

## **Problem Statement**

Heart attacks are the number one cause of death worldwide. According to WHO data there are 17 million people in the world who die from heart disease. Many factors make a person suffer from a heart attack such as obesity, cholesterol etc. From the dataset that we have taken from Kaggle, namely the heart attack dataset, we want to predict how many patients get the heart attack using several machine learning models what is good to use in this dataset.

# **3** Objective

We're trying to predict the 'target' value from the dataset using numerical and encoded categorical features. We predict the 'target' value to help someone know the possibility of having a heart attack and determine whether they need a treatment or not.

### Machine Learning Modelling







Accuracy score: 0.639344262295082

# **Decision Tree**

Accuracy score: 0.7540983606557377 F1-Score: 0.7692307692307693



Accuracy score: 0.8360655737704918 F1-Score: 0.8611111111111111

## **Data Overview**

- Key attributes of the data: age:int
- sex:int
- chest pain type (cp): int
- resting blood pressure: int
- serum cholesterol in mg/dl: int
- fasting blood sugar: int
- restecg: int
- thalach: int exang: int
- oldpeak: float
- ca: int
- thal:int
- target : int

The data has 303 rows and no null values.

#### ata Understanding





## **Evaluation**

Of the four classification methods that we have done. We get the results of the classification report of each method. As can be seen from the results of the accuracy value and F-1 score of each method. Decision Tree and Random forests had the highest yields, meaning that they performed better at predicting someone would have a heart attack than other models. However, we still need to know whether the Decision Tree and Random Forest model is really consistent in doing good predictability on new data that has not yet been seen

#### Conclusion

Classification Report				
	precision	recall	f1-score	support
9 1	1.00 1.00	1.00	1.00 1.00	27 34
accuracy macro avg weighted avg	1.00 1.00	1.00	1.00 1.00 1.00	61 61 61











