## hw1 full code

October 5, 2020

## 1 Homework 1

This is the template for the first homework assignment. The only function that you are required to fill in and turn in to Gradescope is "compute\_features". Please do not edit definition of "compute\_features" so the Gradescope unit tests run successfully.

```
[1]: # Uncomment and run this code if you want to verify your `sklearn` installation.
# If this cell outputs 'array([1])', then it's installed correctly.

# from sklearn import tree
# X = [[0, 0], [1, 1]]
# y = [0, 1]
# clf = tree.DecisionTreeClassifier(criterion='entropy')
# clf = clf.fit(X, y)
# clf.predict([[2, 2]])
```

```
[2]: from sklearn.metrics import accuracy_score import numpy as np
```

```
[3]: # Loading in training data
    madelon_x_train = np.load('madelon/train/X_train.npy')
    madelon_y_train = np.load('madelon/train/y_train.npy')
    madelon_x_test = np.load('madelon/test/X_test.npy')
    madelon_y_test = np.load('madelon/test/y_test.npy')

# print(madelon_x_train)
# print(madelon_y_train)
# print(madelon_x_test)
# print(madelon_y_test)
```

```
[4]: # Loading in badges data

badges_x_train = np.loadtxt('badges/train/train.names.txt', dtype=str)
badges_y_train = np.load('badges/train/y_train_badges.npy')
badges_x_test = np.loadtxt('badges/test/test_names_badges.txt', dtype=str)
badges_y_test = np.load('badges/test/y_test_badges.npy')

# print(badges_x_train)
```

```
# print(badges_y_train)
# print(badges_x_test)
# print(badges_y_test)
badges_x_train_transformed = np.empty(len(badges_x_train), dtype=np.

→dtype('U100'))
badges_x_test_transformed= np.empty(len(badges_x_test), dtype=np.dtype('U100'))
for i in range(0, len(badges_x_train)):
    badges_x_train_transformed[i] = (badges_x_train[i][0] + " " +__
→badges_x_train[i][1])
for i in range(0, len(badges_x_test)):
   badges_x_test_transformed[i] = badges_x_test[i][0] + " " +__
→badges_x_test[i][1]
# print(badges_x_train_transformed)
# print(badges_x_test_transformed)
# ord(badges_x_train_transformed[0].lower().split()[0][0])
# ord('a')
# min(5,6,7)
# badges_x_train_transformed[0].lower().split()
```

```
return feature_array
```

```
[6]: compute_features(badges_x_train_transformed)[0]
```

