## **TECHNOLUTION ASSIGNMENT-DOCUMENT**

## 1.Introduction to the problem

The dataset shared herewith is from the informatics department of a hospital which has patient level vitals for each patient they have on ICU beds and the corresponding alerts based on the status of their experiencing a code-blue event (viz. "Coded" = "high-risk":1 or 'low-risk':0).

Now, predict whether code blue event occurs or not based on given patient vitals.

## **2.ALGORITHMS USED**

To solve this problem we could use classification based algorithms as the outcome is a classification problem where code blue event occurred or no(It is YES or NO type).

There are many classification algorithms. Inorder to solve this I used LogisticRegression, KNN Classification, DecisionTree based algorithms.

**LogisticRegression**: Is one of classification algorithm which is used to determine/predict output for discrete values with help of sigmod function.

SIGMOD FUNCTION:  $F(x)=1/1+e^{-(-x)}$ 

F(x)=wanted probability

X=input of the equation (eg: mx+b)

KNN-Classification: In this algorithm we calculate and set "K" value using elbow function and then calculate distances from each point using methods

**EUCLIDEAN DISTANCE:** : d=((x1-y1)^2+(x2-y2)^2))^1/2

MANHATTAN DISTANCE: d=|x1-y1|+|x2-y2|+....+|xn-yn|

**Decision Tree**:In this algorithm we making decisions based on splitting nodes into multiple sub nodes.

Split uses following algorithms:

**Information Gain: (1-entropy)** here we consider node which gives highest gain.

**Gini Impurity: (1-Gini)** we calculate gini impurity of each split and select node with low gini impurity.

#### 3. EXPERIMENTAL ENVIRONMENT:

I've used Jupyter notebook from anaconda which is a free and open-source distribution of python, R programming languages for scientific computing. In jupyter note book

I have installed all required libraries which are imported as of our model requirement.

Libraries like pandas ,numpy are imported as they are used for data extraction and data manipulation .

Matplotlib package is a visualization tool used to plot graphs for understanding the relation between various attributes.

Scikit-learn package is installed in order to import wanted libraries of various pre implemented algorithm methods.

### 4. Brief introduction to dataset used

Brief introduction to dataset used At first step we assign attributes dependent and independent variables .

vitals\_datetime Time at which the patient got admitted to the hospital

heart rate Patient vital

respiration\_over\_impedence Patient vital

spirometry\_oxygen\_saturation Patient vital

pulse Patient vital

blood pressure systolic Patient vital

blood\_pressure\_diastolic Patient vital

blood\_pressure\_average Patient vital

patient id Unique-id of the patient

machine\_id Unique machine id (ICU bed ID)

Coded Actual alerts [high-risk/low-risk]

Then, we divide dataset into training and testing .. where train data is used to build the model and make predictions on the test data.

For further proceeding we choose **coded** as target variable.

After this we predict Accuracy which gives correctly predicted observations of total observations. DATA SET PREPARATION

## 1. Imputing missing values.

We can fill the missing values with help of computing the mean.

### 2. Removing categorical variables.

Sklearn cannot compute categorical variables like(machine\_id,patient\_id,vitals\_datetime) we need to assign numbers to each variable. We can use dummy encoding to achieve this.

DATASET IS CLEANED AND MANIPULATED (i.e imputed categorical variables and missing values)

#### 5. Performance metrics:

As above algorithms are classification based algorithms, we can generate classification report which contains following metrices:

**Accuracy:** It denotes how efficiently our model performing.(Model is good if accuracy is greater than 70%)

**Precision:** Tells us amount of meaningful information present among retrieved information.

Recall: Tells us amount of meaningful information retrieved

**F1score:** its is also measurement of accuracy.

## **LOGISTIC REGRESSION**

**KNN-Classification:** 

Accuracy:0.93

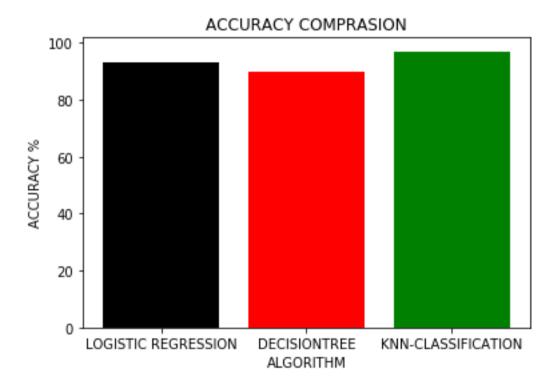
Accuracy:0.97

### **Decision Tree:**

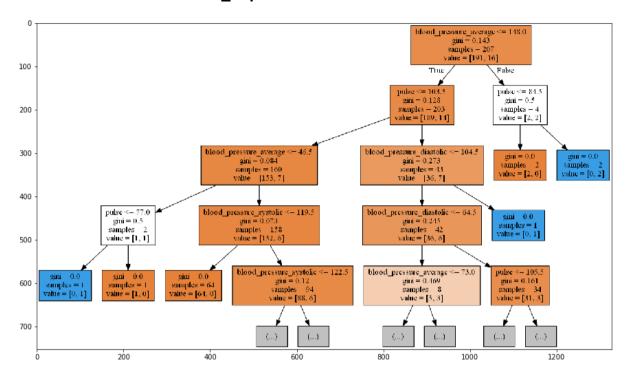
Accuracy:0.90

6.Comparison charts and observations

## **ACCURACY COMPRASION**



# **DECISION TREE where max depth = 4**



### **Observations:**

- 1. From above comparison charts we can notice that KNN-Classification predicts model better when compared to Logistic and DecisionTree.
- 2. As we see based on given patient vitals on testdata there is solid conclusion that we could predict if patient will undergo code blue event or not .
- 3. In decision tree as we change max\_depth the tree grows eventually. Sometimes decision tree can lead to overfitting.

## **Final comparison**

4. KNN CLASSIFICATION>LOGISTICREGRESSION>DECISIONTREE. According to above results.

SUBMITTED BY,

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