ITD105 Case Study #1

Comparing Machine Learning Algorithms

| | IFICATION | | |
|---------|---------------------------------------------|----------------------------------------------------------------------------------------------|---|
| Evaluat | | nachine learning algorithms designed for cl different resampling techniques and utilizing | |
| Classif | ication Dataset | | |
| Datase | t Name : | | |
| Feature | 98: | | |
| | | | |
| Set | a A sampling Technique : | | |
| | ssification Metric : Confusion Matrix and C | | |
| | | | |
| | ML Algorithm (Classification) | Confusion Matrix | |
| | | (Provide the matrix and classification report of each algorithm) | |
| C | ART (Classification and Regression Trees) | | |
| G | aussian Naive Bayes/Naive Bayes | | |
| G | radient Boosting Machines (AdaBoost) | | |
| K- | Nearest Neighbors (K-NN) | | |
| Lo | ogistic Regression | | |
| М | ulti-Layer Perceptron (MLP) | | |
| Pe | erceptron | | |
| Ra | andom Forest | | |
| ٥. | upport Vector Machines (SVM) | | |
| 50 | | | - |

| ML Algorithm (Classification) | |
|--------------------------------------------|--|
| CART (Classification and Regression Trees) | |
| Gaussian Naive Bayes/Naive Bayes | |
| Gradient Boosting Machines (AdaBoost) | |
| K-Nearest Neighbors (K-NN) | |
| Logistic Regression | |
| Multi-Layer Perceptron (MLP) | |
| Perceptron | |
| Random Forest | |
| Support Vector Machines (SVM) | |

| Resampling 1 | Technique: | |
|----------------|-------------------------------|---|
| Classification | n Metric: | |
| | | |
| | | ı |
| | ML Algorithm (Classification) | |

Set C (should use different resampling technique and classification metric)

| ML Algorithm (Classification) | |
|--------------------------------------------|--|
| CART (Classification and Regression Trees) | |
| Gaussian Naive Bayes/Naive Bayes | |
| Gradient Boosting Machines (AdaBoost) | |
| K-Nearest Neighbors (K-NN) | |
| Logistic Regression | |
| Multi-Layer Perceptron (MLP) | |
| Perceptron | |
| Random Forest | |
| Support Vector Machines (SVM) | |

Results interpretation (Set A , Set B and Set C):

Based on the results, perform algorithm/hyperparameter tuning (at least 3) of the chosen ML algorithm.

EXAMPLE:

ML Algorithm: Support Vector Machines (SVM)
Sampling Technique - Train/Test Split (80:20)

Classification Metrics - Accuracy

| | | SVM Hyperparameters | | | |
|-----------|--------------|---------------------|-----|----------|--|
| | random_state | Kernel | С | Accuracy | |
| Model I | seed | linear | 1.0 | 77.922 | |
| Model II | 5 | poly | 1.5 | 79.870 | |
| Model III | 10 | poly | 2 | 79.221 | |

Results interpretation:

II REGRESSION

Regression Dataset

Train the **regression dataset** using various machine learning algorithms designed for regression. Evaluate and compare these models by applying different resampling techniques and utilizing appropriate performance metrics.

| Dataset Nam | e: | |
|-------------|--------------------------------------------------------------------------------|-----------------|
| Features: | | |
| | ng Technique : | |
| Regressio | on Metric : | |
| | ML Algorithm (Regression) | |
| | CART (Classification and Regression Trees) | |
| | Elastic Net | |
| | Gradient Boosting Machines (AdaBoost) | |
| | K-Nearest Neighbors (K-NN) | |
| | Lasso Regression | |
| | Ridge Regression | |
| | Linear Regression | |
| | Multi-Layer Perceptron (MLP) | |
| | Random Forest | |
| Resampli | ould use different resampling technique and reg ng Technique: on Metric: | ression metric) |
| | ML Algorithm (Regression) | |
| | CART (Classification and Regression Trees) | |
| | Elastic Net | |
| | Gradient Boosting Machines (AdaBoost) | |
| | K-Nearest Neighbors (K-NN) | |
| | Lasso Regression | |
| | Ridge Regression | |

Results interpretation (Set A and Set B):

Random Forest

Linear Regression

Multi-Layer Perceptron (MLP)

Based on the results, perform at algorithm tuning (at least 3) of the chosen ML algorithm.

EXAMPLE:

ML Algorithm: Support Vector Machines (SVM) Sampling Technique - Train/Test Split (80:20)

Regression Metrics - MAE

| | | SVM Hyperparameters | | | |
|-----------|---------|---------------------|------|-------|--|
| | epsilon | Kernel | С | MAE | |
| Model I | 0.1 | rbf | 1.0 | 5.754 | |
| Model II | 0.2 | linear | 1.5 | 3.754 | |
| Model III | 0.15 | poly | 1.25 | 5.761 | |

Results interpretation:

Submit the following:

- a. Pdf copy of the results.
- b. Video link demonstrating the case study.