You are not required to use the functions defined below, but they may be useful for you to think about how to structure your code.

```
[7]: # Function to implement SGDClassifier model

from sklearn.linear_model import SGDClassifier

def train_and_evaluate_sgd(X_train, y_train, X_test, y_test):
    model = SGDClassifier(loss='log', max_iter=10000)
    model = model.fit(X_train, y_train)
    pred = model.predict(X_test)

test_acc=accuracy_score(y_test, pred)

return model.score(X_train, y_train), test_acc, model
```

```
[8]: # Function to implement Decision Tree and Decision Stump models

from sklearn.tree import DecisionTreeClassifier

def train_and_evaluate_decision_tree(X_train, y_train, X_test, y_test):
    model = DecisionTreeClassifier(criterion='entropy')
    model = model.fit(X_train, y_train)
    pred = model.predict(X_test)

test_acc=accuracy_score(y_test, pred)
```

```
return model.score(X_train, y_train), test_acc, model

def train_and_evaluate_decision_stump(X_train, y_train, X_test, y_test):
    model = DecisionTreeClassifier(criterion='entropy', max_depth=4)
    model = model.fit(X_train, y_train)
    pred = model.predict(X_test)

    test_acc=accuracy_score(y_test, pred)
    return model.score(X_train, y_train), test_acc, model
```

```
[9]: # Old function
     # import random as rd
     # import numpy as np
     \# def train_and_evaluate_sqd_with_stumps(X_train, y_train, X_test, y_test):
           stump\_outputs = np.empty(shape=(len(X\_train), 100))
           test\_outputs = np.empty(shape=(len(X\_test), 100))
           for i in range (0,100):
               feature set=rd.sample(range(0,len(X train[0])), k=int(len(X train[0]))
      →2))
               x_train_split=X_train[:,feature_set]
               model=DecisionTreeClassifier(criterion='entropy', max_depth=4)
               model=model.fit(x_train_split, y_train)
               pred=model.predict(x_train_split)
               for j in range (0, len(pred)):
     #
                   stump_outputs[j,i]=pred[j]
     #
               x_test_split=X_test[:,feature_set]
               test_pred = model.predict(x_test_split)
               for k in range (0, len(test_pred)):
     #
                   test_outputs[k,i]=test_pred[k]
     #
           model_final = SGDClassifier(loss='log', max_iter=10000)
           model_final = model_final.fit(stump_outputs, y_train)
     #
           pred_final = model_final.predict(test_outputs)
           test_acc=accuracy_score(y_test, pred_final)
           return model_final.score(stump_outputs, y_train), test_acc, model_final
```

```
[10]: import random as rd
      import numpy as np
      def create_feature_set(X_train):
          feature_set = np.empty(shape=(100, int(len(X_train[0])/2)), dtype=int)
          for i in range(0,100):
              feature_set[i]=rd.sample(range(0,len(X_train[0])),__
       \rightarrowk=int(len(X_train[0])/2))
          return feature_set
      def create_stump_models(X_train, y_train, feature_set):
          stump_models = np.empty(shape=100, dtype=object)
          for i in range (0,100):
              x_train_split=X_train[:,feature_set[i]]
              model=DecisionTreeClassifier(criterion='entropy', max_depth=4)
              model=model.fit(x train split, y train)
              stump_models[i]=model
          return stump_models
      def create_stump_pred_features(X_train, feature_set, stump_models):
          stump_pred_features = np.empty(shape=(len(X_train),100))
          for i in range (0,100):
              x_train_split=X_train[:,feature_set[i]]
              pred=stump_models[i].predict(x_train_split)
              for j in range (0, len(pred)):
                  stump_pred_features[j,i]=pred[j]
          return stump_pred_features
      def create_model_final(stump_pred_features, y_train):
          model_final = SGDClassifier(loss='log', max_iter=10000)
          model_final = model_final.fit(stump_pred_features, y_train)
          return model_final
      def sgd_with_stumps_predictions(model_final, input_data):
          return model final.predict(input data)
      def train and evaluate sgd with stumps(X train, y train, X test, y test):
          feature_set = create_feature_set(X_train)
          stump_models = create_stump_models(X_train, y_train, feature_set)
          stump_predicted_features = create_stump_pred_features(X_train, feature_set,_
       →stump_models)
          final_model = create_model_final(stump_predicted_features, y_train)
```

```
predictions = sgd_with_stumps_predictions(final_model,__
       →stump_predicted_features)
          test stump predicted features = create stump pred features(X test,
       →feature_set, stump_models)
          test_predictions = sgd_with_stumps_predictions(final_model,__
       →test_stump_predicted_features)
          train_acc = final_model.score(stump_predicted_features, y_train)
          test acc = accuracy score(test predictions, y test)
          return train_acc, test_acc, final_model, feature_set, stump_models
[11]: # train and evaluate sqd with stumps(madelon x train, madelon y train,
       \rightarrow madelon x test, madelon y test)
[12]: def load_cv_split(fold):
          train_x= np.load('madelon/cross_validation/cv-train-X.' + str(fold) + '.
          train y= np.load('madelon/cross validation/cv-train-y.' + str(fold) + '.
       →npy')
          test_x= np.load('madelon/cross_validation/cv-heldout-X.' + str(fold) + '.
       test y= np.load('madelon/cross validation/cv-heldout-y.' + str(fold) + '.
       →npy')
          return train_x, train_y, test_x, test_y
[13]: sgd_train_acc_array=[]
      dt train acc array=[]
      ds_train_acc_array=[]
      cmbd_train_acc_array=[]
      sgd_heldout_acc_array=[]
      dt_heldout_acc_array=[]
      ds_heldout_acc_array=[]
      cmbd_heldout_acc_array=[]
      for i in range(0,5):
          dl = load_cv_split(i)
          sgd = train_and_evaluate_sgd(d1[0], d1[1], d1[2], d1[3])
          tree = train_and_evaluate_decision_tree(dl[0], dl[1], dl[2], dl[3])
          stump = train and_evaluate_decision_stump(d1[0], d1[1], d1[2], d1[3])
          combined = train_and_evaluate_sgd_with_stumps(dl[0], dl[1], dl[2], dl[3])
```

```
sgd_train_acc_array.append(sgd[0])
         dt_train_acc_array.append(tree[0])
         ds_train_acc_array.append(stump[0])
         cmbd_train_acc_array.append(combined[0])
         sgd heldout acc array.append(sgd[1])
