OSS Project #2

Automated Binary Classification

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Automated Machine Learning (AutoML)

- AutoML is the process of automating the time-consuming, iterative tasks of machine learning model development
- In this project, students will implement simple script about AutoML

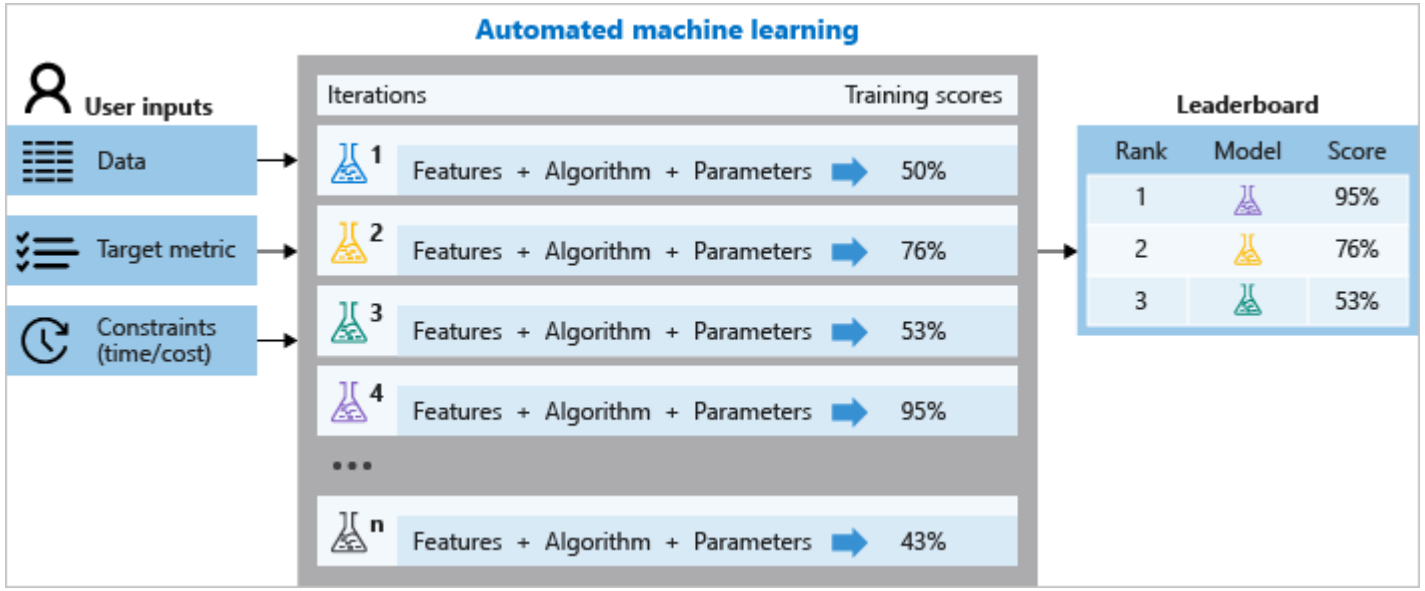


Image from: https://learn.microsoft.com/en-us/azure/machine-learning/concept-automated-ml

Project Goals

- The implemented script should automatically load data, split train dataset and test dataset, train various models and report performances
- Usage : **python template.py DATA_CSV_PATH TEST_SIZE**
  - TEST_SIZE in the form of ratio
- In this project, datasets must satisfy following requirements
  ① Binary classification
  ② All values are numeric
  ③ No missing values
  ④ The name of the label column is “target”

Example Commands & Results

```
python solution.py default_credit_card.csv 0.1
```

```
Number of features:  23
Number of data for class 0:  23364
Number of data for class 1:  6636

Splitting the dataset with the test size of  0.1

Decision Tree Performances
Accuracy:  0.7256666666666667
Precision:  0.3912448700410397
Recall:  0.43072289156626506

Random Forest Performances
Accuracy:  0.819
Precision:  0.6501240694789082
Recall:  0.39457831325301207

SVM Performances
Accuracy:  0.8156666666666667
Precision:  0.6627565982404692
Recall:  0.34036144578313254
```

```
python solution.py ../heart.csv 0.4
```

```
Number of features:  13
Number of data for class 0:  499
Number of data for class 1:  526

Splitting the dataset with the test size of  0.4

Decision Tree Performances
Accuracy:  0.9682926829268292
Precision:  0.9861751152073732
Recall:  0.9553571428571429

Random Forest Performances
Accuracy:  0.9780487804878049
Precision:  0.9864253393665159
Recall:  0.9732142857142857

SVM Performances
Accuracy:  0.9195121951219513
Precision:  0.9569377990430622
Recall:  0.8928571428571429
```

