



Work Integrated Learning Programmes Division
M.Tech (AIML)
Machine Learning
S1-22 AIMLCZG565
Second Semester

Assignment 1 – PS-6

Heart attack possibility Dataset- [Weightage 10%]

Instructions for Assignment Evaluation

1. Please follow the naming convention as <Group no>_<Dataset name> ipynb.
Eg – for group 1 with a weather dataset your notebooks should be named as - Group1_Heartattack.ipynb.
2. Inside each jupyter notebook, you are required to mention your name, Group details and the Assignment dataset you will be working on.
3. Organize your code in separate sections for each task. Add comments to make the code readable.
4. Deep Learning Models are strictly not allowed. You are encouraged to learn classical Machine learning techniques and experience their behavior. For comparison of output with classical model you can use, if needed.
5. Notebooks without output shall not be considered for evaluation.
6. Delete unnecessary error messages and long outputs.
7. Display the analysis of attributes in one frame rather than one after one. However, special treatment to attributes can be displayed separately.
8. Prepare a jupyter notebook (recommended - Google Colab) to build, train and evaluate a Machine Learning model on the given dataset. Please read the instructions carefully.
9. Each group consists of up to 3 members. All members of the group will work on the same problem statement.
10. Only two files should be uploaded in canvas **without zipping** them. One is **ipynb file** and other one **html output of the ipynb file**. No other files should be uploaded.
11. Each group should upload in CANVAS in respective locations under ASSIGNMENT Tab. Assignment submitted via means other than through CANVAS will not be graded.

Problem Statement

Dataset: Heart attack possibility

Import Libraries/Dataset

1. Download the dataset
2. Import the required libraries

Data Visualization and Exploration [1 M]

1. Print 2 rows for sanity check to identify all the features present in the dataset and if the target matches with them.
2. Comment on class imbalance with appropriate visualization method.
3. Provide appropriate visualizations to get an insight about the dataset.
4. Do the correlational analysis on the dataset. Provide a visualization for the same. Will this correlational analysis affect the feature selection that you will perform in the next step? Justify your answer. **Answers must have justification to be awarded marks.**
5. Any other visualisation specific to the problem statement.

2. Data Pre-processing and cleaning [2M]

1. Do the appropriate pre-processing of the data like identifying NULL or Missing Values if any, handling of outliers if present in the dataset, skewed data etc. Mention the pre-processing steps performed in the markdown cell. Explore a few latest data balancing tasks and its effect on model evaluation parameters.
2. Apply appropriate feature engineering techniques for them. Apply the feature transformation techniques like Standardization, Normalization, etc. You are free to apply the appropriate transformations depending upon your dataset's structure and complexity. Provide proper justification. **Techniques used without justification will not be awarded marks.** Explore a few techniques for identifying feature importance for your feature engineering task.

3. Model Building [5M]

1. Split the dataset into training and test sets. **Answers without justification will not be awarded marks. [1 M]**

Case 1 : Train = 80 % Test = 20% [x_train1,y_train1] = 80% ;
 [x_test1,y_test1] = 20% ;

Case 2 : Train = 10 % Test = 90% [x_train2,y_train2] = 10% ;
 [x_test2,y_test2] = 90%

2. Explore k-fold cross-validation. **[1M]**

3. Build Model/s using 1) Logistic Regression [3M]

1. Do the prediction for the test data and display the results for the inference. Calculate all the evaluation metrics and choose the best for your model. Justify your answer. **Answers must have justification to be awarded marks. [1M]**

2. List out the performance measures in a tabular format. **(Accuracy, F1-score, Efficiency, sensitivity or specificity) [1M]**