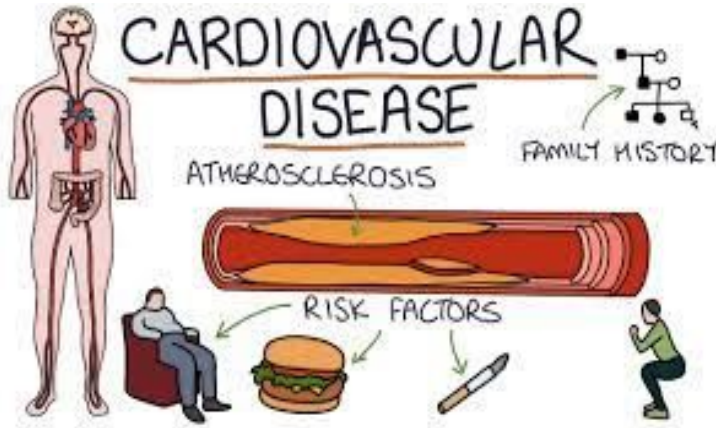


# Capstone Project

## Cardiovascular Risk Prediction

# Problem Statement:

The dataset is from an ongoing cardiovascular study on residents of the town of Framingham, Massachusetts. The classification goal is to predict whether the patient has a 10-year risk of future coronary heart disease (CHD). The dataset provides the patients' information. It includes over 4,000 records and 15 attributes.



# Key Steps:

- Defining the problem statement
- Data Cleaning
- EDA and data visualization
- Data preprocessing
- Feature selection
- Preparing Dataset for model
- Applying model
- Model validation and selection

Key steps



# Why is predictive analytics useful for Cardiovascular risk ?

- To know which patients are in risk:
- To know which disease lead to Cardiovascular risk:
- To know what habit lead to Cardiovascular risk:
- To know what should be BMI, BP, Diabetes and Cholesterol level:

# Dataset:

Rows : 3390

Columns : 17

	id	age	education	sex	is_smoking	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate	glucose	TenYearCHD
0	0	64	2.0	F	YES	3.0	0.0	0	0	0	221.0	148.0	85.0	NaN	90.0	80.0	1
1	1	36	4.0	M	NO	0.0	0.0	0	1	0	212.0	168.0	98.0	29.77	72.0	75.0	0
2	2	46	1.0	F	YES	10.0	0.0	0	0	0	250.0	116.0	71.0	20.35	88.0	94.0	0
3	3	50	1.0	M	YES	20.0	0.0	0	1	0	233.0	158.0	88.0	28.26	68.0	94.0	1
4	4	64	1.0	F	YES	30.0	0.0	0	0	0	241.0	136.5	85.0	26.42	70.0	77.0	0

# Variable Names:

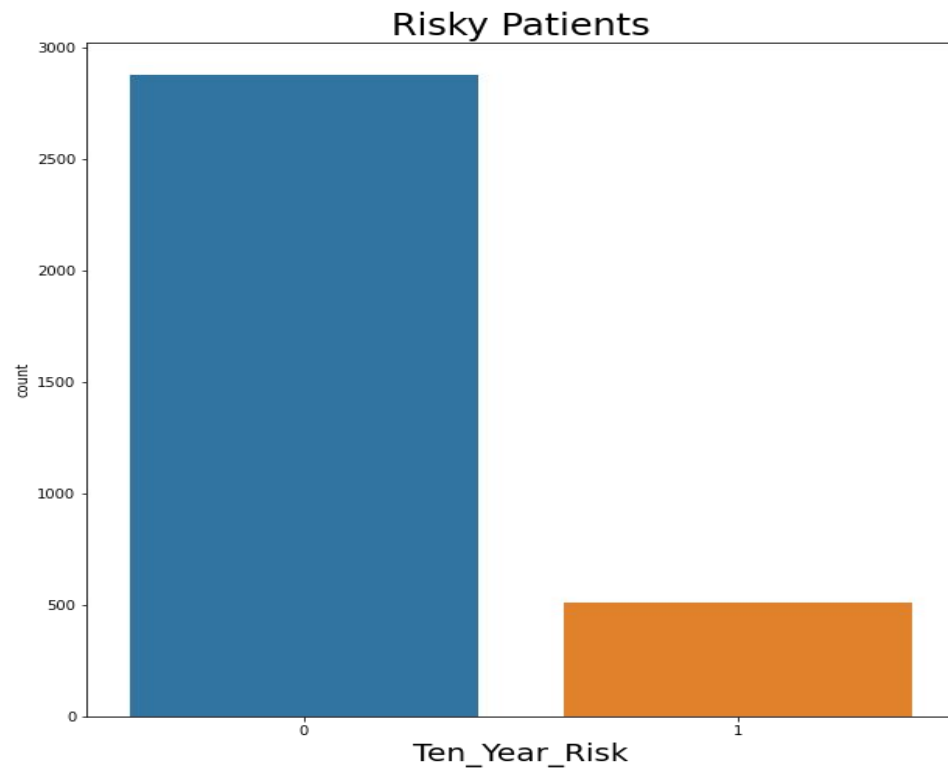
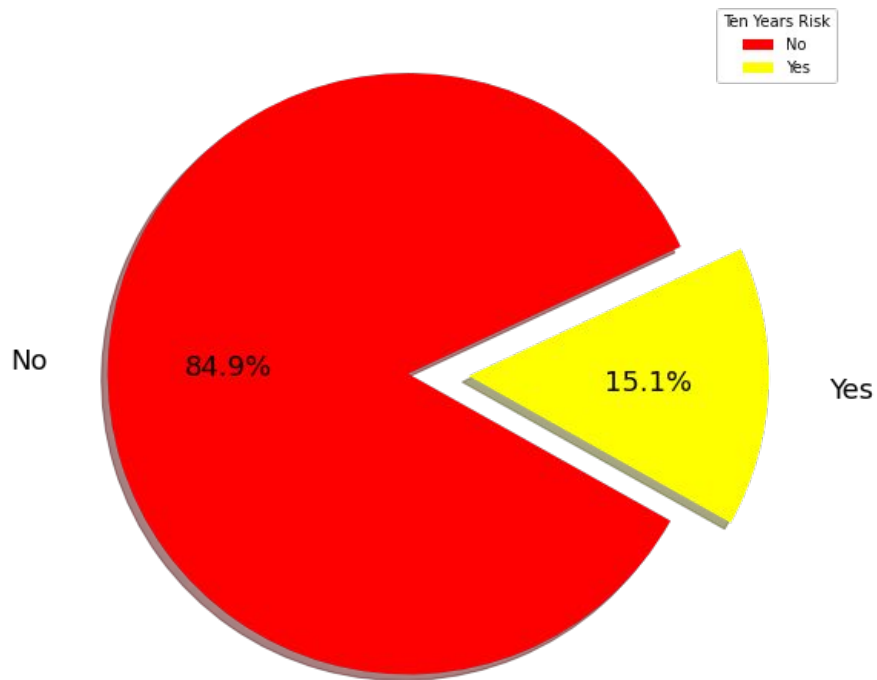
- Sex:
- Age:
- is\_smoking:
- Cigs Per Day:
- BP Meds:
- Prevalent Stroke:
- Prevalent Hyp:
- Diabetes:
- Tot Chol:
- Sys BP:
- Dia BP:
- BMI:
- Heart Rate:
- Glucose:
- 10-year risk:

# Exploratory Data Analysis:

EDA is used for analyzing what the data can tell us before the modeling or by applying any set of instructions/code.