Project Requirements

- All requirements and corresponding scores are represented in the below
  - Upload the completed code on the GitHub (10 points)
  - Complete the function **load_dataset** (10 points)
  - Complete the function **dataset_stat** (15 points)
  - Complete the function **split_dataset** (20 points)
  - Complete the function **decision_tree_train_test** (15 points)
  - Complete the function **random_forest_train_test** (15 points)
  - Complete the function **svm_train_test** (15 points)

  - A total of 100 points

Project template

- The template for the code is provided on the I-Class and you must implement functions in the template and submit the completed code

- Please upload the completed code on your personal GitHub, and write the URL as the comment in the second line of the script!
  - If you missed writing URL or the script is not on the written URL, you cannot acquire **10 points about GitHub uploading**

```
#PLEASE WRITE THE GITHUB URL BELOW!
#

import sys

def load_dataset(dataset_path):
    #To-Do: Implement this function

def dataset_stat(dataset_df):
    #To-Do: Implement this function

def split_dataset(dataset_df, testset_size):
    #To-Do: Implement this function

def decision_tree_train_test(x_train, x_test, y_train, y_test):
    #To-Do: Implement this function

def random_forest_train_test(x_train, x_test, y_train, y_test):
    #To-Do: Implement this function

def svm_train_test(x_train, x_test, y_train, y_test):
    #To-Do: Implement this function

def print_performances(acc, prec, recall):
    #Do not modify this function!
    print ("Accuracy: ", acc)
    print ("Precision: ", prec)
    print ("Recall: ", recall)

```

Project template

- You can import additional modules that you need to implement each function
- Do not modify the function header (function name and parameter names)
- Please implement within the functions with “To-Do” comments. Do not edit other function and scripts

```
#PLEASE WRITE THE GITHUB URL BELOW!
#

import sys

def load_dataset(dataset_path):
    #To-Do: Implement this function

def dataset_stat(dataset_df):
    #To-Do: Implement this function

def split_dataset(dataset_df, testset_size):
    #To-Do: Implement this function

def decision_tree_train_test(x_train, x_test, y_train, y_test):
    #To-Do: Implement this function

def random_forest_train_test(x_train, x_test, y_train, y_test):
    #To-Do: Implement this function

def svm_train_test(x_train, x_test, y_train, y_test):
    #To-Do: Implement this function

def print_performances(acc, prec, recall):
    #Do not modify this function!
    print ("Accuracy: ", acc)
    print ("Precision: ", prec)
    print ("Recall: ", recall)

```

Project template

```python
if __name__ == '__main__':
    #Do not modify the main script!
    data_path = sys.argv[1]
    data_df = load_dataset(data_path)

    n_feats, n_class0, n_class1 = dataset_stat(data_df)
    print ("Number of features: ", n_feats)
    print ("Number of class 0 data entries: ", n_class0)
    print ("Number of class 1 data entries: ", n_class1)

    print ("\nSplitting the dataset with the test size of ", float(sys.argv[2]))
    x_train, x_test, y_train, y_test = split_dataset(data_df, float(sys.argv[2]))

    acc, prec, recall = decision_tree_train_test(x_train, x_test, y_train, y_test)
    print ("\nDecision Tree Performances")
    print_performances(acc, prec, recall)

    acc, prec, recall = random_forest_train_test(x_train, x_test, y_train, y_test)
    print ("\nRandom Forest Performances")
    print_performances(acc, prec, recall)

    acc, prec, recall = svm_train_test(x_train, x_test, y_train, y_test)
    print ("\nSVM Performances")
    print_performances(acc, prec, recall)
```

Function Descriptions

- load_dataset
  - Load the csv file at the given path into the pandas DataFrame and return the DataFrame

- dataset_stat
  - For the given DataFrame, return the following statistical analysis results in order
    - Number of features
    - Number of data for class 0
    - Number of data for class 1

