# **COMP5318 Assignment 1: Classification**

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```
In [73]: import data as data
         # Import all libraries
         from sklearn.ensemble import BaggingClassifier, AdaBoostClassifier, GradientBoos
         from sklearn.model_selection import StratifiedKFold
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.model_selection import cross_val score
         from sklearn.model_selection import train_test_split
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.model selection import GridSearchCV
         from sklearn.naive_bayes import GaussianNB
         from sklearn.linear model import LogisticRegression
         from sklearn.impute import SimpleImputer
         from sklearn.metrics import f1 score
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.preprocessing import MinMaxScaler
         from sklearn.preprocessing import LabelEncoder
         from sklearn.datasets import data
         from pandas.core import frame
         from sklearn.svm import SVC
         from sklearn import datasets
         from sklearn import preprocessing
         from sklearn.datasets import load iris
         import numpy as np
         import pandas as pd
In [74]: # Load dataset
         df = pd.read csv('breast-cancer-wisconsin.csv')
In [75]: # Pre-process dataset
         df = df.replace('?', np.nan).replace('class1',0).replace('class2',1)
         feature = df.iloc[:, 0:-1]
         imputer = SimpleImputer(missing_values=np.nan, strategy='mean')
         features = imputer.fit_transform(feature)
         scaler = MinMaxScaler()
         X = scaler.fit_transform(features)
         classes = df.iloc[:, -1].tolist()
         labels = np.unique(classes)
         lEnc = LabelEncoder()
         lEnc.fit(labels)
         label_encoder = lEnc.transform(classes)
         numClass = len(labels)
         y = label_encoder.astype(np.float64)
In [76]: # Print first ten rows of pre-processed dataset to 4 decimal places as per assig
         # A function is provided to assist
```

```
dataframe = df.replace('?',np.nan).replace('class1',0).replace('class2',1)
simputer = SimpleImputer(missing_values=np.nan,strategy='mean')
minmaxscaler = preprocessing.MinMaxScaler()
dataset = imputer.fit_transform(dataframe)
dataset = minmaxscaler.fit transform(dataset)
def print data(X, n rows=10):
    """Takes a numpy data array and target and prints the first ten rows.
    Arguments:
        X: numpy array of shape (n_examples, n_features)
        y: numpy array of shape (n_examples)
        n_rows: numpy of rows to print
   for example_num in range(n_rows):
        for feature in X[example num][0:-1:1]:
            print("{:.4f}".format(feature), end=",")
        print(int(X[example num][-1]))
print_data(dataset)
```

```
0.4444,0.0000,0.0000,0.0000,0.1111,0.0000,0.2222,0.0000,0.0000,0\\0.4444,0.3333,0.3333,0.4444,0.6667,1.0000,0.2222,0.1111,0.0000,0\\0.2222,0.0000,0.0000,0.0000,0.1111,0.1111,0.2222,0.0000,0.0000,0\\0.5556,0.7778,0.7778,0.0000,0.2222,0.3333,0.2222,0.6667,0.0000,0\\0.3333,0.0000,0.0000,0.2222,0.1111,0.0000,0.2222,0.0000,0.0000,0\\0.7778,1.0000,1.0000,0.7778,0.6667,1.0000,0.8889,0.6667,0.0000,1\\0.0000,0.0000,0.0000,0.0000,0.1111,1.0000,0.2222,0.0000,0.0000,0\\0.1111,0.0000,0.1111,0.0000,0.1111,0.0000,0.2222,0.0000,0.0000,0\\0.3333,0.1111,0.0000,0.0000,0.01111,0.0000,0.1111,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.0000,0.00
```

## Part 1: Cross-validation without parameter tuning

```
In [77]: ## Setting the 10 fold stratified cross-validation
         cvKFold=StratifiedKFold(n_splits=10, shuffle=True, random_state=0)
         # The stratified folds from cvKFold should be provided to the classifiers
In [78]: # Logistic Regression
         def logregClassifier(X, y):
             logreg = LogisticRegression(random state=0)
             scores = cross_val_score(logreg, np.asarray(X, dtype='float64'), y, cv=cvKFd
             return scores.mean()
In [79]: #Naïve Bayes
         def nbClassifier(X, y):
             nb = GaussianNB()
             scores = cross_val_score(nb, np.asarray(X, dtype='float64'), y, cv=cvKFold)
             return scores.mean()
In [80]: # Decision Tree
         def dtClassifier(X, y):
             tree = DecisionTreeClassifier(criterion='entropy', random_state=0)
             scores = cross_val_score(tree, np.asarray(X, dtype='float64'), y, cv=cvKFold
             return scores.mean()
```

#### Part 1 Results