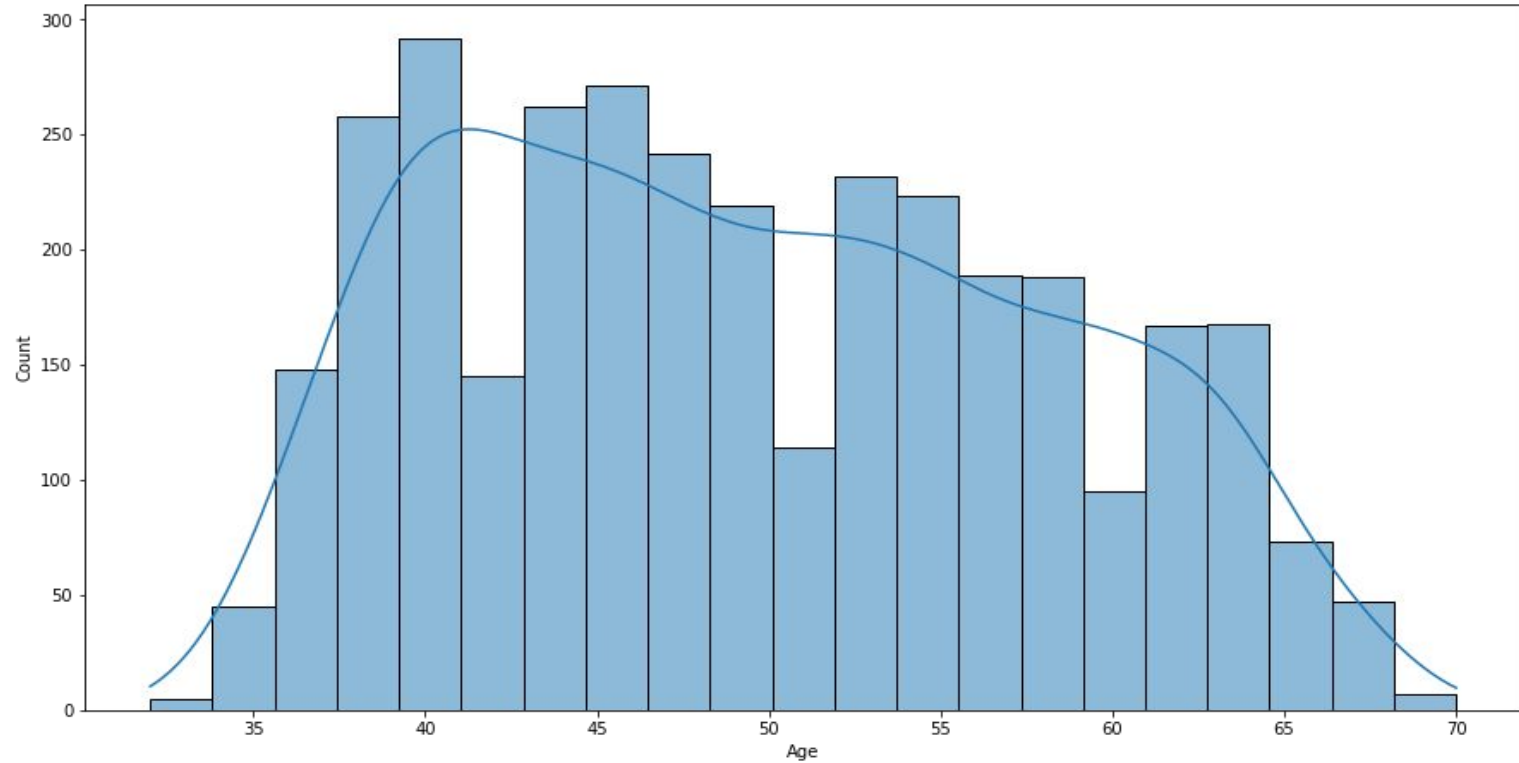


# Risky Patients:

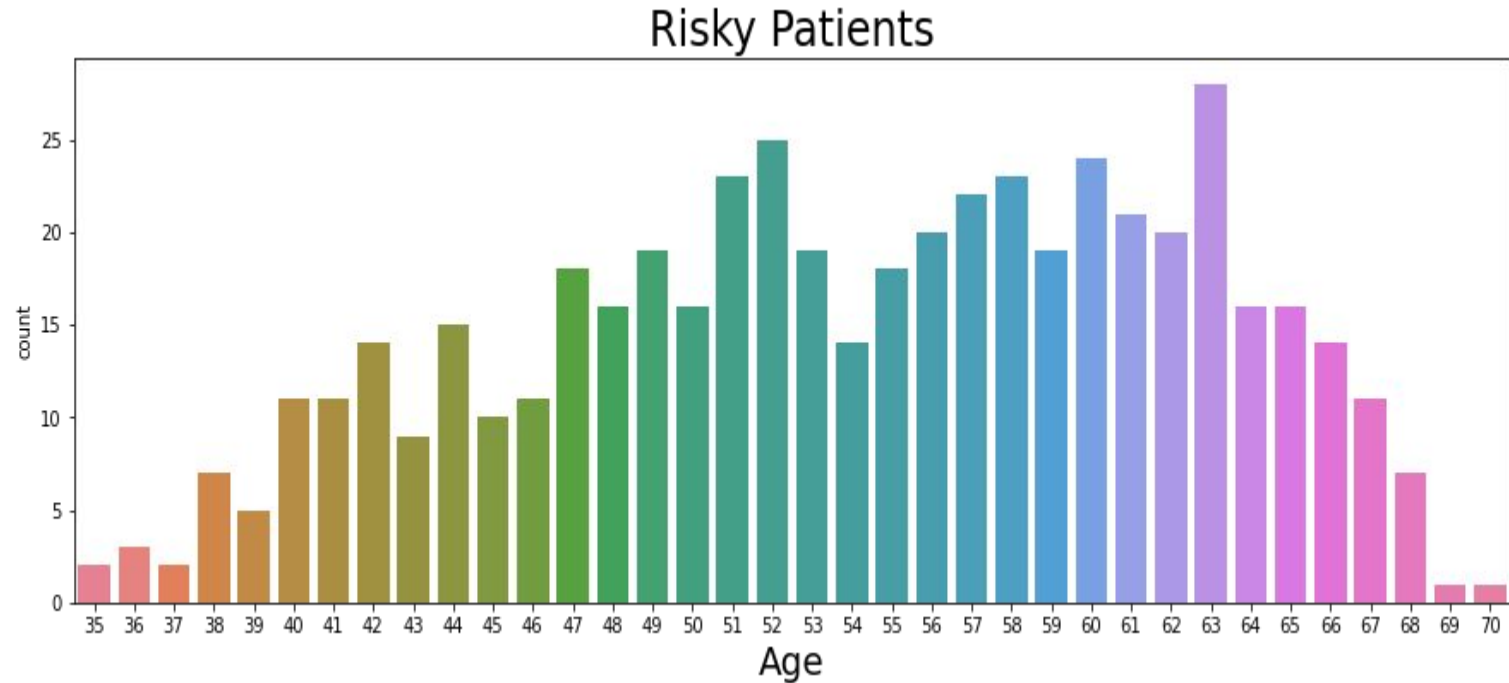




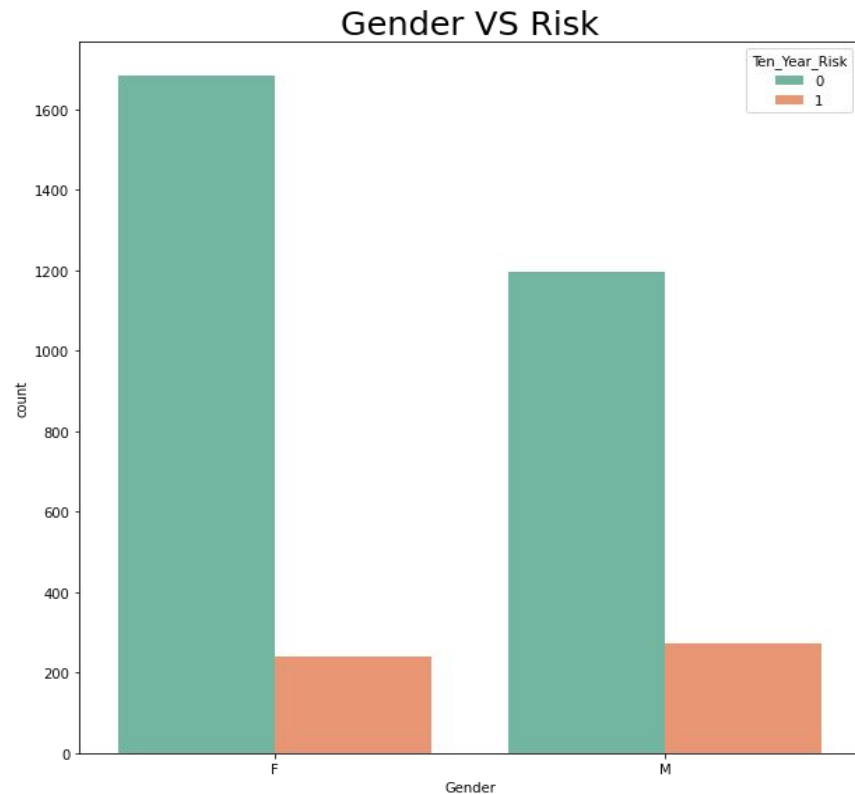
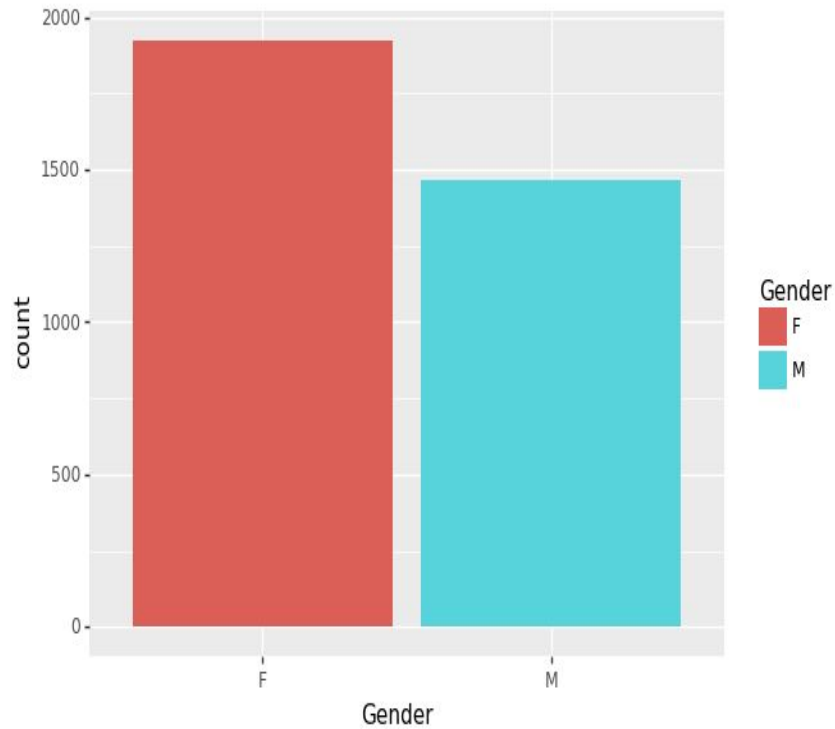
# Distribution of age:



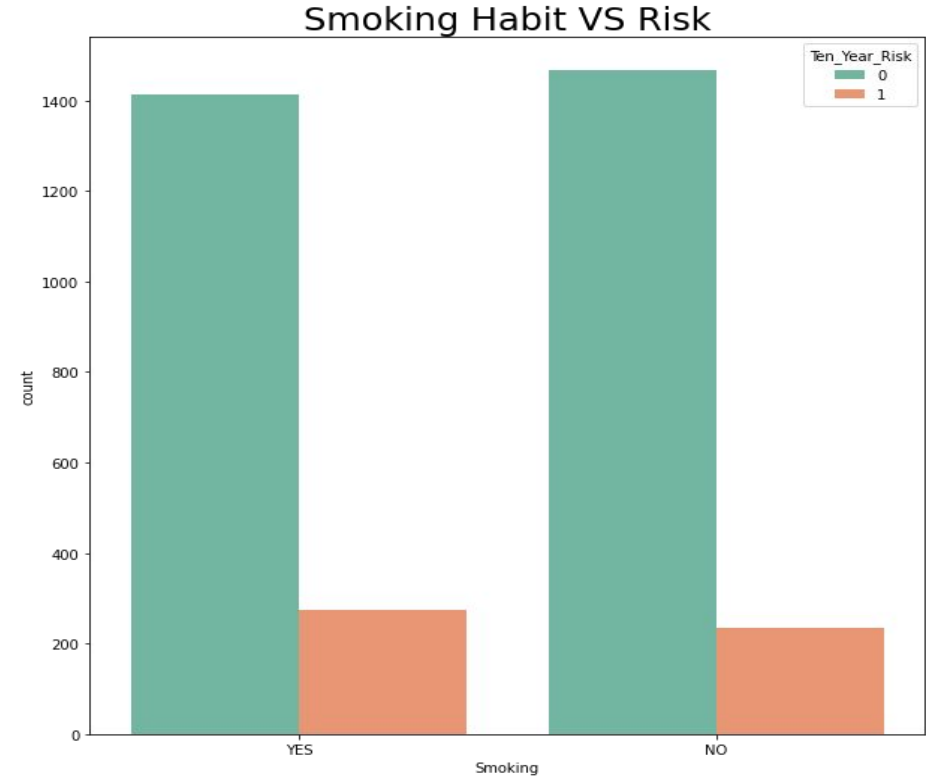
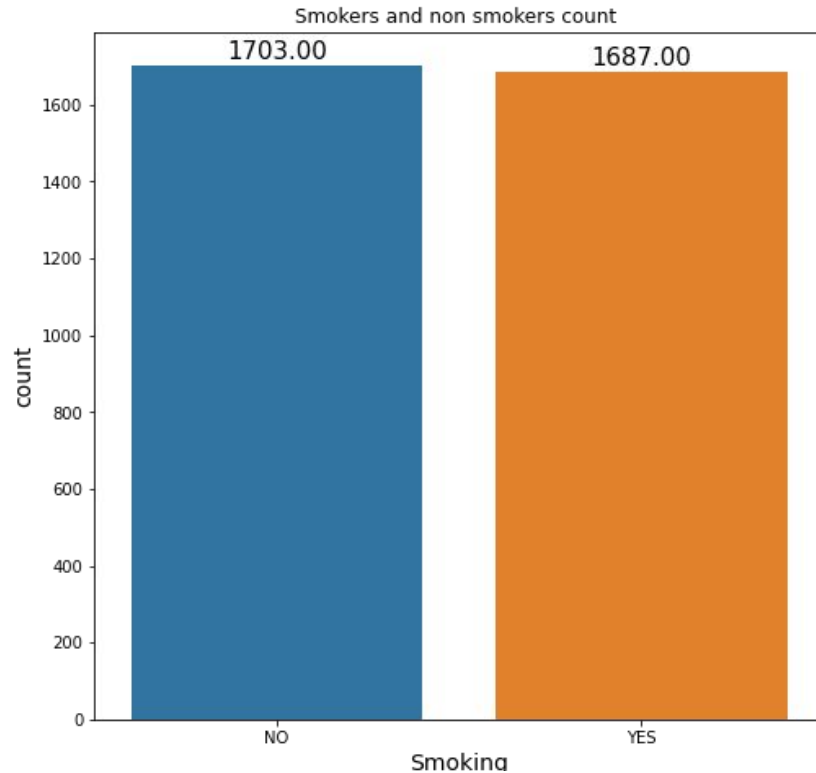
# Risky Patients With Respect To Age:



# Risk with respect to gender:

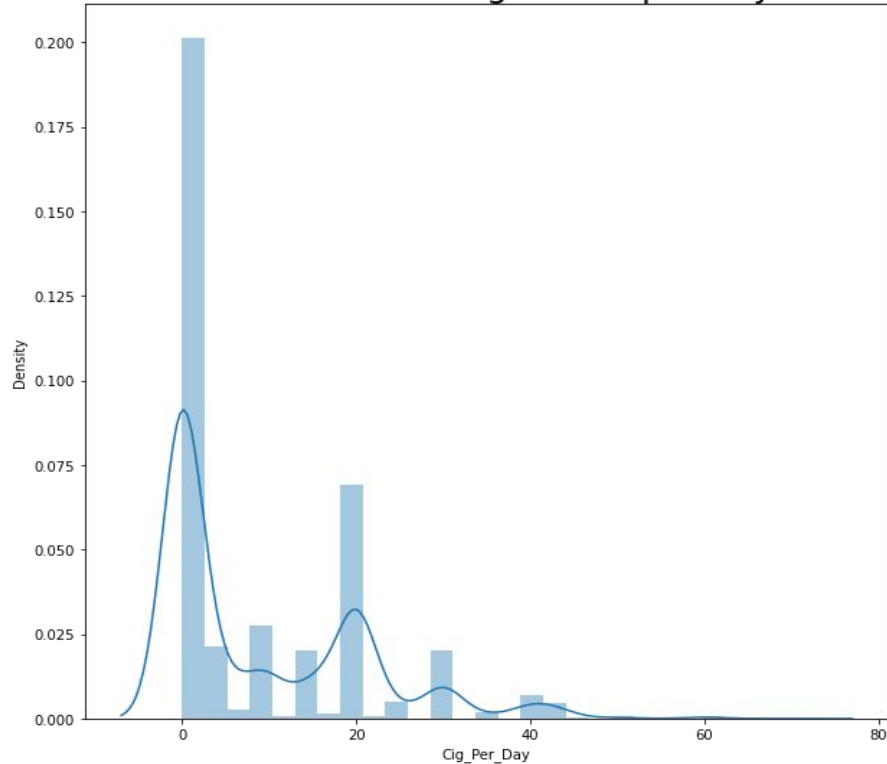


# Risk With Respect To Smoking Habit:

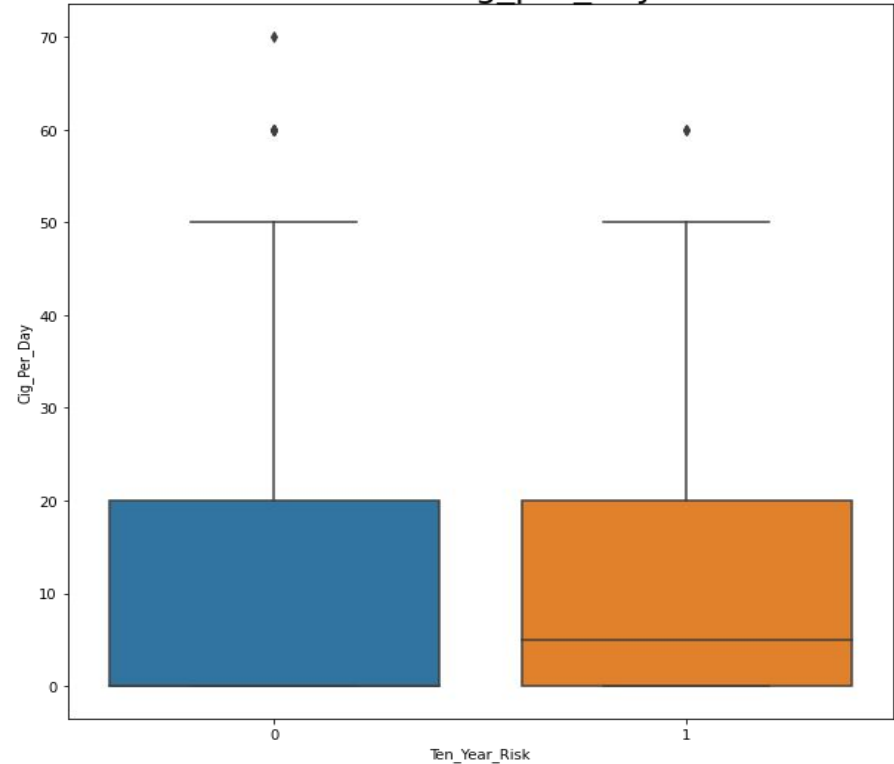


# Risk With Respect To Cigarettes Per Day:

Distribution of cigarettes per day

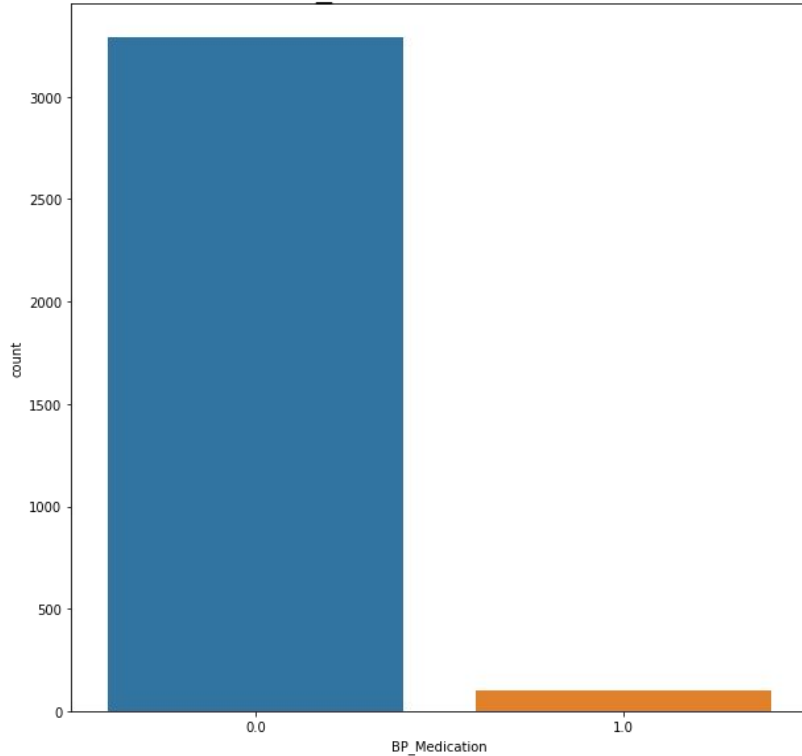


Risk vs Cig\_per\_day

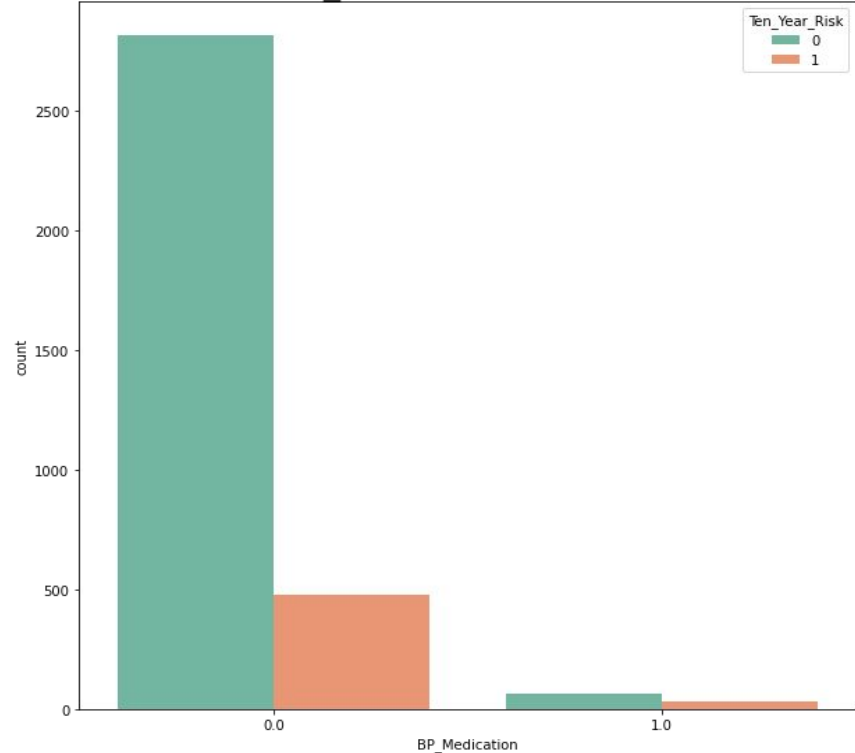


# Risk With Respect To BP Medication:

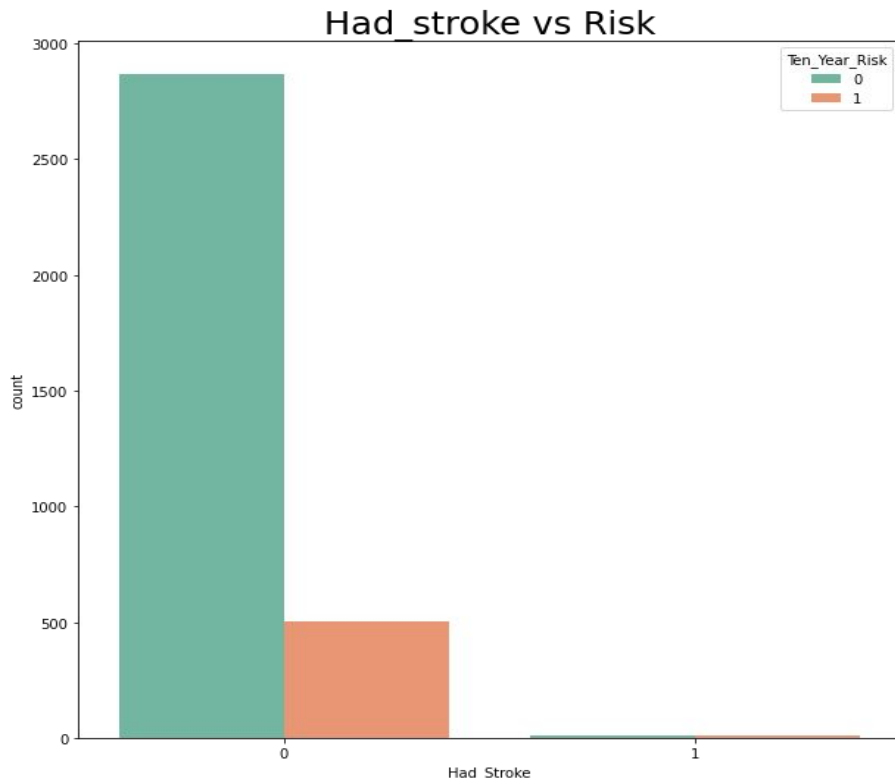
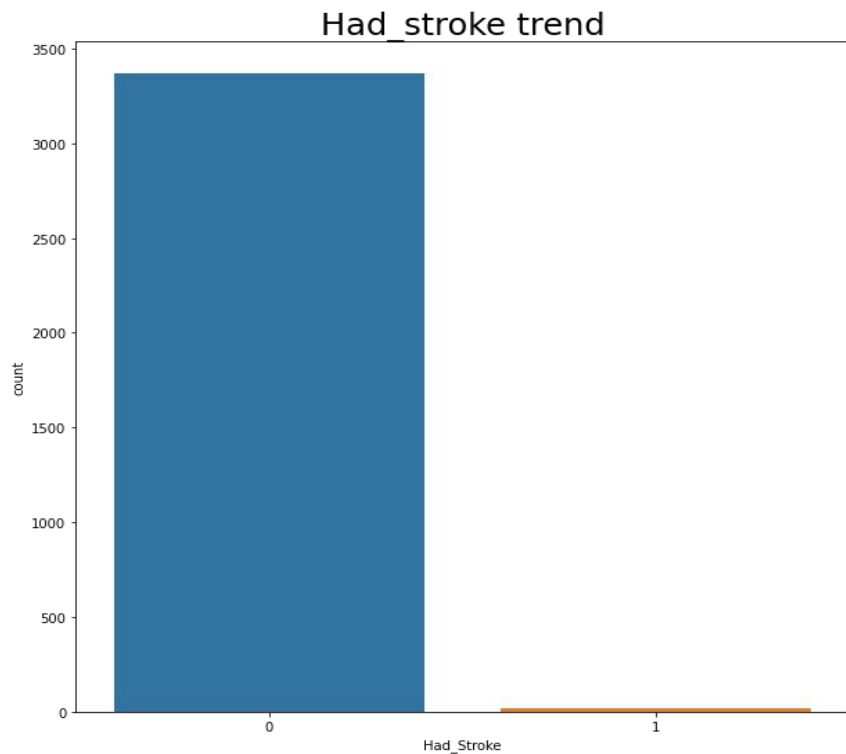
BP\_medication trend



