

Machine Learning Homework 02

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1 Problem 8

Prove or disprove: Consider two arbitrary points $x, y \in \mathbb{R}^2$. If x is the nearest neighbor of y regarding the L2-norm, then x is the nearest neighbor of y regarding the L1-norm.

1.1 Solution

We will use the following points as a counterexample:

- Target point: y = (0,0)
- Point $x_1 = (2,0)$
- Point $x_2 = (1.5, 1)$

For the L2 - distance between y = (0,0) and $x_1 = (2,0)$:

$$d_2(y, x_1) = \sqrt{(2-0)^2 + (0-0)^2} = \sqrt{4} = 2$$

For the L2 - distance between y = (0,0) and $x_2 = (1.5,1)$:

$$d_2(y, x_2) = \sqrt{(1.5 - 0)^2 + (1 - 0)^2} = \sqrt{2.25 + 1} = \sqrt{3.25} \approx 1.803$$

Result in L2 norm: The nearest neighbor to y = (0,0) in the L2 norm is $x_2 = (1.5,1)$, as it has a smaller distance of approximately 1.803, compared to the distance of 2 for $x_1 = (2,0)$.

For the L1 - distance between y = (0,0) and $x_1 = (2,0)$:

$$d_1(y, x_1) = |2 - 0| + |0 - 0| = 2 + 0 = 2$$

For the L1 - distance between y = (0,0) and $x_2 = (1.5,1)$:

$$d_1(y, x_2) = |1.5 - 0| + |1 - 0| = 1.5 + 1 = 2.5$$

Result in L1 norm: The nearest neighbor to y = (0,0) in the L1 norm is $x_1 = (2,0)$, as it has a smaller distance of 2, compared to the distance of 2.5 for $x_2 = (1.5,1)$.

- In the L2 norm, the nearest neighbor to y = (0,0) is $x_2 = (1.5,1)$ with a distance of approximately 1.803.
- In the L1 norm, the nearest neighbor to y = (0,0) is $x_1 = (2,0)$ with a distance of 2.

This disproves the statement that if x is the nearest neighbor of y in the L2 norm, then x must also be the nearest neighbor in the L1 norm.

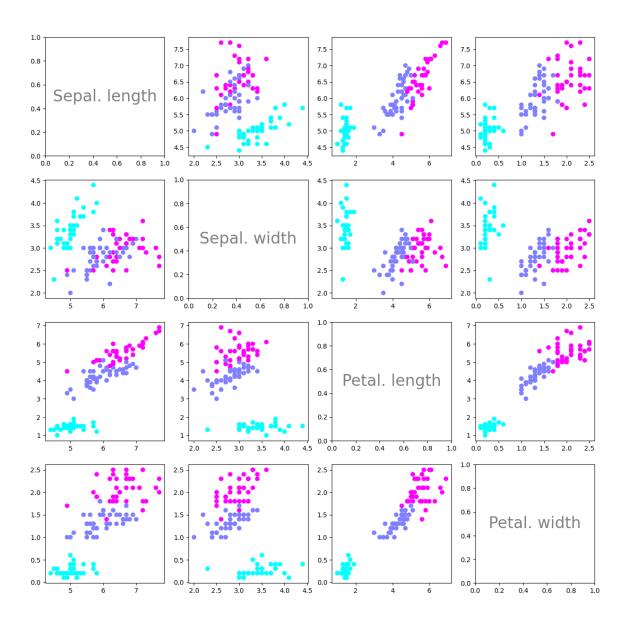
2 Problem 9

2.1 Load dataset (Preparation)



```
dataset = datasets.load_iris()
   X, y = dataset['data'], dataset['target']
   X_train, X_test, y_train, y_test = model_selection.train_test_split(X, y, random_state=123,
                                                    test_size=(1 - split))
   return X_train, X_test, y_train, y_test
# prepare data
split = 0.75
X_train, X_test, y_train, y_test = load_dataset(split)
f, axes = plt.subplots(4, 4, figsize=(15, 15))
for i in range(4):
   for j in range(4):
       if j == 0 and i == 0:
           axes[i,j].text(0.5, 0.5, 'Sepal. length', ha='center', va='center', size=24, alpha=.5)
        elif j == 1 and i == 1:
           axes[i,j].text(0.5, 0.5, 'Sepal. width', ha='center', va='center', size=24, alpha=.5)
        elif j == 2 and i == 2:
           axes[i,j].text(0.5, 0.5, 'Petal. length', ha='center', va='center', size=24, alpha=.5)
        elif j == 3 and i == 3:
           axes[i,j].text(0.5, 0.5, 'Petal. width', ha='center', va='center', size=24, alpha=.5)
           axes[i,j].scatter(X_train[:,j],X_train[:,i], c=y_train, cmap=plt.cm.cool)
```





2.2 Task: Euclidean distance



2.3 Task: get k nearest neighbors' labels

```
def get_neighbors_labels(X_train, y_train, X_new, k):
Get the labels of the k nearest neighbors of the datapoints x\_{new}.
Parameters
X_train : array, shape (N_train, D)
    Training features.
y_train : array, shape (N_train,)
    Training labels.
X_new : array, shape (M, D)
   Data points for which the neighbors have to be found.
k : int
    Number of neighbors to return.
Returns
neighbors_labels : array, shape (M, k)
Array containing the labels of the k nearest neighbors for each data point in X_new.
distances = np.sqrt(((X_new[:, np.newaxis, :] - X_train) ** 2).sum(axis=2))
nearest_indices = np.argsort(distances, axis=1)[:, :k]
neighbors_labels = y_train[nearest_indices]
return neighbors_labels
```

2.4 Task: get the majority label

```
def get_response(neighbors_labels, num_classes=3):
"""Predict label given the set of neighbors.
Parameters
neighbors_labels : array, shape (M, k)
   Array containing the labels of the k nearest neighbors per data point.
num_classes : int
   Number of classes in the dataset.
Returns
y : int array, shape (M,)
   Majority class among the neighbors.
# TODO
M = neighbors_labels.shape[0]
predictions = np.zeros(M, dtype=int)
for i in range(M):
    counts = np.bincount(neighbors_labels[i], minlength=num_classes)
    predictions[i] = np.argmax(counts)
return predictions
```



2.5 Task: compute accuracy

```
def compute_accuracy(y_pred, y_test):
"""Compute accuracy of prediction.
Parameters
y_pred : array, shape (N_test)
   Predicted labels.
y_test : array, shape (N_test)
True labels.
# TODO
assert y_pred.shape == y_test.shape, "Shapes of y_pred and y_test must match"
correct_predictions = np.sum(y_pred == y_test)
accuracy = correct_predictions / len(y_test)
return accuracy
```

2.6 Testing

```
# This function is given, nothing to do here.
def predict(X_train, y_train, X_test, k):
    """Generate predictions for all points in the test set.
   Parameters
   X_train : array, shape (N_train, 4)
       Training features.
   y_train : array, shape (N_train)
       Training labels.
   X_test : array, shape (N_test, 4)
       Test features.
   k : int
       Number of neighbors to consider.
   Returns
   y_pred : array, shape (N_test)
   Predictions for the test data.
   neighbors = get_neighbors_labels(X_train, y_train, X_test, k)
   y_pred = get_response(neighbors)
   return y_pred
```

```
# prepare data
split = 0.75
X_train, X_test, y_train, y_test = load_dataset(split)
print('Training set: {0} samples'.format(X_train.shape[0]))
print('Test set: {0} samples'.format(X_test.shape[0]))
# generate predictions
k = 3
y_pred = predict(X_train, y_train, X_test, k)
accuracy = compute_accuracy(y_pred, y_test)
print('Accuracy = {0}'.format(accuracy))
```

2.7 Result:

Training set: 112 samples Test set: 38 samples

Accuracy = 0.9473684210526315