         dt_heldout_acc_array.append(tree[1])
         ds heldout acc array.append(stump[1])
          cmbd_heldout_acc_array.append(combined[1])
     print("sgd_train_acc_array: " + str(sgd_train_acc_array))
     print("dt_train_acc_array: " + str(dt_train_acc_array))
     print("ds_train_acc_array: " + str(ds_train_acc_array))
     print("cmbd_train_acc_array: " + str(cmbd_train_acc_array))
     print()
     print("sgd_heldout_acc_array: " + str(sgd_heldout_acc_array))
     print("dt_heldout_acc_array: " + str(dt_heldout_acc_array))
     print("ds_heldout_acc_array: " + str(ds_heldout_acc_array))
     print("cmbd_heldout_acc_array: " + str(cmbd_heldout_acc_array))
     sgd_train_acc_array: [0.7791666666666667, 0.8071428571428572,
     0.7857142857142857, 0.7875, 0.7696428571428572]
     dt_train_acc_array: [1.0, 1.0, 1.0, 1.0, 1.0]
     ds_train_acc_array: [0.7636904761904761, 0.7488095238095238, 0.756547619047619,
     0.7690476190476191, 0.7690476190476191
     cmbd_train_acc_array: [0.8857142857142857, 0.8601190476190477,
     0.8845238095238095, 0.7589285714285714, 0.8809523809523809]
     sgd heldout acc array: [0.5880952380952381, 0.611904761904762,
     0.5642857142857143, 0.5119047619047619, 0.6357142857142857]
     dt heldout acc array: [0.6523809523809524, 0.7214285714285714,
     0.6833333333333333, 0.6880952380952381, 0.6333333333333333333
     0.7071428571428572, 0.6952380952380952, 0.716666666666667]
     cmbd_heldout_acc_array: [0.7904761904761904, 0.7642857142857142,
     0.7761904761904762, 0.6595238095238095, 0.8071428571428572
[14]: sgd_avg_train_acc = np.mean(sgd_train_acc_array)
     dt_avg_train_acc = np.mean(dt_train_acc_array)
     ds_avg_train_acc = np.mean(ds_train_acc_array)
     cmbd avg train acc = np.mean(cmbd train acc array)
     sgd avg heldout acc = np.mean(sgd heldout acc array)
     dt_avg_heldout_acc = np.mean(dt_heldout_acc_array)
     ds avg heldout acc = np.mean(ds heldout acc array)
     cmbd_avg_heldout_acc = np.mean(cmbd_heldout_acc_array)
```

```
sgd_std_train_acc = np.std(sgd_train_acc_array)
dt_std_train_acc = np.std(dt_train_acc_array)
ds_std_train_acc = np.std(ds_train_acc_array)
cmbd_std_train_acc = np.std(cmbd_train_acc_array)
sgd_std_heldout_acc = np.std(sgd_heldout_acc_array)
dt_std_heldout_acc = np.std(dt_heldout_acc_array)
ds std heldout acc = np.std(ds heldout acc array)
cmbd_std_heldout_acc = np.std(cmbd_heldout_acc_array)
sgd_train_acc_ci = [sgd_avg_train_acc - 1.96*sgd_std_train_acc/(np.
sgd_avg_train_acc + 1.96*sgd_std_train_acc/(np.
→sqrt(len(sgd_train_acc_array)))]
dt_train_acc_ci = [dt_avg_train_acc - 1.96*dt_std_train_acc/(np.
dt_avg_train_acc + 1.96*dt_std_train_acc/(np.
→sqrt(len(dt_train_acc_array)))]
ds_train_acc_ci = [ds_avg_train_acc - 1.96*ds_std_train_acc/(np.

→sqrt(len(ds_train_acc_array))),
                 ds_avg_train_acc + 1.96*ds_std_train_acc/(np.
→sqrt(len(ds_train_acc_array)))]
cmbd_train_acc_ci = [cmbd_avg_train_acc - 1.96*cmbd_std_train_acc/(np.

→sqrt(len(cmbd_train_acc_array))),
                 cmbd_avg_train_acc + 1.96*cmbd_std_train_acc/(np.
→sqrt(len(cmbd_train_acc_array)))]
sgd_heldout_acc_ci = [sgd_avg_heldout_acc - 1.96*sgd_std_heldout_acc/(np.
sgd_avg_heldout_acc + 1.96*sgd_std_heldout_acc/(np.
→sqrt(len(sgd_heldout_acc_array)))]
dt_heldout_acc_ci = [dt_avg_heldout_acc - 1.96*dt_std_heldout_acc/(np.
dt_avg_heldout_acc + 1.96*dt_std_heldout_acc/(np.

¬sqrt(len(dt_heldout_acc_array)))]
ds_heldout_acc_ci = [ds_avg_heldout_acc - 1.96*ds_std_heldout_acc/(np.
ds_avg_heldout_acc + 1.96*ds_std_heldout_acc/(np.
→sqrt(len(ds_heldout_acc_array)))]
cmbd_heldout_acc_ci = [cmbd_avg_heldout_acc - 1.96*cmbd_std_heldout_acc/(np.
→sqrt(len(cmbd_heldout_acc_array))),
                 cmbd_avg_heldout_acc + 1.96*cmbd_std_heldout_acc/(np.
→sqrt(len(cmbd_heldout_acc_array)))]
print("sgd_avg_train_acc: " + str(sgd_avg_train_acc))
```

```
print("dt_avg_train_acc: " + str(dt_avg_train_acc))
print("ds_avg_train_acc: " + str(ds_avg_train_acc))
print("cmbd_avg_train_acc: " + str(cmbd_avg_train_acc))
print()
print("sgd_avg_heldout_acc: " + str(sgd_avg_heldout_acc))
print("dt_avg_heldout_acc: " + str(dt_avg_heldout_acc))
print("ds_avg_heldout_acc: " + str(ds_avg_heldout_acc))
print("cmbd_avg_heldout_acc: " + str(cmbd_avg_heldout_acc))
print()
print("sgd_std_train_acc: " + str(sgd_std_train_acc))
print("dt_std_train_acc: " + str(dt_std_train_acc))
print("ds_std_train_acc: " + str(ds_std_train_acc))
print("cmbd_std_train_acc: " + str(cmbd_std_train_acc))
print()
print("sgd_std_heldout_acc: " + str(sgd_std_heldout_acc))
print("dt_std_heldout_acc: " + str(dt_std_heldout_acc))
print("ds_std_heldout_acc: " + str(ds_std_heldout_acc))
print("cmbd_std_heldout_acc: " + str(cmbd_std_heldout_acc))
print()
print("sgd_train_acc_ci: " + str(sgd_train_acc_ci))
print("dt_train_acc_ci: " + str(dt_train_acc_ci))
print("ds_train_acc_ci: " + str(ds_train_acc_ci))
print("cmbd_train_acc_ci: " + str(cmbd_train_acc_ci))
print()
print("sgd_heldout_acc_ci: " + str(sgd_heldout_acc_ci))
print("dt_heldout_acc_ci: " + str(dt_heldout_acc_ci))
print("ds_heldout_acc_ci: " + str(ds_heldout_acc_ci))
print("cmbd_heldout_acc_ci: " + str(cmbd_heldout_acc_ci))
sgd avg train acc: 0.78583333333333334
dt_avg_train_acc: 1.0
ds avg train acc: 0.7614285714285713
cmbd_avg_train_acc: 0.8540476190476192
sgd_avg_heldout_acc: 0.5823809523809523
dt_avg_heldout_acc: 0.6757142857142857
ds_avg_heldout_acc: 0.7047619047619047
cmbd_avg_heldout_acc: 0.7595238095238095
sgd_std_train_acc: 0.012356890537840505
dt_std_train_acc: 0.0
ds_std_train_acc: 0.007801026154955195
cmbd std train acc: 0.04845603706809577
sgd std heldout acc: 0.042527835355744975
dt std heldout acc: 0.03046874909151694
ds std heldout acc: 0.013964645998681122
```

```
cmbd_std_heldout_acc: 0.052011685413002474
     sgd_train_acc_ci: [0.7750020412179444, 0.7966646254487224]
     dt_train_acc_ci: [1.0, 1.0]
     ds train acc ci: [0.7545906705160917, 0.7682664723410509]
     cmbd_train_acc_ci: [0.8115740298682506, 0.8965212082269878]
     sgd_heldout_acc_ci: [0.5451036611107374, 0.6196582436511673]
     dt heldout acc ci: [0.6490072496043431, 0.7024213218242283]
     ds_heldout_acc_ci: [0.6925213528498743, 0.7170024566739351]
     cmbd_heldout_acc_ci: [0.7139335571543488, 0.8051140618932702]
[27]: import os
      import matplotlib.pyplot as plt
      def plot results(sgd train_acc, sgd_train_std, sgd_heldout_acc,__
       →sgd_heldout_std, sgd_test_acc,
