# CARDIVASCULAR RISK PREDICTION



#### PROJECT SUMMARY

#### Data Description

```
Demographic:
Sex: male or female("M" or "F")
Age: Age of the patient; (Continuous -
Although the recorded ages have been trunc
ated to whole numbers, the concept of age i
s continuous)
```

#### Behavioral

```
is_smoking: whether or not the patient is a
  current smoker ("YES" or "NO")
Cigs Per Day: the number of cigarettes that
  the person smoked on average in one day.(c
  an be
  considered continuous as one can have any n
  umber of cigarettes, even half a cigarette.
)
```

#### Medical ( history)

BP Meds: whether or not the patient was on blood pr essure medication (Nominal)

**Prevalent Stroke:** whether or not the patient had previously had a stroke (Nominal)

Prevalent Hyp: whether or not the patient was hyper tensive (Nominal)

**Diabetes:** whether or not the patient had diabetes ( Nominal)

Tot Chol: total cholesterol level (Continuous)

Sys BP: systolic blood pressure (Continuous)

Dia BP: diastolic blood pressure (Continuous)

BMI: Body Mass Index (Continuous)

Heart Rate: heart rate (Continuous -

In medical research, variables such as heart rate though in

fact discrete, yet are considered continuous becaus e of large number of possible values.)

#### PROBLEM STATEMENT

The Cardivascular data consists of set of columns that sp eaks about the patient is suffering from heart disease or not

first we have to load the dataset and by understanding the dataset with the Meaningfull features and rows we perform the model

There are 3390 rows and 16 columns with in the dataset

#### **FEATURES**

age: Speaks about the age of the patient

**education :** shows the education the the patient done in f our different phases 1. schooling, 2. intermediate, 3.UG, 4.P  $_{\rm G}$ 

\*\*sex\*\*: Represents whether the patient is male or female

is\_smoking:Represent whether the patient is smoking or no
t

cigsPerDay: If smoking then how many cigars the patient ca
n take per a day

**BPMEDS:** is there bp in thee patient

**PREVALENTSTROKE:** is there any previous stroke in the patient

diabetes: Is the patient suffering with dia
betes

TOTCHOL : Total cholestrol

SYSBP :systollic blood pressure

DIABP : Diastollic blood pressure

BMI :body mass index

**HEARTRATE** :represent the heart rate of the patient

glucose :glucose levels of the patient

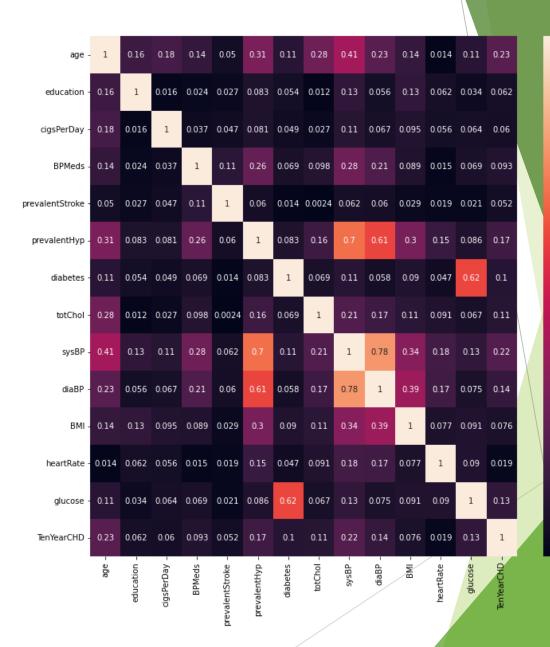
TENYEARCHD: Predicting ten year coronary heart disease with in the patient



#### Finding the co related features

Here we can see that different featues co related with each other

- 1. Sysbp and diabp features are highly co related with each other with .78 percentage among others
- 2. There is no significant correlation between independent and dependent varibles
- 3.but there is nhigh correlation between independent variables
- 4. Also some of the other features tht are correlatd with eeach oter with some percentage