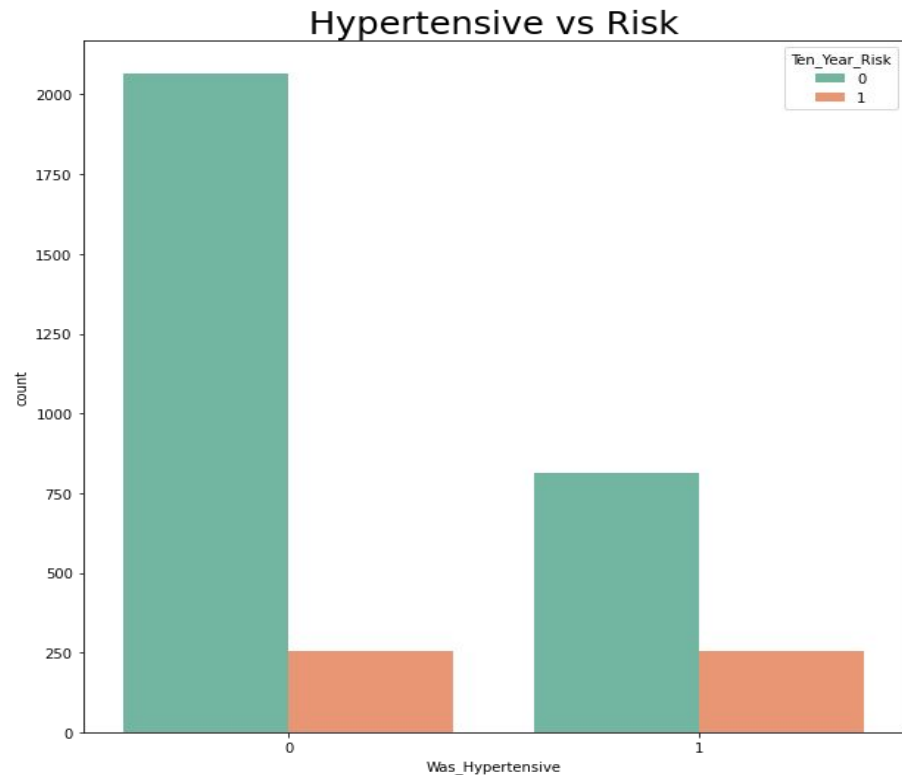
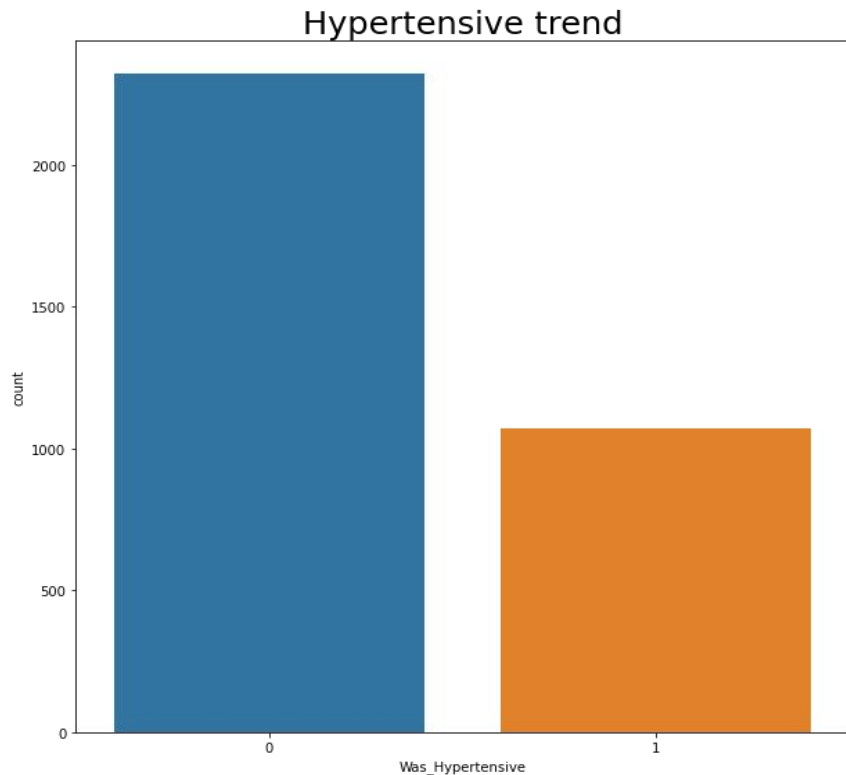
BP\_medication vs Risk



# Risk With Respect To Stroke:

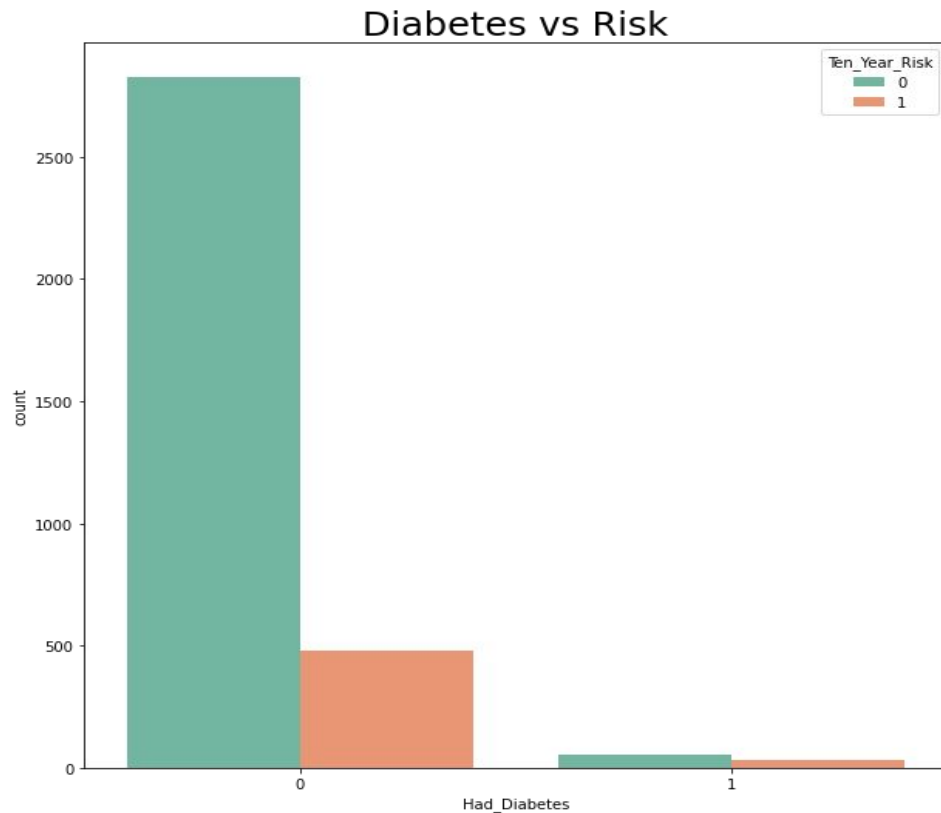
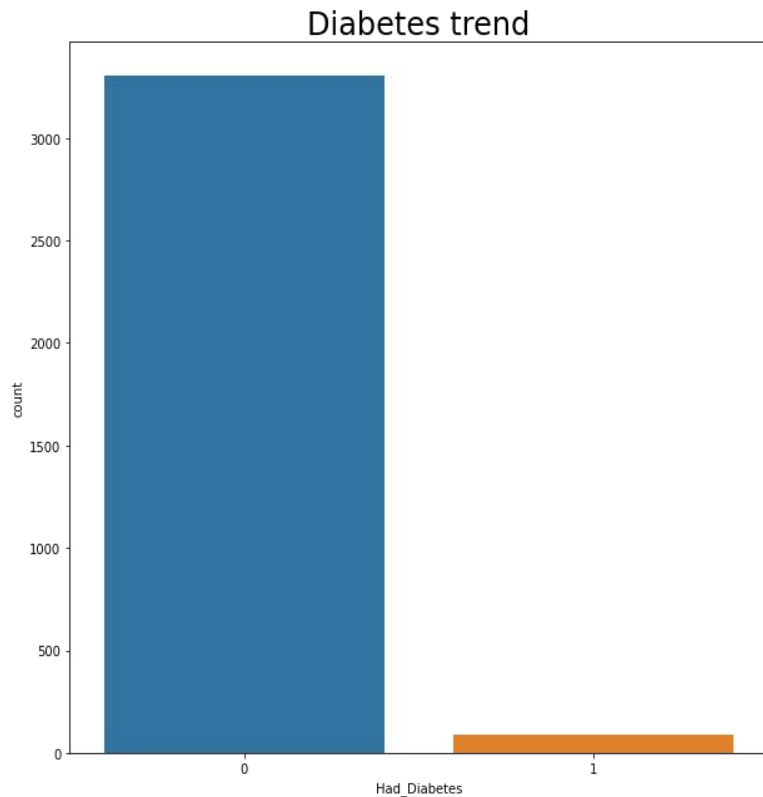


# Risk With Respect To Hypertensive:

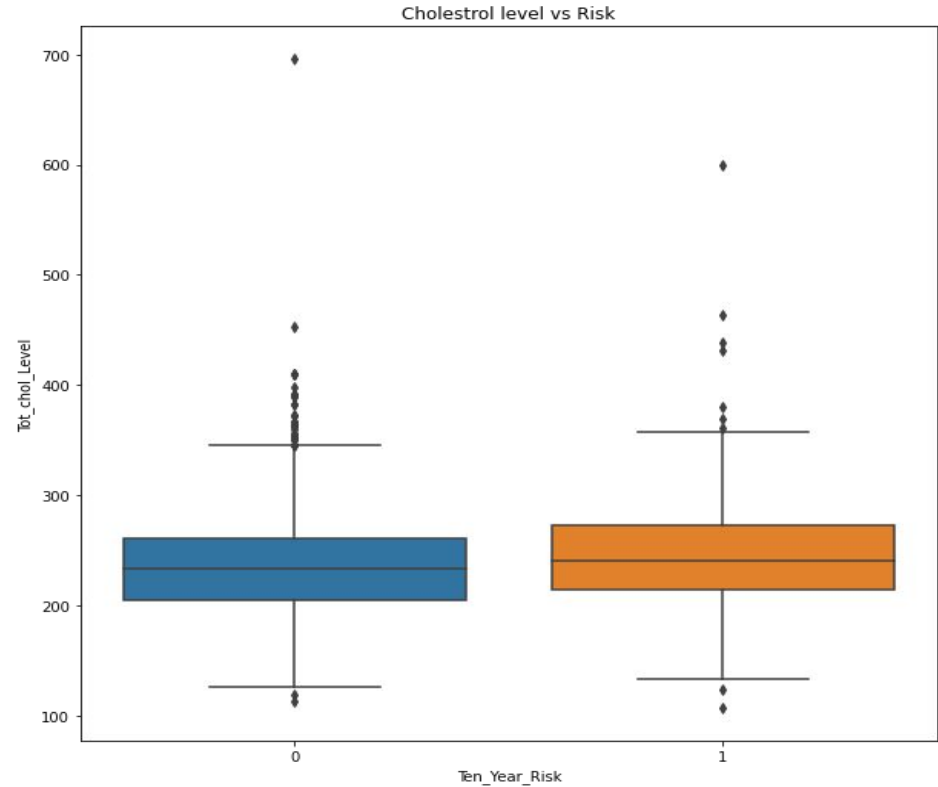
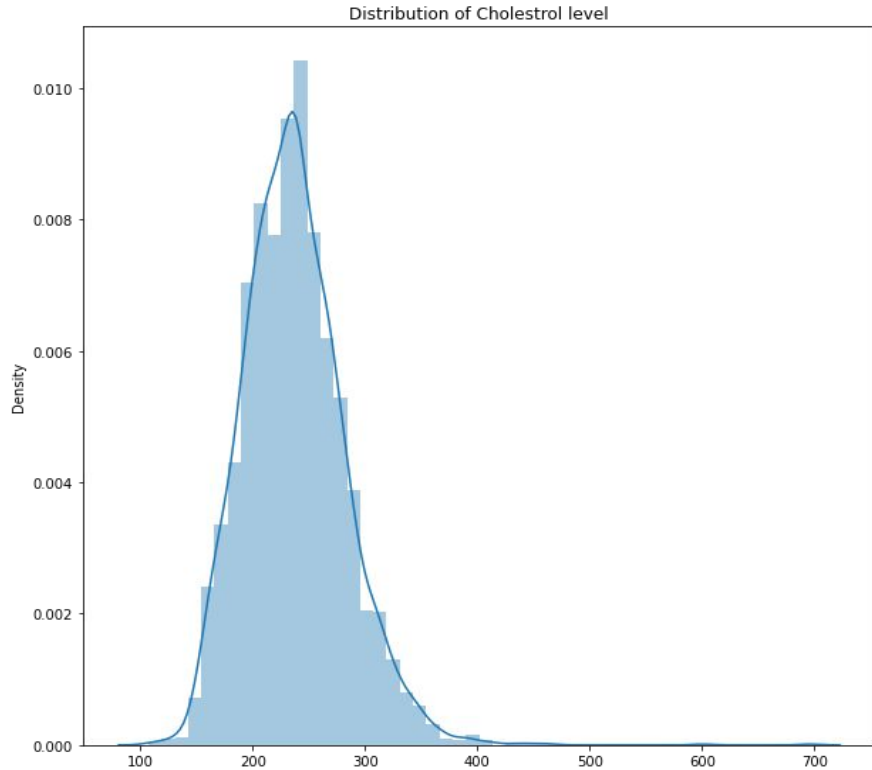




# Risk With Respect To Diabetes:

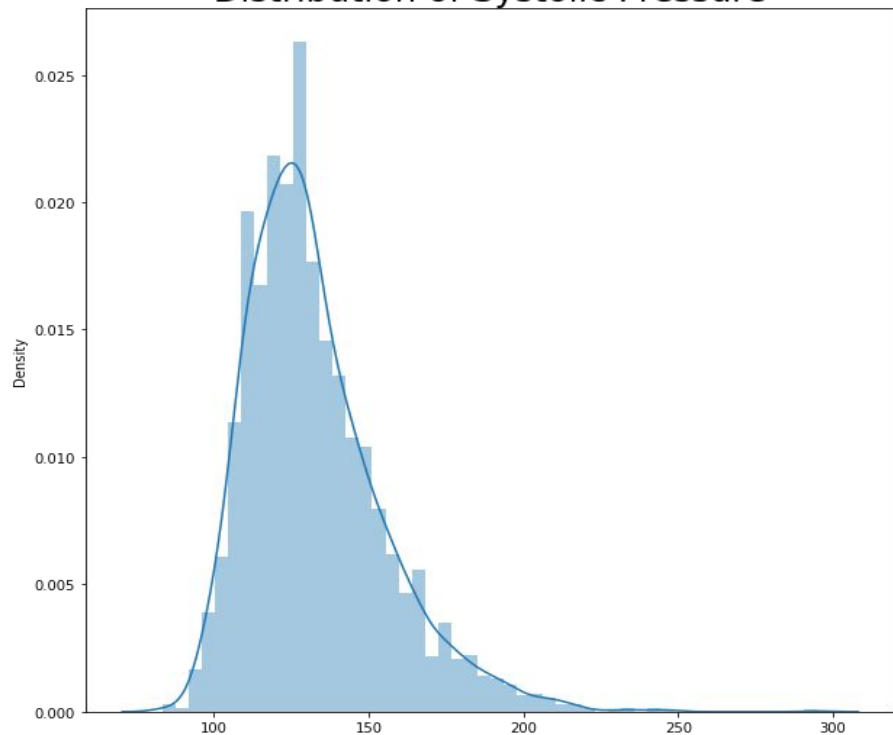


# Risk With Respect To Cholesterol level:

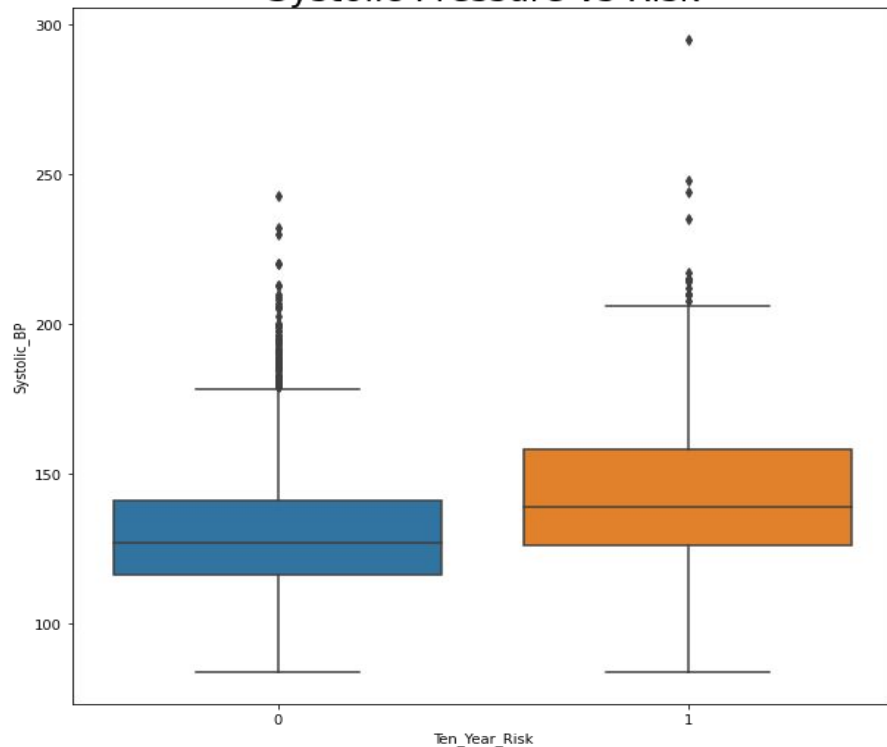


# Risk With Respect To Systolic Pressure:

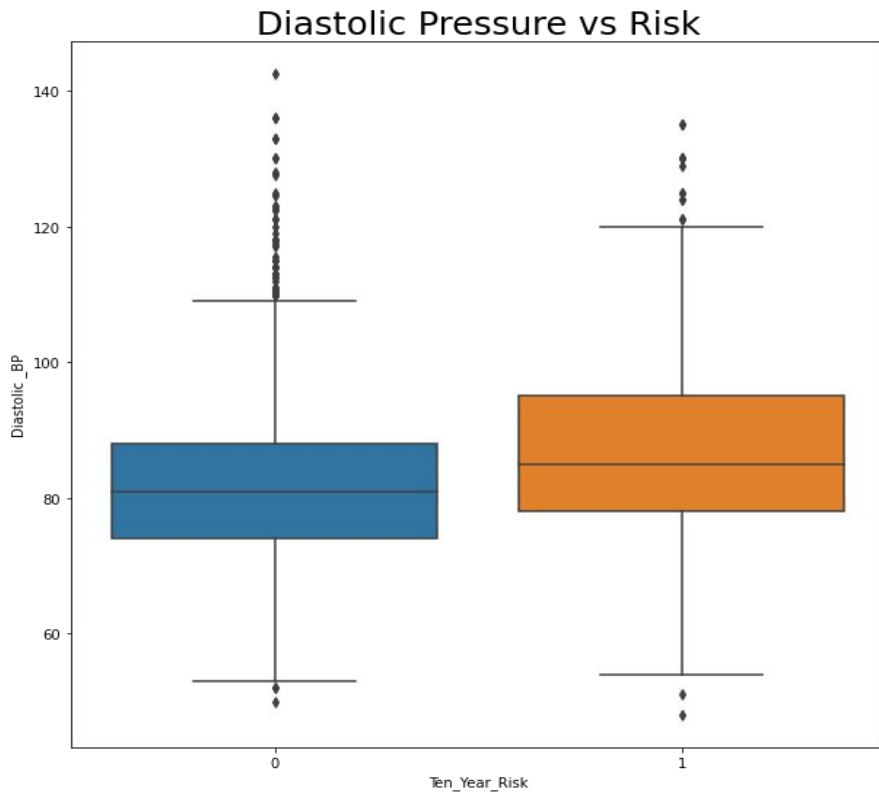
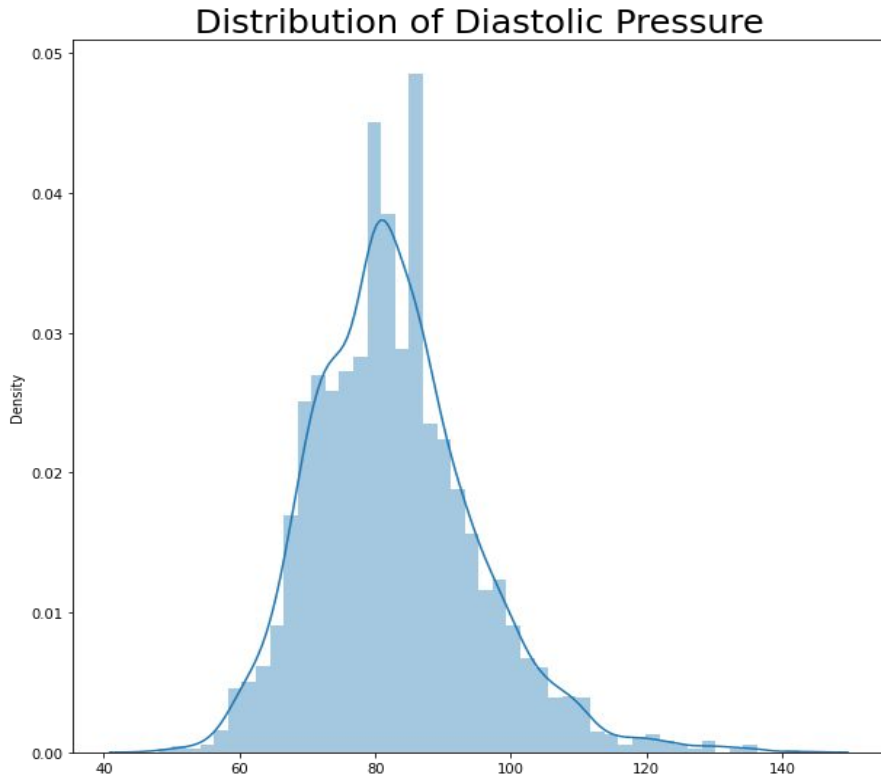
Distribution of Systolic Pressure



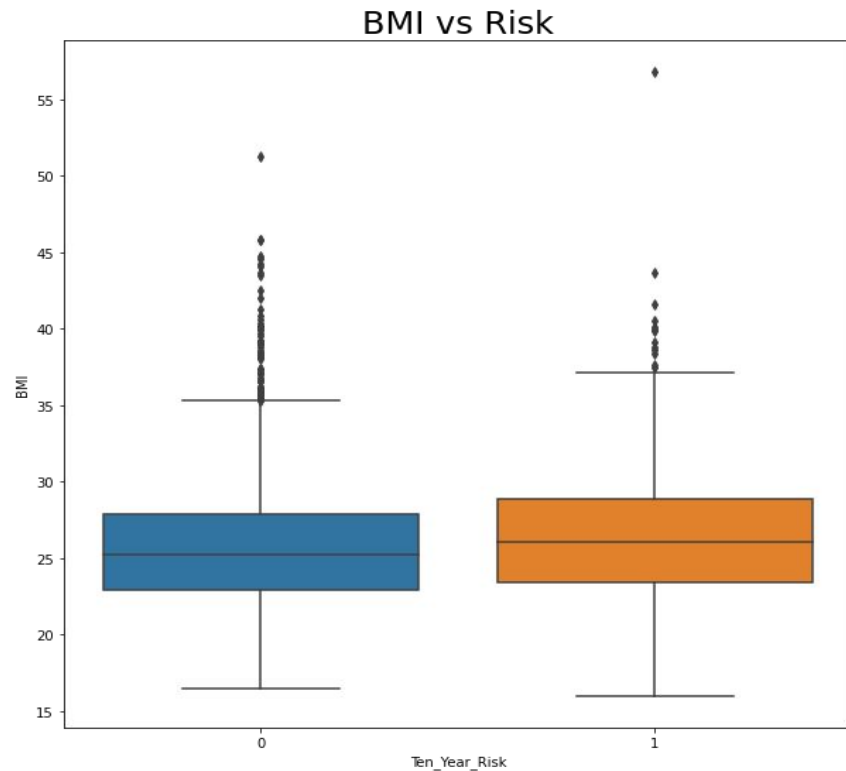
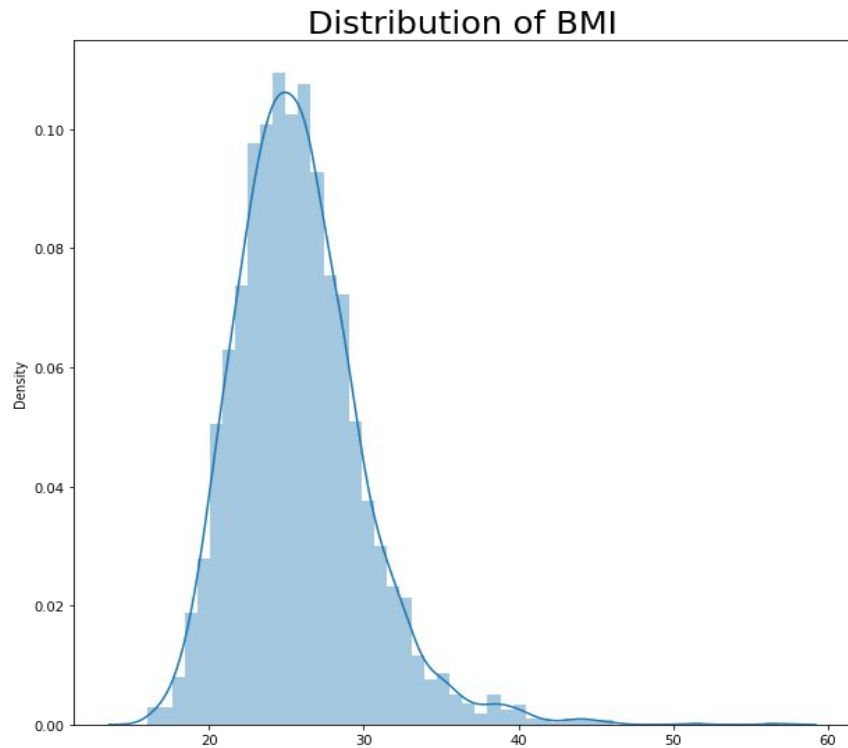
Systolic Pressure vs Risk



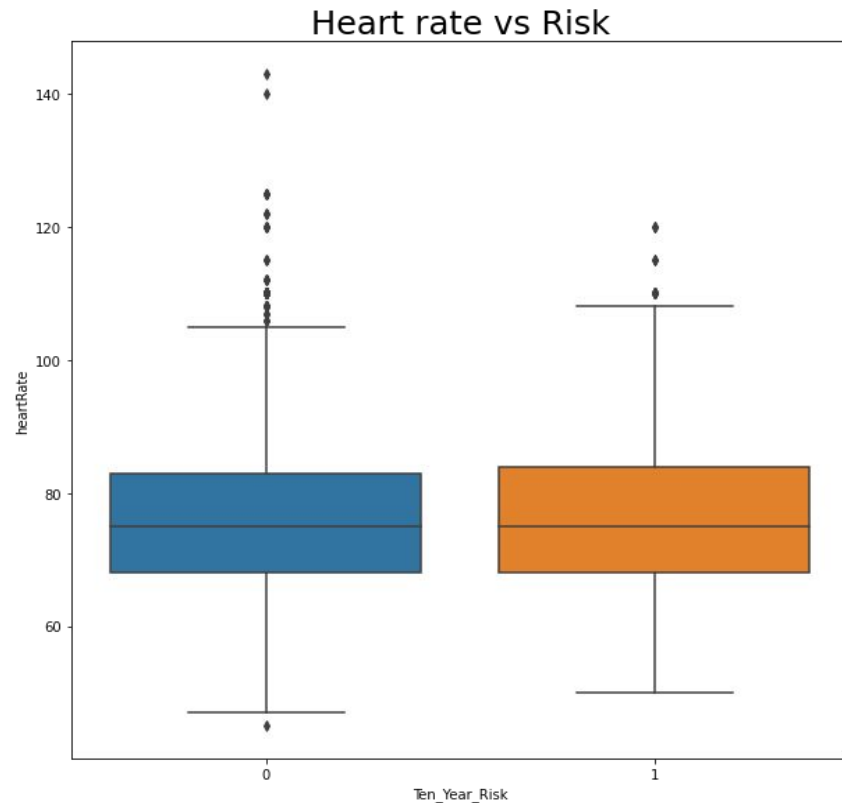
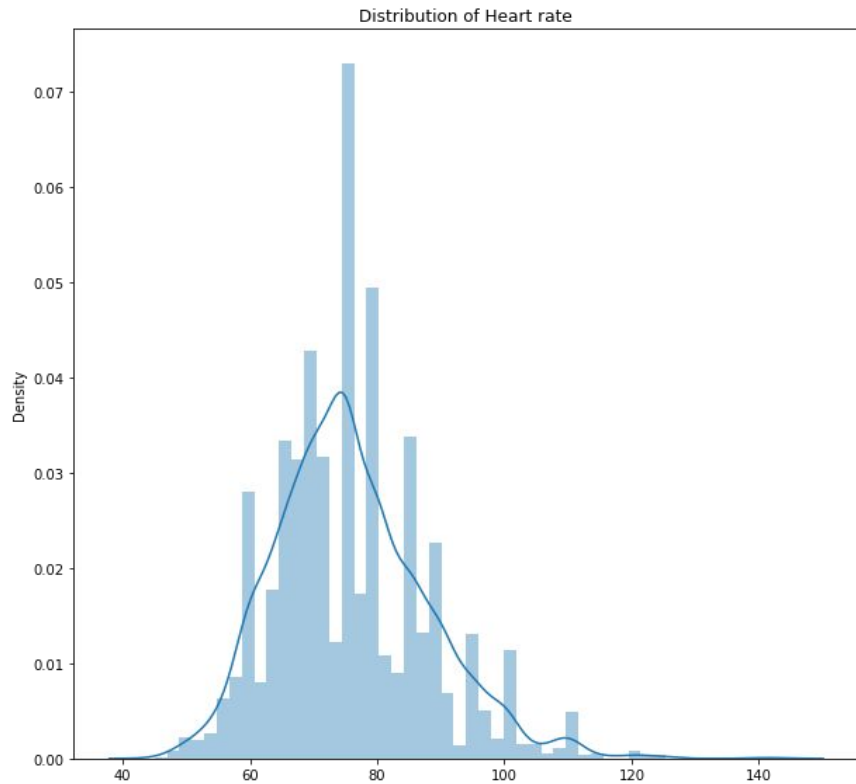
# Risk With Respect To Diastolic Pressure:



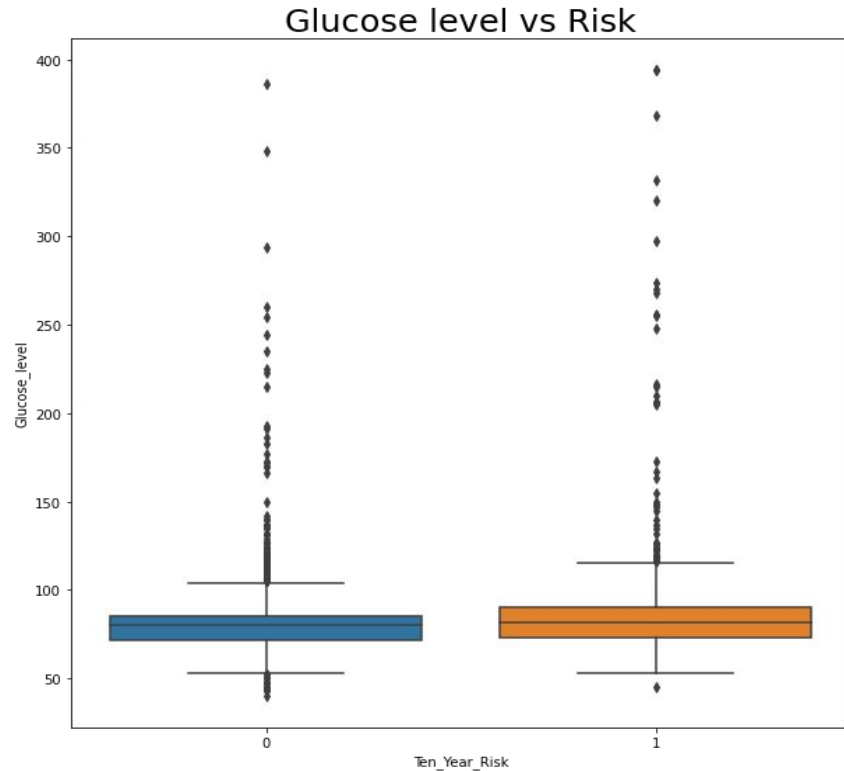
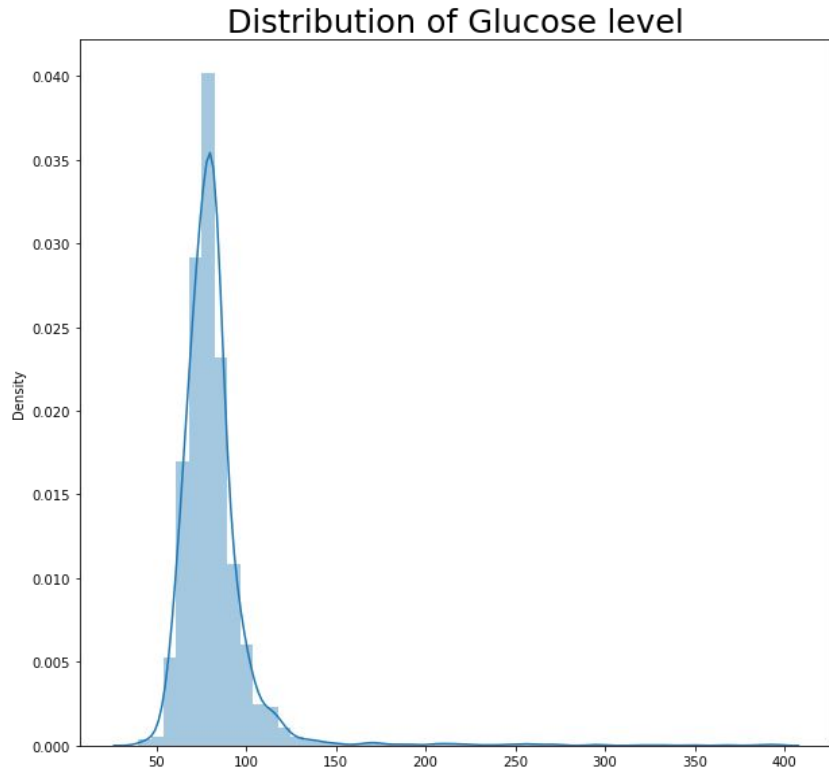
# Risk With Respect To BMI:



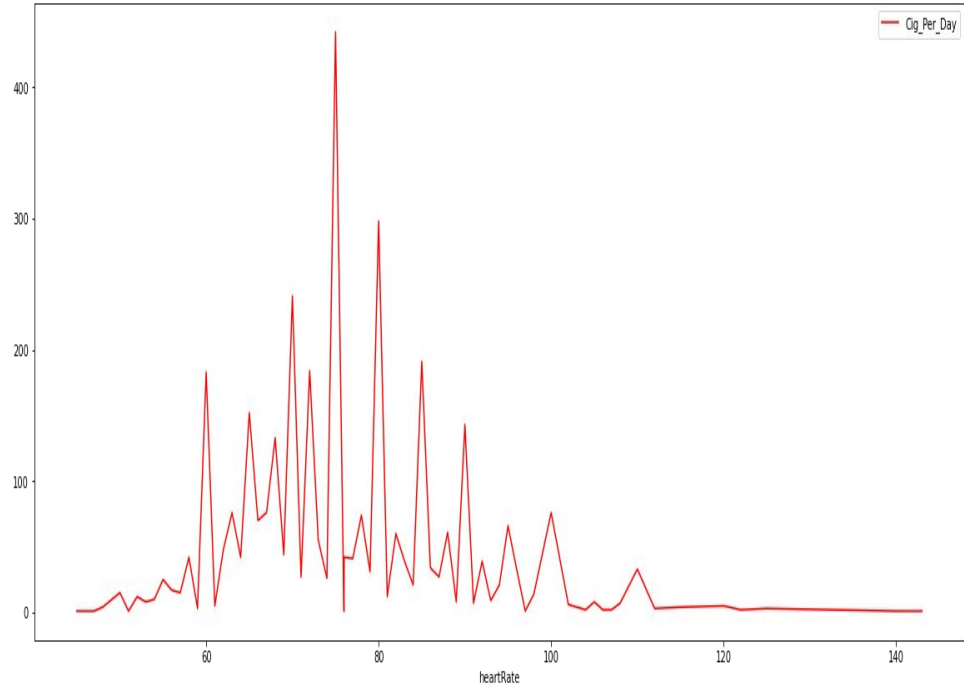
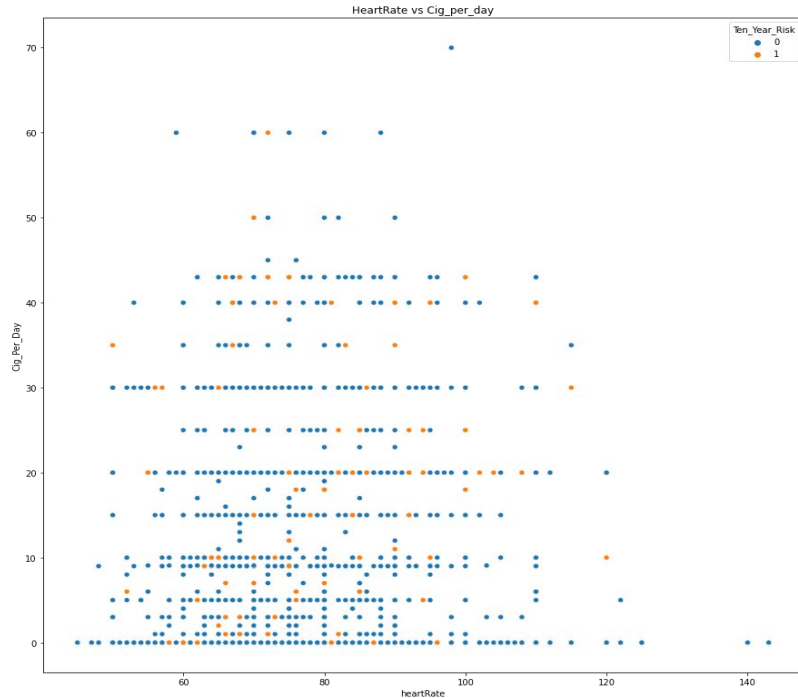
# Risk With Respect To Heart Rate:



# Risk With Respect To Glucose Level:

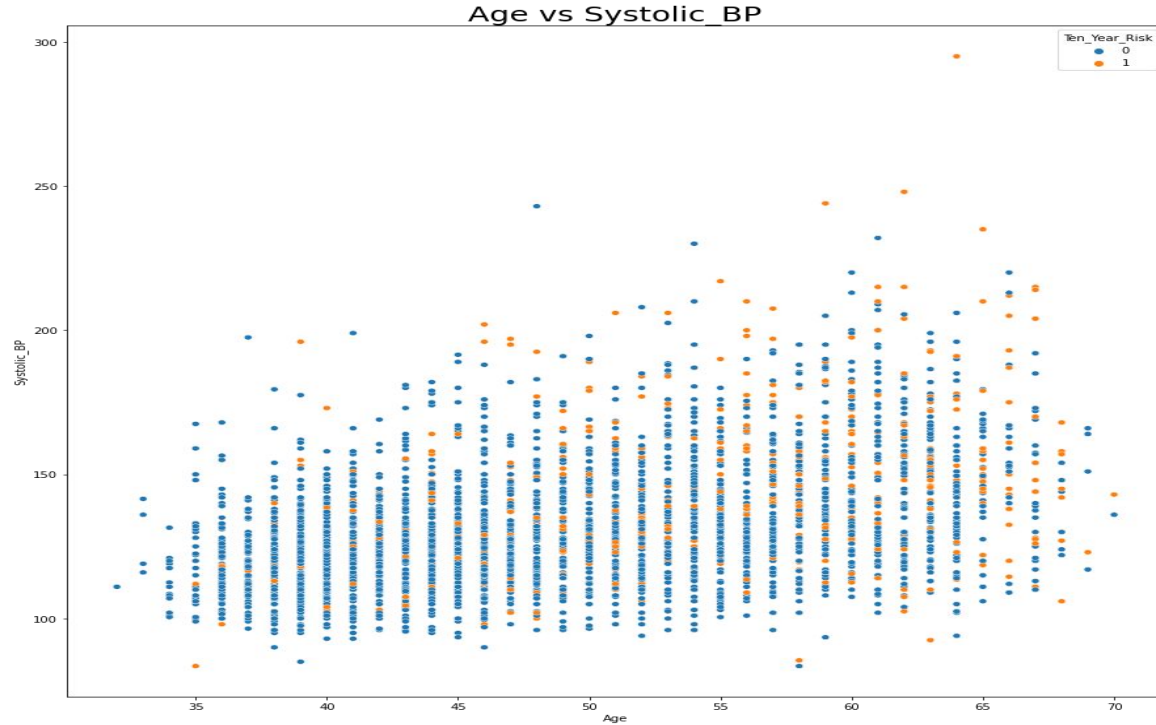


# Heart Rate With Respect To Cigarettes Per Day:

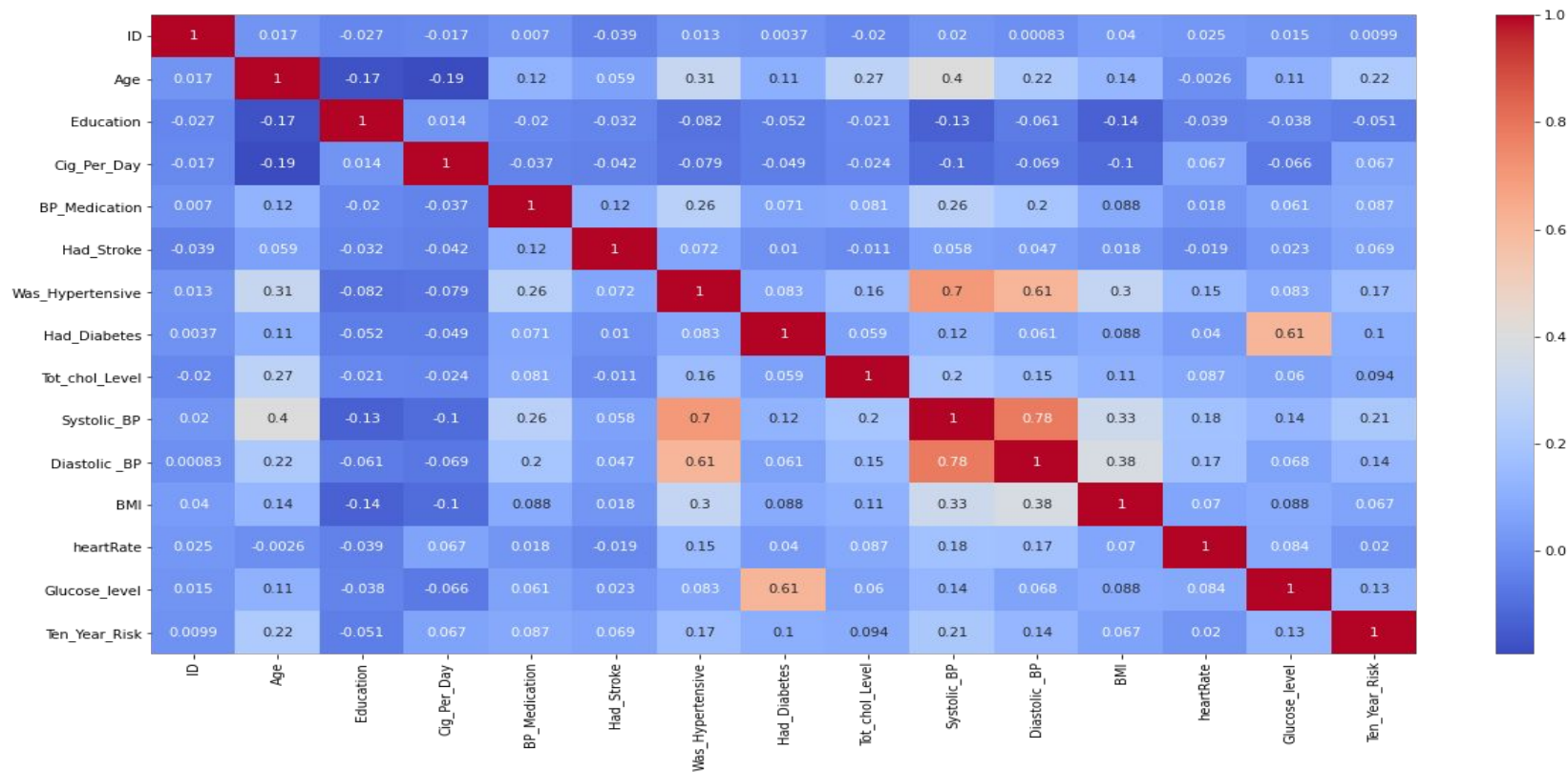




# Systolic BP With Respect To Age:



# Correlation Matrix:



# Data Preprocessing:

Data After Converting to Numerical :

Rows: 3390

Columns: 18

Which columns we have converted ?

- Smoking
- Gender

## Models Used:

1. Logistic Regression
2. Support Vector Machine (SVM)
3. SVM with Linear Kernel
4. SVM with Polynomial Kernel
5. SVM with Gaussian Radial Base Kernel
6. Neural Networks
7. Random Forest Classifier
8. XGBoost

# Model Validation and Selection:

## Observation 1:

At first I've tried Logistic Regression but scores were not that satisfying.

## Observation 2:

Then I've tried using Support Vector Machine, i got good scores as compared to logistic regression. Then I used SVM with Kernel Tricks I have used Linear kernel, Polynomial Kernel and Gaussian Radial Base Kernel I got good scores with polynomial kernel and radial base kernel.

## Observation 3:

Then I've used Neural networks with one hidden layer but we did not got good scores.

