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import operator

import numpy as np
from past.builtins import xrange

class KNearestNeighbor(object):
    """ a KNN classifier with L2 distance """

    def __init__(self):
        pass

    def train(self, X, y):
        """
        Train the classifier. For k-nearest neighbors this is just
        memorizing the training data.

        Inputs:
        - X: A numpy array of shape (num_train, D) containing the training data
            consisting of num_train samples each of dimension D.
        - y: A numpy array of shape (N,) containing the training labels, where
            y[i] is the label for X[i].
        """
        self.X_train = X # m here (5000, 3072)
        self.y_train = y # m here (500,)

    def predict(self, X, k=1, num_loops=0):
        """
        Predict labels for test data using this classifier.

        Inputs:
        - X: A numpy array of shape (num_test, D) containing test data consisting
            of num_test samples each of dimension D.
        - k: The number of nearest neighbors that vote for the predicted labels.
        - num_loops: Determines which implementation to use to compute distances
            between training points and testing points.

        Returns:
        - y: A numpy array of shape (num_test,) containing predicted labels for the
            test data, where y[i] is the predicted label for the test point X[i].
        """
        if num_loops == 0:
            dists = self.compute_distances_no_loops(X)
        elif num_loops == 1:
            dists = self.compute_distances_one_loop(X)
        elif num_loops == 2:
            dists = self.compute_distances_two_loops(X)
        else:
            raise ValueError('Invalid value %d for num_loops' % num_loops)

        return self.predict_labels(dists, k=k)

    def compute_distances_two_loops(self, X):
        """
        Compute the distance between each test point in X and each training point
        in self.X_train using a nested loop over both the training data and the
        test data.

        Inputs:
        - X: A numpy array of shape (num_test, D) containing test data.
            # m here (500, 3072)

        Returns:
        - dists: A numpy array of shape (num_test, num_train) where dists[i, j]
            # m here (500, 5000)
            is the Euclidean distance between the ith test point and the jth training
            point.
        """
        num_test = X.shape[0] # m here num_test=500

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num_train = self.X_train.shape[0] # m here num_train=5000
dists = np.zeros((num_test, num_train))
for i in xrange(num_test): # m 0 to 499
    for j in xrange(num_train): # m 0 to 4999
        #####
        # m completed
        # Compute the l2 distance between the ith test point and the jth #
        # training point, and store the result in dists[i, j]. You should #
        # not use a loop over dimension. #
        #####
        test_image = X[i]
        memorized_training_image = self.X_train[j]
        dists[i, j] = np.sqrt(np.sum(np.square(test_image - memorized_training_image)))
        #####
        #                                     END OF YOUR CODE #
        #####
    return dists

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def compute_distances_one_loop(self, X):
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    """
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Compute the distance between each test point in X and each training point in self.X_train using a single loop over the test data.

Input / Output: Same as compute_distances_two_loops

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    """
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    num_test = X.shape[0]
    num_train = self.X_train.shape[0]
    dists = np.zeros((num_test, num_train))
    for i in xrange(num_test):
        #####
        # m done but takes longer than two loops:
        #
        # Compute the l2 distance between the ith test point and all training #
        # points, and store the result in dists[i, :]. #
        #####
        test_image = X[i]
        all_memorized_training_images = self.X_train
        dists[i, :] = np.sqrt(np.sum(np.square(all_memorized_training_images - test_image),
axis=1))
        #####
        #                                     END OF YOUR CODE #
        #####
    return dists

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def compute_distances_no_loops(self, X):
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    """
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Compute the distance between each test point in X and each training point in self.X_train using no explicit loops.

Input / Output: Same as compute_distances_two_loops

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    """
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    num_test = X.shape[0]
    num_train = self.X_train.shape[0]
    dists = np.zeros((num_test, num_train))
    #####
    # m done:
    # Compute the l2 distance between all test points and all training #
    # points without using any explicit loops, and store the result in #
    # dists. #
    #
    # You should implement this function using only basic array operations; #
    # in particular you should not use functions from scipy. #
    #
    # HINT: Try to formulate the l2 distance using matrix multiplication #
    #       and two broadcast sums. #
    #####
    # dists = np.sqrt(np.sum(X**2, axis=1).reshape(num_test, 1) + np.sum(self.X_train**2,
axis=1) - 2 * X.dot(self.X_train.T))
    a_squared = np.sum(X ** 2, axis=1).reshape(num_test, 1)

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b_squared = np.sum(self.X_train ** 2, axis=1)
two_a_b = 2 * X.dot(self.X_train.T)
dists = np.sqrt(a_squared + b_squared - two_a_b)
#####
#                               END OF YOUR CODE                               #
#####
return dists

def predict_labels(self, dists, k=1):
    """
    Given a matrix of distances between test points and training points,
    predict a label for each test point.

    Inputs:
    - dists: A numpy array of shape (num_test, num_train) where dists[i, j]
      gives the distance between the ith test point and the jth training point.

    Returns:
    - y: A numpy array of shape (num_test,) containing predicted labels for the
      test data, where y[i] is the predicted label for the test point X[i].
    """
    # m here dists = (500, 5000)
    # m here return is (500,)
    num_test = dists.shape[0]
    y_pred = np.zeros(num_test)
    for i in xrange(num_test):
        # A list of length k storing the labels of the k nearest neighbors to
        # the ith test point.
        closest_y = []
        #####
        # m done
        # Use the distance matrix to find the k nearest neighbors of the ith
        # testing point, and use self.y_train to find the labels of these
        # neighbors. Store these labels in closest_y.
        # Hint: Look up the function numpy.argsort.
        #####
        sorted_indices = np.argsort(dists[i])
        selected_closest_neighbours = list(sorted_indices[0:k])
        for neighbour in selected_closest_neighbours:
            closest_y.append(self.y_train[neighbour])
        #####
        # m done:
        # Now that you have found the labels of the k nearest neighbors, you
        # need to find the most common label in the list closest_y of labels.
        # Store this label in y_pred[i]. Break ties by choosing the smaller
        # label.
        #####
        y_pred[i] = self.mostCommonLabel(closest_y)
        #####
        #                               END OF YOUR CODE                               #
        #####

    return y_pred

@staticmethod
def mostCommonLabel(items):
    counts = {}
    counted = []
    for item in items:
        if item in counted:
            continue
        counted.append(item)
        counts[item] = (items.count(item))

    sorted_items = sorted(counts.items(), key=operator.itemgetter(1))
    sorted_items.reverse()
    first_item = sorted_items[0][0]
    highest = first_item
    try:

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second_item = sorted_items[1][0]
if sorted_items[0][1] == sorted_items[1][1]:
    length_of_first_item = len(first_item)
    length_of_second_item = len(second_item)

    if length_of_first_item > length_of_second_item:
        print("first item is", first_item, "length is", length_of_first_item)
        print("second item is", second_item, "length is", length_of_second_item)
        print(first_item, "length is greater than", second_item, "length")
        highest = second_item
        print("highest is ", highest)
    return highest
except:
    return highest
```