## **SVM Implementation**

- 1. Import libraries
- 2. Import dataset
- 3. Explore the data to gain insights about the data

(View dimension, preview the dataset,

view the column names of the data frame,

remove leading spaces from column names,

rename column names,

check the distribution of the target\_class column,

View the percentage distribution of the target\_class column,

View the summary of the dataset,

Explore missing values in variables,

view summary statistics in numerical variables,

draw boxplots to visualize outliers,

Check the distribution of variables)

- 5. Split data into separate training and test set
- 6. Feature Scaling
- 7. Run SVM with default hyperparameters

Run SVM with rbf kernel and C=1.0 and gamma=auto

Run SVM with rbf kernel and C=100.0

Run SVM with rbf kernel and C=1000.0

8. Run SVM with linear kernel

Run SVM with linear kernel and C=1.0

Run SVM with linear kernel and C=100.0

Run SVM with linear kernel and C=1000.0

9. Run SVM with polynomial kernel

Run SVM with polynomial kernel and C=1.0

Run SVM with polynomial kernel and C=100.0

Run SVM with polynomial kernel and C=1000.0

10. Run SVM with sigmoid kernel

Run SVM with sigmoid kernel and C=1.0

Run SVM with sigmoid kernel and C=100.0

Run SVM with sigmoid kernel and C=1000.0

- 11. Compare the train-set and test-set accuracy
- 12. Check for overfitting and underfitting
- 13. visualize confusion matrix with seaborn heatmap
- 14. Generate classification report
- 15. Evaluate classification accuracy
- 16. Evaluate Classification error
- 17. Plot ROC Curve
- 18. Compute ROC AUC
- 19. Stratified k-fold cross-validation with shuffle split with linear kernel
- 21. Stratified k-fold cross-validation with shuffle split with rbf kernel
- 22. Stratified k-fold cross-validation with shuffle split with polynomial kernel
- 23. Stratified k-fold cross-validation with shuffle split with sigmoid kernel

## **Naive Bayes implementation**

- 1. Import libraries
- 2. Import dataset(adult.csv)
- 3. Exploratory data analysis

View dimensions of the dataset

Rename column names

View summary of the dataset

Explore categorical variables and view them

check missing values in categorical variables
view frequency counts of values in categorical variables
view frequency distribution of categorical variables
Explore workclass variable -

check labels in the workclass variable

check the frequency distribution of values in the workclass variable replace '?' values in the workclass variable with `NaN`

Explore occupation variable

check the frequency distribution of values in the occupation variable replace '?' values in occupation variable with `NaN`

Explore native\_country variable

check labels in the native\_country variable
check the frequency distribution of values in the native\_country variable
replace '?' values in the native\_country variable with `NaN`

Check missing values in categorical variables again

Explore Numerical Variables

view the numerical variables
check missing values in numerical variables

- 4. Declare feature vector and target variable
- 5. Split data into separate training and test set
- 6. print the percentage of missing values in the categorical variables in the training set
- 7. print categorical variables with missing data
- 8. impute missing categorical variables with the most frequent value
- 9. Encode categorical variables
- 10. Feature Scaling
- 11. Model training
- 12. Predict the results
- 13. Check the accuracy score
- 14. Compare the train-set and test-set accuracy
- 15. Visualize confusion matrix with seaborn heatmap
- 16. Generate Classification Report
- 17. Evaluate Classification accuracy
- 18. Evaluate Classification error
- 19. Calculate class probabilities and store the probabilities in data frame with appropriate heading

- 20. Plot histogram of predicted probabilities
- 21. Plot ROC Curve
- 22. Compute ROC AUC
- 23. Calculate cross-validated ROC AUC
- 24. Applying 10-Fold Cross Validation
- 25. Compute the Average cross-validation score