                       dt_train_acc, dt_train_std, dt_heldout_acc, dt_heldout_std,__
       →dt_test_acc,
                       dt4_train_acc, dt4_train_std, dt4_heldout_acc,__

→dt4_heldout_std, dt4_test_acc,
                       stumps_train_acc, stumps_train_std, stumps_heldout_acc,_
       →stumps_heldout_std, stumps_test_acc):
          11 11 11
          Plots the final results from problem 2. For each of the 4 classifiers, pass
          the training accuracy, training standard deviation, held-out accuracy, \Box
       \hookrightarrow held-out
          standard deviation, and testing accuracy.
          Although it should not be necessary, feel free to edit this method.
          train_x_pos = [0, 4, 8, 12]
          cv_x_pos = [1, 5, 9, 13]
          test_x_pos = [2, 6, 10, 14]
          ticks = cv_x_pos
          labels = ['sgd', 'dt', 'dt4', 'stumps (4 x 50)']
          train_accs = [sgd_train_acc, dt_train_acc, dt4_train_acc, stumps_train_acc]
          train_errors = [sgd_train_std, dt_train_std, dt4_train_std,__
       →stumps_train_std]
          cv_accs = [sgd_heldout_acc, dt_heldout_acc, dt4_heldout_acc,_
       →stumps_heldout_acc]
          cv_errors = [sgd_heldout_std, dt_heldout_std, dt4_heldout_std,__
       →stumps heldout std]
```

```
test_accs = [sgd_test_acc, dt_test_acc, dt4_test_acc, stumps_test_acc]
          fig, ax = plt.subplots()
          ax.bar(train_x_pos, train_accs, yerr=train_errors, align='center', alpha=0.

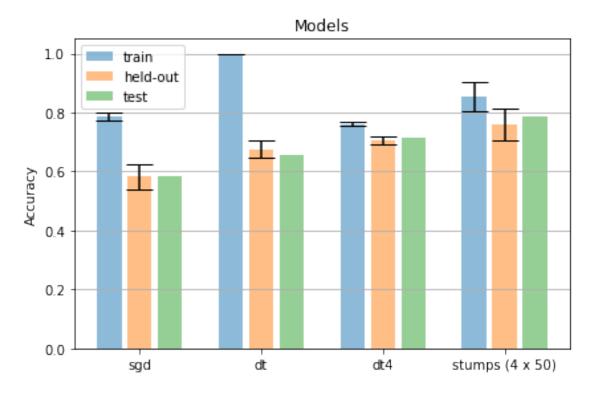
→5, ecolor='black', capsize=10, label='train')
          ax.bar(cv_x_pos, cv_accs, yerr=cv_errors, align='center', alpha=0.5,

→ecolor='black', capsize=10, label='held-out')
          ax.bar(test_x_pos, test_accs, align='center', alpha=0.5, capsize=10,__
       →label='test')
          ax.set_ylabel('Accuracy')
          ax.set xticks(ticks)
          ax.set_xticklabels(labels)
          ax.set_title('Models')
          ax.yaxis.grid(True)
          ax.legend()
          plt.tight_layout()
          plt.savefig('ian_plt.png')
[16]: sgd_testing_results = train_and_evaluate_sgd(madelon_x_train, madelon_y_train,
      →madelon_x_test, madelon_y_test)
      dt_testing results = train and_evaluate_decision_tree(madelon_x_train,__
       →madelon_y_train, madelon_x_test, madelon_y_test)
      ds testing results = train and evaluate decision stump(madelon x train,
       →madelon_y_train, madelon_x_test, madelon_y_test)
      cmbd_testing_results = train_and_evaluate_sgd_with_stumps(madelon_x_train,_
       →madelon_y_train, madelon_x_test, madelon_y_test)
      print(sgd_testing_results)
      print(dt_testing_results)
      print(ds_testing_results)
      print(cmbd_testing_results)
     (0.7728571428571429, 0.585, SGDClassifier(loss='log', max_iter=10000))
     (1.0, 0.655, DecisionTreeClassifier(criterion='entropy'))
     (0.7538095238095238, 0.715, DecisionTreeClassifier(criterion='entropy',
     max_depth=4))
     (0.856666666666667, 0.785, SGDClassifier(loss='log', max iter=10000), array([[
     62, 565, 303, ..., 54, 108, 124],
            [ 64, 232, 235, ..., 387, 35, 446],
            [328, 311, 433, ..., 318, 475, 333],
            [564, 322, 130, ..., 212, 266, 33],
            [414, 355, 594, ..., 73, 35, 250],
            [ 14, 434, 143, ..., 521, 82, 105]]),
     array([DecisionTreeClassifier(criterion='entropy', max depth=4),
            DecisionTreeClassifier(criterion='entropy', max_depth=4),
```

```
DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
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DecisionTreeClassifier(criterion='entropy', max depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
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DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
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DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
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DecisionTreeClassifier(criterion='entropy', max depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
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DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4)],
dtype=object))
```



```
[18]: def generate_txt_predictions(trained_model, X, filename):

"""

This function will write the predictions txt files needed for your

→ prediction submissions. You can access

your trained model by following the suggested return values in

→ train_and_evaluate_sgd_with_stumps(). You

should also be careful to write the correct filename as described in the

→ write up.

Parameters
```

```
trained\_model: sklearn.base.BaseEstimator
              These are the sklearn models that you trained above on the training \Box
       \hookrightarrow data.
          X_leaderboard: np.array
              The leaderboard features of shape (N leaderboard, k)
          filename: String
              This is the name of the resulting txt file.
          11 11 11
          predicted_labels = trained_model.predict(X)
          np.savetxt("{}.txt".format(filename), predicted_labels, fmt='%i',__
       →newline="\n")
[19]: # Making predictions for unseen madelon datasets
      madelon_leaderboard_x = np.load('madelon/leaderboard/X_leaderboard.npy')
      madelon_hidden_x = np.load('madelon/hidden/X_hidden.npy')
      madelon_final_model = cmbd_testing_results[2]
      madelon_final_feature_set = cmbd_testing_results[3]
      madelon_final_stumps = cmbd_testing_results[4]
      madelon_leaderboard_input_set =_

→create_stump_pred_features(madelon_leaderboard_x,
                                                                 ш
      →madelon_final_feature_set,
                                                                  madelon_final_stumps)
      madelon_hidden_input_set = create stump_pred_features(madelon_hidden_x,
       →madelon_final_feature_set,
                                                                  madelon_final_stumps)
      # generate_txt_predictions(madelon_final_model, madelon_leaderboard_input_set,
      #
                                 'labels_2a_leaderboard')
      # generate txt predictions (madelon final model, madelon hidden input set,
                                 'labels_2a_hidden')
[20]: # Running models for badges data
      badges_x_train_features = compute_features(badges_x_train_transformed)
      badges_x_test_features = compute_features(badges_x_test_transformed)
```

