   int64  
 2   cp          303 non-null   int64  
 3   trestbps    303 non-null   int64  
 4   chol        303 non-null   int64  
 5   fbs         303 non-null   int64  
 6   restecg     303 non-null   int64  
 7   thalach     303 non-null   int64  
 8   exang       303 non-null   int64  
 9   oldpeak     303 non-null   float64 
10   slope       303 non-null   int64  
11   ca          303 non-null   int64  
12   thal        303 non-null   int64  
13   target      303 non-null   int64  
dtypes: float64(1), int64(13)
memory usage: 33.3 KB
```

In [8]:

```
df.describe()
```

Out[8]:

| age | sex | cp | trestbps | chol | fbs | restecg | thalach | ex |
|-----|------------|------------|------------|------------|------------|------------|------------|------------|
| 30 | 303.000000 | 303.000000 | 303.000000 | 303.000000 | 303.000000 | 303.000000 | 303.000000 | 303.000000 |
| 37 | 0.683168 | 0.966997 | 131.623762 | 246.264026 | 0.148515 | 0.528053 | 149.646865 | 0.300000 |
| 31 | 0.466011 | 1.032052 | 17.538143 | 51.830751 | 0.356198 | 0.525860 | 22.905161 | 0.400000 |
| 30 | 0.000000 | 0.000000 | 94.000000 | 126.000000 | 0.000000 | 0.000000 | 71.000000 | 0.000000 |
| 30 | 0.000000 | 0.000000 | 120.000000 | 211.000000 | 0.000000 | 0.000000 | 133.500000 | 0.000000 |
| 30 | 1.000000 | 1.000000 | 130.000000 | 240.000000 | 0.000000 | 1.000000 | 153.000000 | 0.000000 |
| 30 | 1.000000 | 2.000000 | 140.000000 | 274.500000 | 0.000000 | 1.000000 | 166.000000 | 1.000000 |
| 30 | 1.000000 | 3.000000 | 200.000000 | 564.000000 | 1.000000 | 2.000000 | 202.000000 | 1.000000 |

In [10]:

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

Out[10]:

```
target
1    165
0    138
Name: count, dtype: int64
```

#splitting the features and target

In [11]:

```
x=df.drop(columns='target',axis=1)
y=df['target']
```

In [12]:

```
print(x)
```

| | age | sex | cp | trestbps | chol | fbs | restecg | thalach | exang | oldpeak |
|-----|-----|-----|----|----------|------|-----|---------|---------|-------|---------|
| 0 | 63 | 1 | 3 | 145 | 233 | 1 | 0 | 150 | 0 | 2.3 |
| \ | | | | | | | | | | |
| 1 | 37 | 1 | 2 | 130 | 250 | 0 | 1 | 187 | 0 | 3.5 |
| 2 | 41 | 0 | 1 | 130 | 204 | 0 | 0 | 172 | 0 | 1.4 |
| 3 | 56 | 1 | 1 | 120 | 236 | 0 | 1 | 178 | 0 | 0.8 |
| 4 | 57 | 0 | 0 | 120 | 354 | 0 | 1 | 163 | 1 | 0.6 |
| .. | ... | ... | .. | ... | ... | ... | ... | ... | ... | ... |
| 298 | 57 | 0 | 0 | 140 | 241 | 0 | 1 | 123 | 1 | 0.2 |
| 299 | 45 | 1 | 3 | 110 | 264 | 0 | 1 | 132 | 0 | 1.2 |
| 300 | 68 | 1 | 0 | 144 | 193 | 1 | 1 | 141 | 0 | 3.4 |
| 301 | 57 | 1 | 0 | 130 | 131 | 0 | 1 | 115 | 1 | 1.2 |
| 302 | 57 | 0 | 1 | 130 | 236 | 0 | 0 | 174 | 0 | 0.0 |

| | slope | ca | thal |
|-----|-------|----|------|
| 0 | 0 | 0 | 1 |
| 1 | 0 | 0 | 2 |
| 2 | 2 | 0 | 2 |
| 3 | 2 | 0 | 2 |
| 4 | 2 | 0 | 2 |
| .. | ... | .. | ... |
| 298 | 1 | 0 | 3 |
| 299 | 1 | 0 | 3 |
| 300 | 1 | 2 | 3 |
| 301 | 1 | 1 | 3 |
| 302 | 1 | 1 | 2 |

[303 rows x 13 columns]

In [13]:

```
print(y)
```

```

0      1
1      1
2      1
3      1
4      1
..
298    0
299    0
300    0
301    0
302    0

```

Name: target, Length: 303, dtype: int64

In [16]:

```
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, stratify=y, rand
```

In [18]:

```
print(x.shape, x_train.shape, x_test.shape)
```

```
(303, 13) (242, 13) (61, 13)
```

MODEL TRAINING FOR LOGISTIC REGRESSION

In [22]:

```
model = LogisticRegression()
```

In [24]:

```
model.fit(x_train,y_train)
```

C:\Users\shaha\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\linear_model_logistic.py:458: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html> (<https://scikit-learn.org/stable/modules/preprocessing.html>)

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)

```
n_iter_i = _check_optimize_result(
```

Out[24]:

```
▼ LogisticRegression  
LogisticRegression()
```

MODEL PREDICTION

In [31]:

```
# accuracy on training data  
x_train_prediction = model.predict(x_train)  
training_data_accuracy = accuracy_score(x_train_prediction, y_train)
```

In [34]:

```
# accuracy on test data  
x_test_prediction = model.predict(x_test)  
test_data_accuracy = accuracy_score(x_test_prediction, y_test)
```

In [35]:

```
print('Accuracy on Test data : ', test_data_accuracy)
```

```
Accuracy on Test data : 0.819672131147541
```

In []: