1/24/23, 5:38 PM HW2

## **Performance Baselines**

## **Dataset**

Source: https://archive.ics.uci.edu/ml/datasets/Dry+Bean+Dataset (https://archive.ics.uci.edu/ml/datasets/Dry+Bean+Dataset)

```
In [3]: import pandas as pd
        X train = pd.read csv('https://raw.githubusercontent.com/rasbt/stat451-machine-learning-fs21/main/hw0
        2-starter/dataset/X train.csv', header=None).values
        y train = pd.read csv('https://raw.githubusercontent.com/rasbt/stat451-machine-learning-fs21/main/hw0
        2-starter/dataset/y train.csv', header=None).values.ravel().astype(int)
        X test = pd.read csv('https://raw.githubusercontent.com/rasbt/stat451-machine-learning-fs21/main/hw02
        -starter/dataset/X test.csv', header=None).values
        y test = pd.read csv('https://raw.githubusercontent.com/rasbt/stat451-machine-learning-fs21/main/hw02
        -starter/dataset/y test.csv', header=None).values.ravel().astype(int)
        print('X train.shape:', X train.shape)
        print('y train.shape:', y train.shape)
        print('X test.shape:', X test.shape)
        print('y_test.shape:', y test.shape)
        X train.shape: (9119, 16)
        y train.shape: (9119,)
        X test.shape: (4492, 16)
        y test.shape: (4492,)
In [4]: from sklearn import model selection
        from sklearn.model selection import train test split
        X train sub, X valid, y train sub, y valid = \
            train test split(X train, y train, test size=0.2, random state=1, stratify=y train)
        print('Train/Valid/Test sizes:', y train.shape[0], y valid.shape[0], y test.shape[0])
```

HW2

Train/Valid/Test sizes: 9119 1824 4492

## **Baselines**

Compare hyperparameter settings on validation set:

1/24/23, 5:38 PM HW2

```
In [5]: from sklearn.neighbors import KNeighborsClassifier
        knn = KNeighborsClassifier(n neighbors=5)
        knn.fit(X train sub, y train sub)
        print(f"Train Accuracy: {knn.score(X train sub, y train sub)*100:0.3f}%")
        print(f"Valid Accuracy: {knn.score(X valid, y valid)*100:0.3f}%")
        Train Accuracy: 79.657%
        Valid Accuracy: 71.162%
In [6]: knn = KNeighborsClassifier(n neighbors=3)
        knn.fit(X train sub, y train sub)
        print(f"Train Accuracy: {knn.score(X train sub, y train sub)*100:0.3f}%")
        print(f"Valid Accuracy: {knn.score(X valid, y valid)*100:0.3f}%")
        Train Accuracy: 84.003%
        Valid Accuracy: 71.930%
In [7]: knn = KNeighborsClassifier(n neighbors=7)
        knn.fit(X train sub, y train sub)
        print(f"Train Accuracy: {knn.score(X_train_sub, y_train_sub)*100:0.3f}%")
        print(f"Valid Accuracy: {knn.score(X valid, y valid)*100:0.3f}%")
        Train Accuracy: 77.478%
        Valid Accuracy: 69.518%
```

Choose best model and train on whole training set:

Test Accuracy: 71.305%

```
In [8]: model = KNeighborsClassifier(n_neighbors=3)
    model.fit(X_train, y_train)
    print(f"Train Accuracy: {model.score(X_train, y_train)*100:0.3f}%")
    print(f"Test Accuracy: {model.score(X_test, y_test)*100:0.3f}%")
Train Accuracy: 84.965%
```

```
In [11]: # 4.3
         from optuna.integration import LightGBMPruningCallback
         import numpy as np
         import lightgbm
         import optuna
         from sklearn.metrics import log loss
         from sklearn.model selection import StratifiedKFold
         import warnings
         warnings.filterwarnings("ignore", category=UserWarning)
         #optuna.logging.set verbosity(optuna.logging.WARNING)
         def objective(trial, X train, y train, cv=5):
             param grid = {
                 "n estimators": trial.suggest categorical("n estimators", [10, 100]),
                 "learning rate": trial.suggest categorical("learning rate", [0.01]),
             }
             cv iterator = StratifiedKFold(n splits=cv, shuffle=True, random state=123)
             cv scores = np.zeros(cv)
             for idx, (train sub idx, valid idx) in enumerate(cv iterator.split(X train, y train)):
                 X train sub, X valid = X train[train sub idx], X train[valid idx]
                 y train sub, y valid = y train[train sub idx], y train[valid idx]
                 model = lightgbm.LGBMClassifier(objective="multi logloss", **param grid)
                 model.fit(
                     X train sub,
                     y train sub,
                     eval set=[(X valid, y valid)],
                     eval metric="multi logloss",
                     verbose=-1,
                     early stopping rounds=50,
                     callbacks=[
                         LightGBMPruningCallback(trial=trial, metric="multi logloss")
                     ], # Add a pruning callback to eliminate unpromising candidates
```