```
In [82]: from sklearn.preprocessing import LabelEncoder
         # Parameters for Part 1:
         X_train, X_test, y_train, y_test = train_test_split(X, y, stratify=y, random_sta
         #Bagging
         bag n estimators = 60
         bag max samples = 100
         bag_max_depth = 6
         #AdaBoost
         ada n estimators = 60
         ada_learning_rate = 0.5
         ada bag max depth = 6
         #GR
         gb_n_estimators = 60
         gb_learning_rate = 0.5
         # Print results for each classifier in part 1 to 4 decimal places here:
         print("LogR average cross-validation accuracy: ",logregClassifier(X, y))
         print("NB average cross-validation accuracy: ",nbClassifier(X, y))
         print("DT average cross-validation accuracy: ",dtClassifier(X, y))
         print("Bagging average cross-validation accuracy: ",bagDTClassifier(X, y, bag_n_
         print("AdaBoost average cross-validation accuracy: ",adaDTClassifier(X, y, ada_r
         print("GB average cross-validation accuracy: ",gbClassifier(X, y, gb_n_estimator
         LogR average cross-validation accuracy: 0.9642443064182193
         NB average cross-validation accuracy: 0.9585093167701864
         DT average cross-validation accuracy: 0.9384679089026916
         Bagging average cross-validation accuracy: 0.9571014492753622
         AdaBoost average cross-validation accuracy: 0.9585093167701864
         GB average cross-validation accuracy: 0.9613457556935817
```

## Part 2: Cross-validation with parameter tuning

```
In [83]: # KNN
         k = [1, 3, 5, 7, 9]
         p = [1, 2]
         def bestKNNClassifier(X, y):
             param_grid = {'n_neighbors':k, 'p':p}
             knn = KNeighborsClassifier()
             scores = GridSearchCV(knn, param_grid, cv=cvKFold, return_train_score=True)
             scores.fit(X, y)
             return scores
In [84]: # SVM
         # You should use SVC from sklearn.svm with kernel set to 'rbf'
         C = [0.01, 0.1, 1, 5, 15]
         gamma = [0.01, 0.1, 1, 10, 50]
         def bestSVMClassifier(X, y):
             param grid = {'C':C, 'gamma':gamma}
             grid_search = GridSearchCV(SVC(kernel='rbf', random_state=0), param_grid, cv
             grid_search.fit(X, y)
             return grid_search
In [85]: # Random Forest
         # You should use RandomForestClassifier from sklearn.ensemble with information g
         n_{estimators} = [10, 30, 60, 100, 150]
         max_leaf_nodes = [6, 12, 18]
         def bestRFClassifier(X, y):
             param_grid = {'n_estimators':n_estimators, 'max_leaf_nodes':max_leaf_nodes}
             X_train, X_test, y_train, y_test = train_test_split(X, y, stratify=y, random
             grid_search = GridSearchCV(RandomForestClassifier(random_state=0, criterion=
             grid_search.fit(X_train, y_train)
             return grid search
```

### Part 2: Results

```
In [86]: # Perform Grid Search with 10-fold stratified cross-validation (GridSearchCV in
         # The stratified folds from cvKFold should be provided to GridSearchV
         # This should include using train_test_split from sklearn.model_selection with s
         # Print results for each classifier here. All results should be printed to 4 dec
         # "k", "p", n estimators" and "max leaf nodes" which should be printed as intege
         X_train, X_test, y_train, y_test = train_test_split(X, y, stratify=y, random_sta
         bestKNN = bestKNNClassifier(X_train,y_train)
         print("KNN best k: ", bestKNN.best_params_['n_neighbors'])
         print("KNN best p: ", bestKNN.best_params_['p'])
         print("KNN cross-validation accuracy: ", bestKNN.best_score_)
         print("KNN test set accuracy: ", bestKNN.score(X_test, y_test))
         print()
         bestSVM = bestSVMClassifier(X_train, y_train)
         print("SVM best C: ", bestSVM.best_params_['C'])
         print("SVM best gamma: ", bestSVM.best_params_['gamma'])
         print("SVM cross-validation accuracy: ", bestSVM.best_score_)
```

```
print("SVM test set accuracy: ", bestSVM.score(X_test, y_test))
        print()
        bestRFC = bestRFClassifier(X, y)
        y_predict = bestRFC.predict(X_test)
        print("RF best n_estimators: ", bestRFC.best_params_['n_estimators'])
        print("RF best max_leaf_nodes: ", bestRFC.best_params_['max_leaf_nodes'])
        print("RF cross-validation accuracy: ", bestRFC.best_score_)
        print("RF test set accuracy: ", bestRFC.score(X_test, y_test))
        print("RF test set macro average F1: ", f1_score(y_test, y_predict, average='mac
        print("RF test set weighted average F1: ", f1_score(y_test, y_predict, average="
        KNN best k: 3
        KNN best p: 1
        KNN cross-validation accuracy: 0.9694847605224963
        KNN test set accuracy: 0.9542857142857143
        SVM best C: 5
        SVM best gamma: 0.1
        SVM cross-validation accuracy: 0.9676342525399129
        SVM test set accuracy: 0.9714285714285714
        RF best n_estimators: 150
        RF best max_leaf_nodes: 6
        RF cross-validation accuracy: 0.9675253991291728
        RF test set accuracy: 0.9657142857142857
        RF test set macro average F1: 0.9627976190476191
        RF test set weighted average F1: 0.9660714285714286
In [ ]:
```