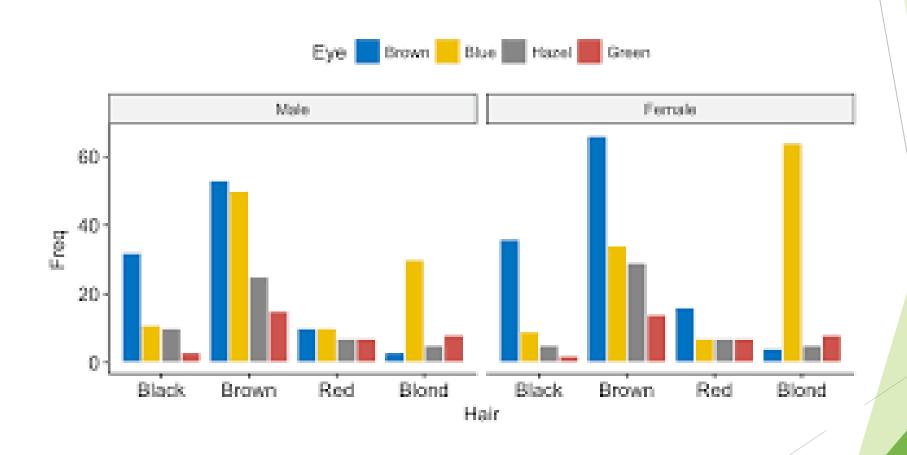


- 0.8

- 0.6

0.2

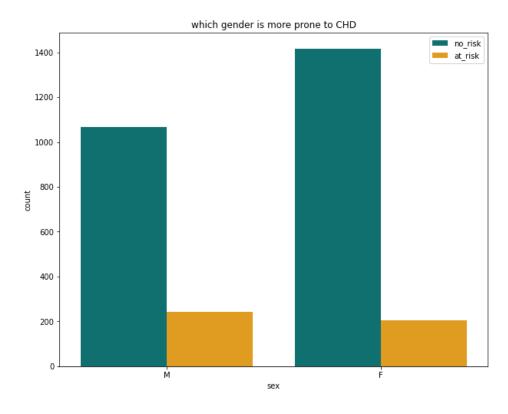
# HERE WE HAVE DONE SOME CATEGORICAL VISUALISATION ON THE DATASET



# Gender prone too heart disease

By the graph we intend that females are less effective of disease when compared to males

 As the green represents the no risk of disease and yellow represents at risk

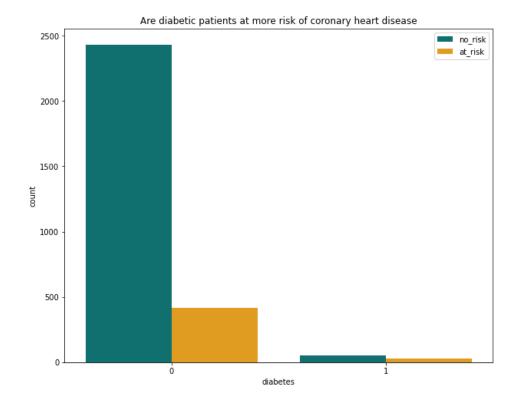


# Are diabetic patients at more risk of coronary heart disease

we representts the diabetic patients at more risk o f coronary heart disease

0 indicates tenyearCHD occurs and 1 indicates not

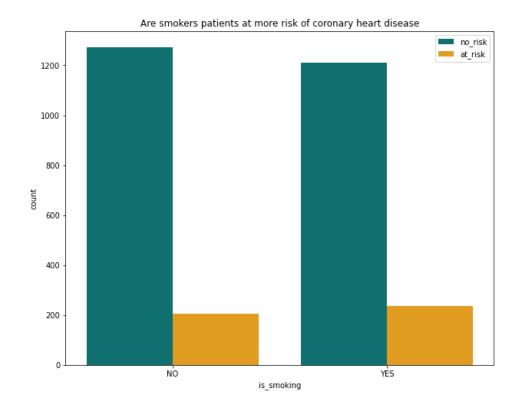
here we ca see that diabetes patients are more ser ioiusly occurs with CHD



## Are smokers patients at more risk of coronary heart diseas

WE compared that smoking effects the CHD then we noptices that

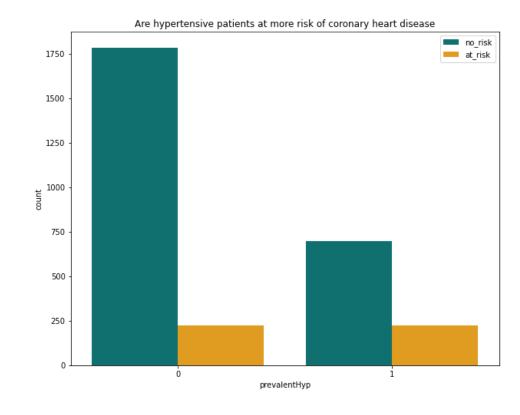
there is no relation of smoiking causes tha CHD by this graph



# Are hypertensive patients at more risk of coronary heart disease

by this hypertensive patients at not at all more risk of coronary heart disease by prevlenthyp