HW2

```
preds = model.score(X_valid, y_valid)

cv_scores[idx] = preds

return 1-np.mean(cv_scores)
```

1/24/23, 5:38 PM

```
In [12]: study = optuna.create_study(direction="minimize", study_name="LGBM Classifier")

def func(trial):
    return objective(trial, X_train, y_train)

study.optimize(func, n_trials=50);
```

[I 2021-11-10 22:17:58,256] A new study created in memory with name: LGBM Classifier [I 2021-11-10 22:18:00,553] Trial 0 finished with value: 0.07511800964286763 and parameters: {'n est imators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:00,832] Trial 1 finished with value: 0.48239960158212314 and parameters: {'n est imators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:01,109] Trial 2 finished with value: 0.48239960158212314 and parameters: {'n est imators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:01,384] Trial 3 finished with value: 0.48239960158212314 and parameters: {'n est imators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:01,682] Trial 4 finished with value: 0.48239960158212314 and parameters: {'n est imators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:01,970] Trial 5 finished with value: 0.48239960158212314 and parameters: {'n est imators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:02,257] Trial 6 finished with value: 0.48239960158212314 and parameters: {'n est imators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:02,557] Trial 7 finished with value: 0.48239960158212314 and parameters: {'n est imators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:05,341] Trial 8 finished with value: 0.07511800964286763 and parameters: {'n est imators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:05,640] Trial 9 finished with value: 0.48239960158212314 and parameters: {'n est imators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:08,097] Trial 10 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:10,777] Trial 11 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:13,354] Trial 12 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:15,883] Trial 13 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:18,448] Trial 14 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:21,054] Trial 15 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:23,817] Trial 16 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:26,408] Trial 17 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:28,988] Trial 18 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:31,599] Trial 19 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:34,782] Trial 20 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763.

HW2

[I 2021-11-10 22:18:37,797] Trial 21 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:40,443] Trial 22 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:43,170] Trial 23 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:45,837] Trial 24 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:48,489] Trial 25 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:51,131] Trial 26 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:53,963] Trial 27 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:56,682] Trial 28 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:18:59,412] Trial 29 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:02,214] Trial 30 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:04,989] Trial 31 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:07,699] Trial 32 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:10,396] Trial 33 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:13,137] Trial 34 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:13,479] Trial 35 finished with value: 0.48239960158212314 and parameters: {'n es timators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:16,186] Trial 36 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:16,521] Trial 37 finished with value: 0.48239960158212314 and parameters: {'n es timators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:19,292] Trial 38 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:19,635] Trial 39 finished with value: 0.48239960158212314 and parameters: {'n es timators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:22,396] Trial 40 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:25,221] Trial 41 finished with value: 0.07511800964286763 and parameters: {'n es timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763. [I 2021-11-10 22:19:27,976] Trial 42 finished with value: 0.07511800964286763 and parameters: {'n es

HW2

1/24/23, 5:38 PM HW2

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timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763.
         [I 2021-11-10 22:19:30,683] Trial 43 finished with value: 0.07511800964286763 and parameters: {'n es
         timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763.
         [I 2021-11-10 22:19:33,460] Trial 44 finished with value: 0.07511800964286763 and parameters: {'n es
         timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763.
         [I 2021-11-10 22:19:33,801] Trial 45 finished with value: 0.48239960158212314 and parameters: {'n es
         timators': 10, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763.
         [I 2021-11-10 22:19:36,459] Trial 46 finished with value: 0.07511800964286763 and parameters: {'n es
         timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763.
         [I 2021-11-10 22:19:39,221] Trial 47 finished with value: 0.07511800964286763 and parameters: {'n es
         timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763.
         [I 2021-11-10 22:19:41,913] Trial 48 finished with value: 0.07511800964286763 and parameters: {'n es
         timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763.
         [I 2021-11-10 22:19:44,652] Trial 49 finished with value: 0.07511800964286763 and parameters: {'n es
         timators': 100, 'learning rate': 0.01}. Best is trial 0 with value: 0.07511800964286763.
In [13]: print(f"\tBest value: {study.best value:.5f}")
         print(f"\tBest params:")
         for key, value in study.best params.items():
             print(f"\t\t{key}: {value}")
                 Best value: 0.07512
                 Best params:
                         n estimators: 100
                         learning rate: 0.01
In [14]: model = lightqbm.LGBMClassifier(objective="multi logloss", **study.best params)
         model.fit(X train, y train)
Out[14]: LGBMClassifier(learning rate=0.01, objective='multi logloss')
In [15]: print(f"Training Accuracy: {model.score(X train, y train):0.5f}")
         print(f"Test Accuracy: {model.score(X test, y test):0.5f}")
         Training Accuracy: 0.95646
         Test Accuracy: 0.92164
In [ ]:
```