## Observation 4:

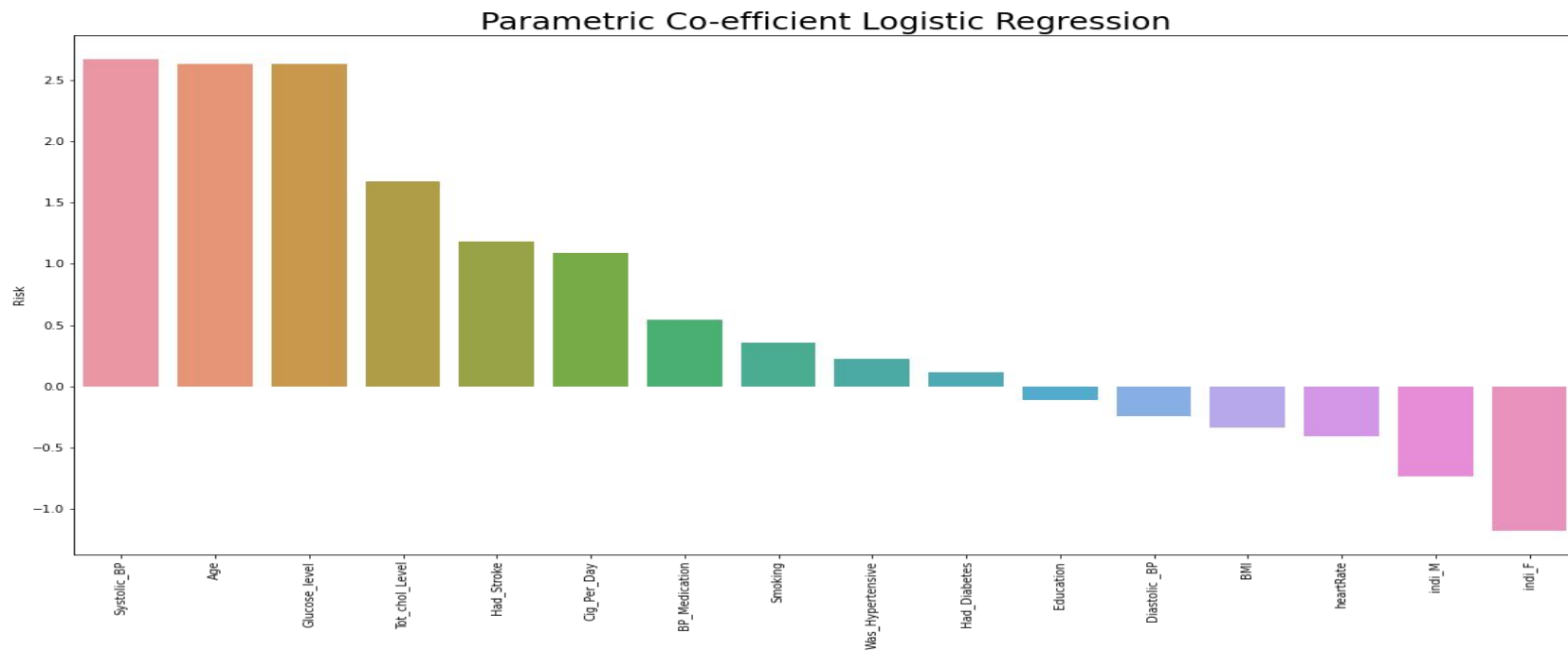
And lastly I've used Ensemble Learning models like Random Forest Classifier and XGBoost Classifier and both algorithms performed really well as compared to other models and with XGBoost I got the best Scores, so my optimal model is XGBoost Classifier.

# Model Validation and Selection:

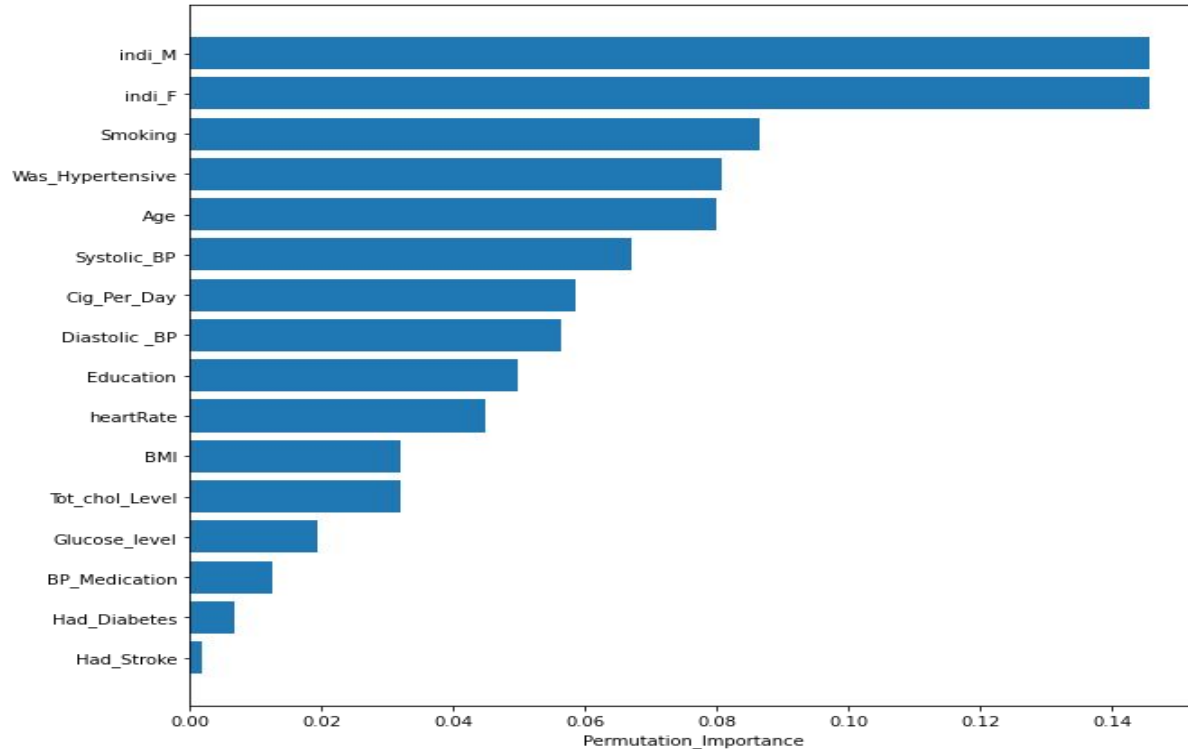
As XGBoost had performed really well so the best hyperparameters we got are:

```
'colsample_bytree': 0.7  
'learning_rate': 0.03  
'max_depth': 7  
'min_child_weight': 4  
'n_estimators': 500  
'nthread': 4  
'objective': 'reg:linear'  
'silent': 1  
'subsample': 0.7
```

# Feature Importance for Logistic Regression:

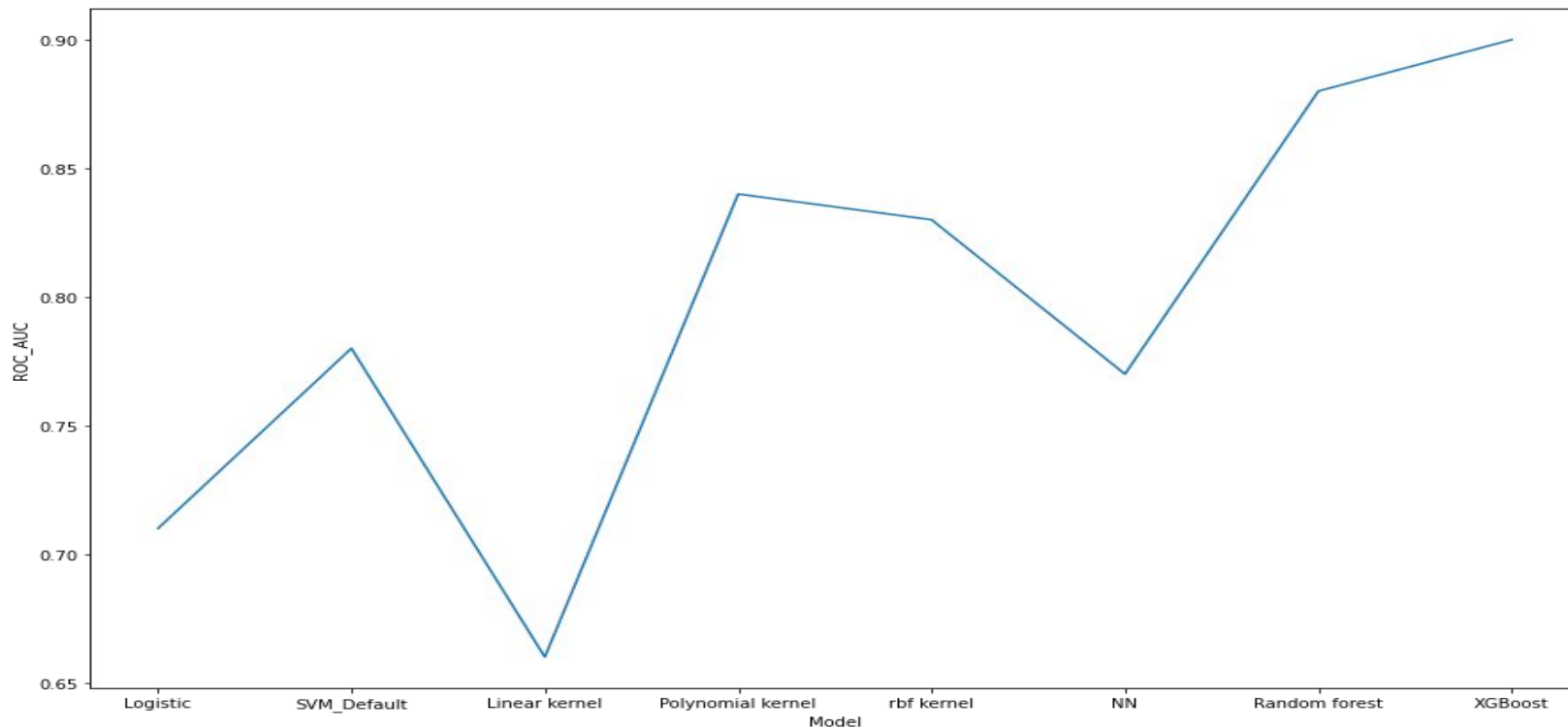


# Feature Importance for Support Vector Machine:





# ROC AUC Comparison:



# Challenges:

- Execution takes time.
- As there were many null values present in data set it took time to clean the dataset.
- Difficulty in selecting the appropriate graph for trend.



# Conclusion:



- There are 15.1 % people in our dataset are is risk for cardiovascular disease and 84.9 % people are safe (ten year risk).
- There is more risk of cardiovascular disease in patients of age between 51 to 63.
- The count of male and female are same in risk which is around 200, though females are more than males in our dataset.
- Around 250 smokers are in risk and around 210 non-smokers are is risk for cardiovascular disease.
- We can't evidentially state smoking will lead to heart disease, as we seen from count plot there is no huge difference between these to commune and also our extreme smoker who smokes 70 cigarettes per day is not having ten year risk.
- There are very few people who are done with BP medication which are around 200 but many people have not taken any BP medication and they are around 3200. We cannot say that after taking medication person is safe.
- Around 500 patients who did not had stroke yet and are at risk and around 2800 patients are safe.
- around 250 people with hypertensive are in risk and around 255 people with no hypertensive are at risk.
- Here we can see people who did not had diabetes are more and around 500 people who did not had diabetes are at risk. And there are very few people who had diabetes are at risk.
- most of the people who are not in risk their Cholesterol level lies between 210 to 280 and people who are in risk their cholestrol level lies between 215 to 285 there in not huge difference it is quite normal.
- most people who are not in risk their systolic BP lies between 110 to 140 and people who are at risk their systolic BP lies between 125 to 160. we can say people with high systolic BP are at risk.
- most people who are not in risk their diastolic BP lies between 75 to 85 and people who are at risk their diastolic BP lies between 89 to 90. we can say there is a slight increase in diastolic BP of people who are in risk.
- most people who are not in risk their BMI lies between 22 to 28 and people who are at risk their BMI lies between 23 to 29 approx. WE cannot see any difference BMI is approx. same of risky and not risky people.

# Conclusion:

- most people who are not in risk their heart rate lies between 68 to 83 and people who are at risk their heart rate lies between 68 to 84. which is same for risky and not risky people.
- there is not that difference between the glucose level of risky and non risky patients. glucose level lies between 70 to 80 for both risky and non risky patients.
- most people smoke cigarettes between 1 to 10 approx. and there heart rate lies between 60 to 100.
- age of people lies between 32 to 70 and most of the people's systolic BP lies between 90 to 200.
- With logistic regression we got the accuracy score of 0.68 on train data and 0.66 on test data.
- With Support vector machine we got the train accuracy score of 0.79 and test accuracy score of 0.78.
- With support vector linear kernel we got the train accuracy score of 0.67 and test accuracy score of 0.67.
- With support vector polynomial kernel we got the train accuracy score of 0.95 and test accuracy score of 0.84.
- With Gaussian radial base kernel we got the train accuracy score of 0.97 and test accuracy score of 0.83.
- With Neural Networks we got the Train Accuracy score of 0.81 and test accuracy score of 0.77.
- With Random forest classifier we got the train accuracy of 0.97 and test accuracy of 0.89.
- With XGBoost classifier we got the train accuracy score of 0.99 and test accuracy of 0.90.

*Thank  
you!*