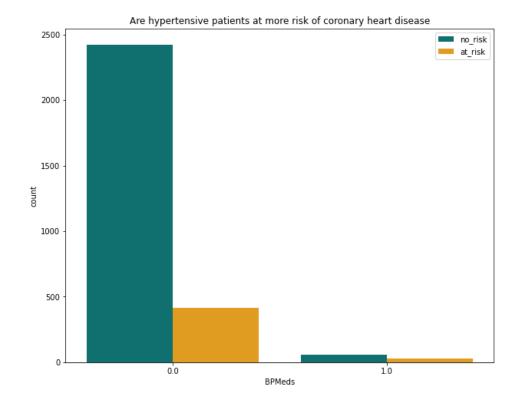
it doesn't occurs difference among them



# Are patients with blood pressure on mediacation at more risk of coronary heart disease

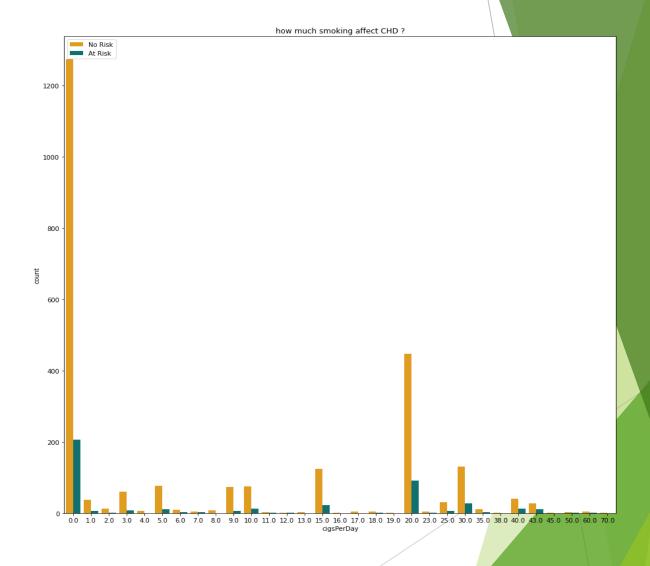
Patience with bloodpressure can indicate symlp toms of occuring the CHD

theres almost equal chancewsn oif getiing the CH D with BP



#### **How much smoking affect CHD**

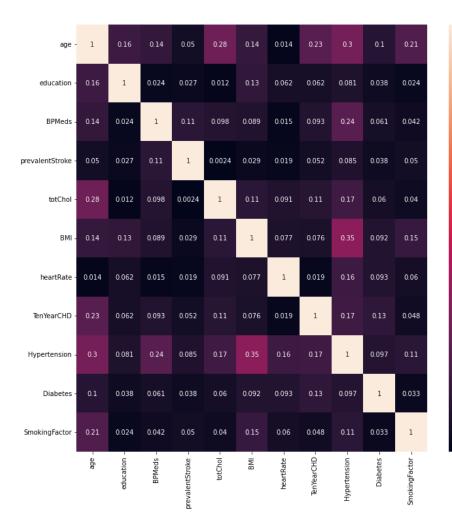
By understaning the graph we may know that smoking doesn't put any efforts to get effected by the heart disease



## Plotting the correlation matrix using heatmap

Here we can see that the multicollinearity doesn't exists between the columns

Every feature consists of its own correlation

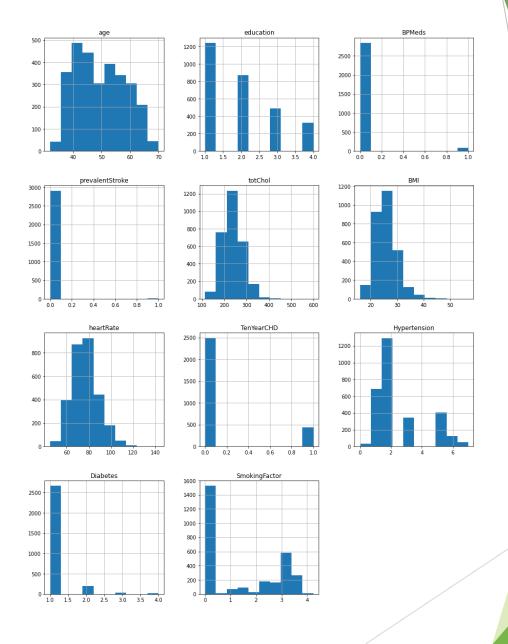


- 0.8

- 0.6

### plot histogram to see the distribution of the dat

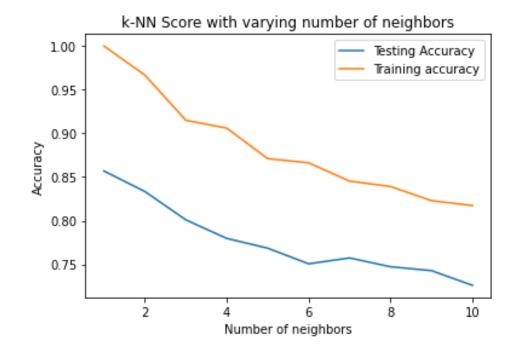
BPMeds and prevalentStroke does no t help in explaining variance so w e can remove those columns



#### **ML** model implementation KNN

Webcan see that test score is incr easing as number of neighbours inc reases.

