Neural Shrubs - Leaves

July 18, 2019

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In [1]: import time
        from sklearn.tree import DecisionTreeRegressor
        from keras.models import Sequential
        from keras.layers import Dense
        from sklearn.preprocessing import MinMaxScaler
        from sklearn import preprocessing
        from sklearn.metrics import mean_absolute_error
        from sklearn.metrics import mean_squared_error
        import numpy as np
        import csv
        import os
/home/shashwati/anaconda3/envs/py35/lib/python3.5/site-packages/h5py/__init__.py:36: FutureWar
  from ._conv import register_converters as _register_converters
Using TensorFlow backend.
In [2]: sc= MinMaxScaler()
In [3]: # function returns the data in the right format
        def get_data():
            # load training dataset
            dataset = np.genfromtxt("YearPredictionMSD.csv", dtype='float', delimiter=",")
            train_X = dataset[0:463715,1:91]
            train_Y = dataset[0:463715,0]
            test_X = dataset[463715:,1:91]
            test_Y = dataset[463715:,0]
            return train_X, train_Y, test_X, test_Y
In [4]: # builds the decision tree of depth 13
        def regression_tree(train, label):
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dt = DecisionTreeRegressor(max_depth=13, min_samples_leaf=5000)
            dt.fit(train, label)
            return dt
In [5]: # builds the neural network for a given class
        def neural_network(class_data):
            num_train = []
            num_label = []
            for x in class_data:
                num_train.append(x[0])
                num_label.append(x[1])
            num_train = np.array(num_train)
            num_label = np.array(num_label)
            # Scale the features so they have 0 mean
            num_train = preprocessing.scale(num_train)
            num_label = num_label.reshape(-1,1)
            num_label = sc.fit_transform(num_label)
            model = Sequential()
            model.add(Dense(90, input_dim=90, kernel_initializer='normal', activation='relu'))
            model.add(Dense(90, kernel_initializer='normal', activation='relu'))
            model.add(Dense(90, kernel_initializer='normal', activation='relu'))
            model.add(Dense(90, kernel_initializer='normal', activation='relu'))
            model.add(Dense(1, kernel_initializer='normal', activation='linear'))
            model.compile(loss='mean_squared_error', optimizer='adam', metrics=['mae', 'accurac']
            model.fit(num_train, num_label, epochs=3, batch_size=32)
            return model
In [6]: # builds the neural shrub
        def neural_shrubs(tree, train, label):
            train = np.array(train)
            label = np.array(label)
            # leave_id: index of the leaf that cantains the instance
            leave_id = tree.apply(train)
            classes = dict()
            for x in range(len(train)):
                leaf = leave_id[x]
                # Gets the class for each leaf
                #.value: contains value of all the tree nodes
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# stores the neural network for each class
         nn_models = dict()
          #stores the max time taken to build a neural network
         max_time = 0;
         for key in classes.keys():
             start = time.time()
             model = neural_network(classes[key])
             end = time.time()
             time_taken = end - start
             if max_time < time_taken:</pre>
                max_time = time_taken
             nn_models[key] = model
          # returns a neural network for each class and the max
          # time taken to build the neural network
         return nn_models, max_time
In [7]: # The algorithm to build the decision tree
      train, train_label, test, test_label = get_data()
      dt_start = time.time()
      tree = regression_tree(train, train_label)
      dt_end = time.time()
      print("Decision tree made in: ", dt_end-dt_start)
Decision tree made in: 32.71749544143677
In [14]: # Neural shrub
       shrubs, max_time = neural_shrubs(tree, train, train_label)
Epoch 1/3
Epoch 2/3
3
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#.value[leaf]: returns the value of the leaf

classes[leaf].append([train[x], label[x]])

classes[leaf] = [[train[x], label[x]]]

#idx = tree.tree_.value[leaf][0][0]

insert the instance into the class

if leaf in classes.keys():

else:

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Epoch 3/3
Epoch 1/3
Epoch 2/3
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Epoch 3/3
Epoch 1/3
Epoch 2/3
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Epoch 3/3
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Epoch 1/3
Epoch 2/3
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Epoch 3/3
Epoch 1/3
Epoch 2/3
Epoch 3/3
In [15]: # training time
  total_time = dt_end - dt_start + max_time
  print("Training time: ", round(total_time, 5))
Training time: 47.95906
In [16]: # predicts using the neural shrub
  def neural_shrub_predict(tree, nn_model, test):
    test = np.asarray(test)
    pred = []
    # getting the predicted value
    tree_pred = tree.apply(test)
    test = preprocessing.scale(test)
    for i in range(len(test)):
     # gets the neural network for the tree predicted val
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x = tree_pred[i]
                 nn_model_class = nn_model[x]
                 # preprocessing
                 1 = np.asarray(test[i])
                 l = l.reshape(-1, 90)
                 # gets the val using the nn associated with
                 # the current predicted value
                 ans = nn_model_class.predict(1)
                 pred.append(ans)
             return pred
In [17]: # predicting for test data
         # results scaled between 0 - 1
         pred = neural_shrub_predict(tree, shrubs, test)
In [18]: pred = np.asarray(pred)
         pred = pred.ravel()
         pred = pred.reshape(-1,1)
         pred = sc.inverse_transform(pred)
         print(mean_absolute_error(test_label, pred), mean_squared_error(test_label, pred))
6.7961220344692705 99.47642204472457
In [19]: print(test_label[0:10])
         print(pred[0:10])
[2007. 2003. 2005. 2003. 2005. 2007. 2003. 2003. 2003. 2005.]
[[1996.5908]
 [1999.8171]
 [2002.4985]
 [2005.3513]
 [2005.3563]
 [1999.4575]
 [1998.3212]
 [1994.4529]
 [1998.7758]
 [1999.3042]]
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