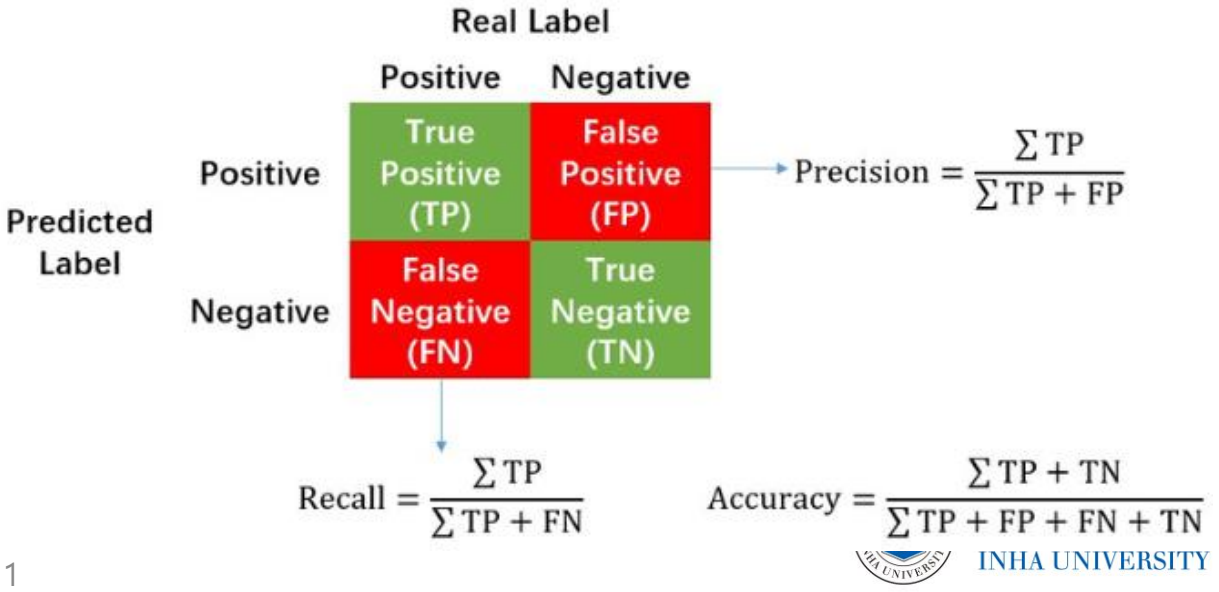
Function Descriptions

- split_dataset
  - Splitting the given DataFrame and return **train data, test data, train label, and test label** in order
  - You must split the data using the given test size

- decision_tree_train_test
  - Using the given train dataset, train the decision tree model
    - You can implement with default arguments
  - After the training, evaluate the performances of the model using the given test dataset
  - Return three performance metrics (accuracy, precision, recall) in order

Function Descriptions

- As described in the below figure, the **precision** is defined as **tp / (tp + fp)** in the confusion matrix. The precision is intuitively the ability of the classifier not to label as positive a sample that is negative
- The **recall** is defined as **tp / (tp + fn)**. The recall is intuitively the ability of the classifier to find all the positive samples
- Accuracy is not a good metric for the imbalanced dataset, therefore we analyze the model performance with additional metrics



Function Descriptions

- Please refer below sklearn functions

sklearn.metrics.**precision_score**(*y_true, y_pred, *, labels=None, pos_label=1, average='binary', sample_weight=None, zero_division='warn'*) [source]

sklearn.metrics.**recall_score**(*y_true, y_pred, *, labels=None, pos_label=1, average='binary', sample_weight=None, zero_division='warn'*) [source]

Function Descriptions

- random_forest_train_test
  - Using the given train dataset, train the random forest model
    - You can implement with default arguments
  - After the training, evaluate the performances of the model using the given test dataset
  - Return three performance metrics (accuracy, precision, recall) in order
- svm_train_test
  - Using the given train dataset, train **the pipeline consists of a standard scaler and SVM**
    - You can implement with default arguments
  - After the training, evaluate the performances of the model using the given test dataset
  - Return three performance metrics (accuracy, precision, recall) in order

Submission

- Submission requires only a completed **template.py** file
  - Don’t forget to write GitHub URL and upload the template.py file on the GitHub
- Please submit your file on the I-Class
- Due date is 11/27(Sun) 23:59
- TA will verify your submissions using the another auto-script and copy checking tools

- If you have any questions, don’t hesitate to send an e-mail (dgkim@inha.ac.kr) or an I-Class message