Let's try to find best parameter f or knn.

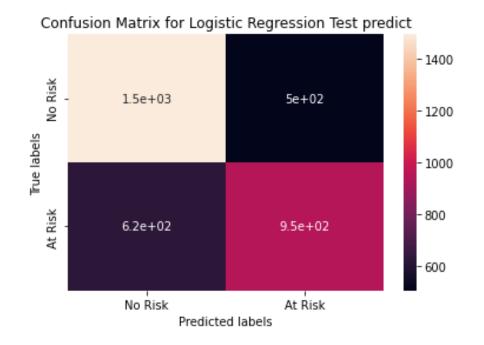


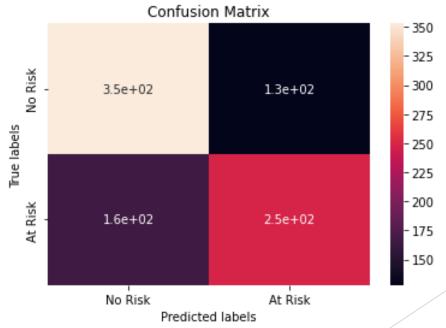
### **LOGISTIC**

### **REGRESSION**

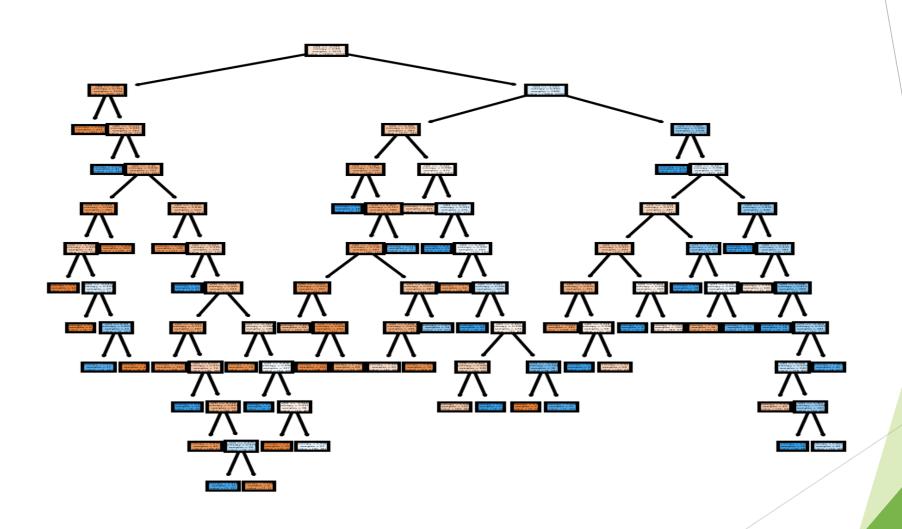
confusion matrix for train data

 $\frac{\text{confusion matrix for test}}{\text{data}}$ 



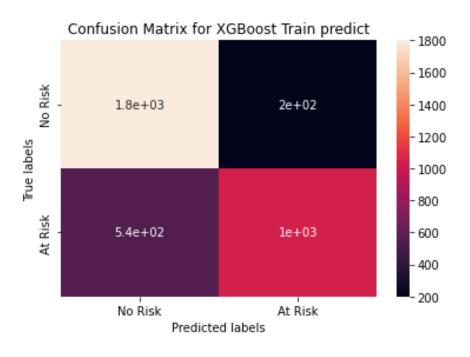


### **DECISION TREE CLASSIFIER**

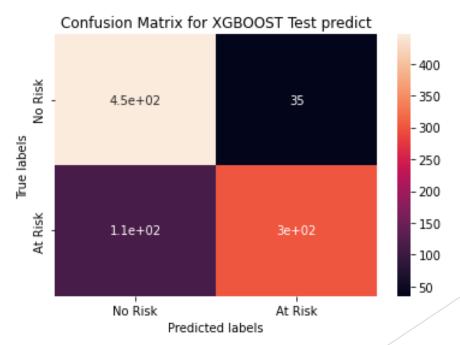


### **XGBOOST**

# CONFUSION matrix for Xgboost(train)



# Confusion matrix for xgboost (test)



### By selecting alla the accuracies of the model gradientboosting classifier appears to be best accuracy

```
The accuracy of KNN: 0.86
The accuracy of Logistic regression: 0.68
the acciracy of decision tree: 0.79
The accuracy of randomforestclassifier: 0.
88
the accuracy of XGBoost: 0.83
The accuiracy of gradientboostingclassifier
 : 0.89
```

# THANK YOU