sgd\_badge = train\_and\_evaluate\_sgd(badges\_x\_train\_features, badges\_y\_train,

```
badges_x_test_features, badges_y_test)
tree_badge = train and_evaluate_decision_tree(badges x_train_features,__
 →badges_y_train,
                             badges_x_test_features, badges_y_test)
stump_badge = train_and_evaluate_decision_stump(badges_x_train_features,_
 →badges y train,
                             badges_x_test_features, badges_y_test)
combined badge = train_and_evaluate_sgd_with_stumps(badges_x_train_features,__
 →badges_y_train,
                             badges_x_test_features, badges_y_test)
print("sgd_badge_results: " + str(sgd_badge))
print("dt_badge_results: " + str(tree_badge))
print("ds_badge_results: " + str(stump_badge))
print("cmbd_badge_results: " + str(combined_badge))
sgd_badge_results: (0.687, 0.575, SGDClassifier(loss='log', max_iter=10000))
dt_badge_results: (1.0, 0.455, DecisionTreeClassifier(criterion='entropy'))
ds_badge_results: (0.55, 0.5, DecisionTreeClassifier(criterion='entropy',
max depth=4))
cmbd_badge_results: (0.597, 0.57, SGDClassifier(loss='log', max_iter=10000),
array([[146, 64, 31, ..., 230, 52, 202],
       [132, 91, 135, ..., 5, 36, 241],
       [139, 165, 151, ..., 147, 100, 149],
       [ 63, 106, 79, ..., 176, 168, 46],
               8, 90, ..., 223, 173, 167],
       [ 22,
       [140, 22, 156, ..., 92, 149, 37]]),
array([DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
       DecisionTreeClassifier(criterion='entropy', max_depth=4),
```

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DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max depth=4),
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DecisionTreeClassifier(criterion='entropy', max_depth=4),
DecisionTreeClassifier(criterion='entropy', max depth=4),
DecisionTreeClassifier(criterion='entropy', max_depth=4),
```

```
DecisionTreeClassifier(criterion='entropy', max_depth=4),
            DecisionTreeClassifier(criterion='entropy', max_depth=4),
            DecisionTreeClassifier(criterion='entropy', max depth=4),
            DecisionTreeClassifier(criterion='entropy', max_depth=4),
            DecisionTreeClassifier(criterion='entropy', max depth=4),
            DecisionTreeClassifier(criterion='entropy', max_depth=4),
            DecisionTreeClassifier(criterion='entropy', max depth=4),
            DecisionTreeClassifier(criterion='entropy', max_depth=4),
            DecisionTreeClassifier(criterion='entropy', max_depth=4)],
           dtype=object))
[21]: # Making predictions for badges data
      badges_leaderboard_x = np.loadtxt('badges/leaderboard/leaderboard_names_badges.
       →txt', dtype=str)
      badges_hidden_x = np.loadtxt('badges/hidden/hidden_names_badges.txt', dtype=str)
      badges_x_leaderboard_transformed = np.empty(len(badges_leaderboard_x), dtype=np.

dtype('U100'))
      badges_x_hidden_transformed= np.empty(len(badges_hidden_x), dtype=np.
       →dtype('U100'))
```

DecisionTreeClassifier(criterion='entropy', max\_depth=4),
DecisionTreeClassifier(criterion='entropy', max\_depth=4),

for i in range(0, len(badges\_leaderboard\_x)):

```
badges_x_leaderboard_transformed[i] = (badges_leaderboard_x[i][0] + " " +__
       →badges_leaderboard_x[i][1])
      for i in range(0, len(badges hidden x)):
          badges_x_hidden_transformed[i] = badges_hidden_x[i][0] + " " +__
       →badges hidden x[i][1]
      badges_x_leaderboard_regular_features =_
      →compute_features(badges_x_leaderboard_transformed)
      badges_x_hidden_regular_features = compute_features(badges_x_hidden_transformed)
      # generate_txt_predictions(sqd_badge[2], badges_x_leaderboard_regular_features,
                                'labels_2b_leaderboard')
      # generate txt predictions(sqd badge[2], badges x hidden regular features,
                                'labels_2b_hidden')
[37]: # Enhanced compute_features function, with more features added
      def compute_features_enhanced(names):
          feature_array = np.zeros(shape=(len(names), 261), dtype=int)
          for i in range (0, len(names)):
              name_split = names[i].lower().split()
              for j in range (0,2):
                  for k in range (0, min(5, len(name_split[j]))):
                      buffer = (26*k) + (130*j) - 97
                      position = ord(name_split[j][k]) + buffer
                      feature_array[i][position] = 1
          for a in range (0, len(names)):
              name_split = names[a].lower().split()
              counter=0
              for b in range (0,2):
                  for c in range (0, len(name_split[b])):
```

letter = name\_split[b][c]

```
[38]: # Making predictions for badges data using enhanced compute features function

# print(badges_x_train_transformed)

# print(badges_y_train)

# for i in range (0,1000):

# print(str(badges_x_train_transformed[i]) + ": " +□

→str(badges_y_train[i]))
```

```
[39]: enh_badges_x_train_features =
      →compute_features_enhanced(badges_x_train_transformed)
      enh_badges_x_test_features =_
       →compute features enhanced(badges x test transformed)
      enh_sgd_badge = train_and_evaluate_sgd(enh_badges_x_train_features,_
      →badges_y_train,
                                   enh_badges_x_test_features, badges_y_test)
      enh_tree_badge = train_and_evaluate_decision_tree(enh_badges_x_train_features,_
       →badges y train,
                                   enh_badges_x_test_features, badges_y_test)
      enh_stump_badge =
       →train_and_evaluate_decision_stump(enh_badges_x_train_features,
      →badges_y_train,
                                   enh_badges_x_test_features, badges_y_test)
      # enh_combined_badge =_
      → train_and_evaluate_sqd_with_stumps(enh_badges_x_train_features, __
      \rightarrow badges_y_train,
                                     enh_badges_x_test_features, badges_y_test)
      print("enh_sgd_badge_results: " + str(enh_sgd_badge))
      print("enh_dt_badge_results: " + str(enh_tree_badge))
      print("enh_ds_badge_results: " + str(enh_stump_badge))
      # print("enh cmbd badge results: " + str(enh combined badge))
```

enh\_sgd\_badge\_results: (1.0, 1.0, SGDClassifier(loss='log', max\_iter=10000))
enh\_dt\_badge\_results: (1.0, 1.0, DecisionTreeClassifier(criterion='entropy'))
enh\_ds\_badge\_results: (0.996, 1.0, DecisionTreeClassifier(criterion='entropy',

```
max_depth=4))
[32]: enhanced_badges_leaderboard_x_features =
       →compute_features_enhanced(badges_x_leaderboard_transformed)
      enhanced_badges_hidden_x_features =_
      →compute_features_enhanced(badges_x_hidden_transformed)
      # generate_txt_predictions(enh_tree_badge[2],__
       →enhanced_badges_leaderboard_x_features,
                                'labels ec leaderboard')
      # generate_txt_predictions(enh_tree_badge[2],__
       →enhanced_badges_hidden_x_features,
                                'labels_ec_hidden')
[33]: # Generating all predictions
      generate_txt_predictions(madelon_final_model, madelon_leaderboard_input_set,
                              'labels_2a_leaderboard')
      generate_txt_predictions(madelon_final_model, madelon_hidden_input_set,
                              'labels_2a_hidden')
      generate_txt_predictions(sgd_badge[2], badges_x_leaderboard_regular_features,
                              'labels_2b_leaderboard')
      generate_txt_predictions(sgd_badge[2], badges_x_hidden_regular_features,
                              'labels 2b hidden')
      generate_txt_predictions(enh_tree_badge[2],__
       →enhanced_badges_leaderboard_x_features,
                              'labels_ec_leaderboard')
      generate_txt_predictions(enh_tree_badge[2], enhanced_badges_hidden_x_features,
                              'labels ec hidden')
```

[]:

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