Lab 07: Handwritten Digit Recognition Using MLPClassifier 1. Import Necessary imports
<pre>import numpy as np import seaborn as sns import matplotlib.pyplot as plt from scipy.io import loadmat from sklearn.neural_network import MLPClassifier</pre>
from sklearn import metrics 2. Upload Dataset In [4]: mnist_raw = loadmat("mnist-original.mat")
MNIST is a database. The acronym stands for "Modified National Institute of Standards and Technology." The MNIST database contains handwritten digits (0 through 9), and can provide a baseling for testing image processing systems. mnist_raw file is a matlab file. It is a dictionary containing some key value pairs. Lets have a look
<pre>In [5]: mnist_raw Out[5]: {'header': b'MATLAB 5.0 MAT-file Platform: posix, Created on: Tue Jan 19 16:10:39 2016',</pre>
<pre>dtype=object), 'data': array([[0, 0, 0,, 0, 0, 0],</pre>
[0, 0, 0,, 0, 0, 0], [0, 0, 0,, 0, 0, 0]], dtype=uint8), 'label': array([[0., 0., 0.,, 9., 9.]])} We are interested in data and labels In [6]: mnist_raw["data"]
Out[6]: array([[0, 0, 0,, 0, 0, 0],
[0, 0, 0,, 0, 0, 0]], dtype=uint8) In [7]: mnist_raw["data"].shape Out[7]: (784, 70000)
Each image is stored in a column. Each row represnts each pixcel value . Each image is 28 X 28. so there are total 784 pixel values in an image. This format is not consistent. In [8]: mnist_raw["data"].T.shape Out[8]: (70000, 784)
<pre>In [9]: mnist_raw["label"] Out[9]: array([[0., 0., 0.,, 9., 9.]]) In [10]: mnist_raw["label"].T.shape</pre>
Out[10]: (70000, 1) In [11]: X,y = mnist_raw['data'].T, mnist_raw["label"].T X.shape, y.shape Out[11]: ((70000, 784), (70000, 1))
3. Train Test Split In [12]: shuffle_index = np.random.permutation(70000)
<pre>In [13]: from sklearn.model_selection import train_test_split X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.1, random_state=101)</pre> In [14]: y
Out[14]: array([[1.],
[1.]]) In [15]: X.shape Out[15]: (70000, 784)
In [16]: y.shape Out[16]: (70000, 1) 4. Dataset Visualization
<pre>In [17]: some_digit = X[36000] In [18]: some_digit.shape Out[18]: (784,)</pre>
<pre>In [19]: some_digit; In [20]: some_digit_image = some_digit.reshape(28, 28) In [21]: some_digit_image;</pre>
<pre>In [22]: plt.imshow(some_digit_image, cmap = plt.cm.binary) plt.title(y[36000]) plt.show() C:\Users\Azlan\anaconda3\envs\tensorflow_env\lib\site-packages\matplotlib\text.py:1242: FutureWarning: elementwise comparison failed; returning scalar instruction.</pre>
d, but in the future will perform elementwise comparison if s != selftext: [O.]
5 -
15 -
25 -
View First 100 Training Data Points
<pre>In [23]: fig, axes = plt.subplots(10, 10, figsize=(8, 8),</pre>
<pre>ax.text(0.05, 0.05, str(int(y_train[i])),</pre>
6 5 8 9 0 5 6 6 4 3 6 5 8 9 0 5 6 6 4 3
1 8 1 8 1 3 0 9 2 7 7 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
500021995777 0588086526
756666655266
3 2 9 3 0 9 7 9 9 3
[0] 0
In [24]: X_train.shape Out[24]: (63000, 784)
<pre>In [25]: mlp = MLPClassifier(hidden_layer_sizes=(100,), max_iter=10, alpha=1e-4,</pre>
<pre>was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel(). y = column_or_1d(y, warn=True) Iteration 1, loss = 1.97830828 Iteration 2, loss = 0.95796046 Iteration 3, loss = 0.70889967 Iteration 4, loss = 0.56051496 Iteration 5, loss = 0.41836474</pre>
<pre>Iteration 6, loss = 0.35549838 Iteration 7, loss = 0.32619268 Iteration 8, loss = 0.30650824 Iteration 9, loss = 0.29068110 Iteration 10, loss = 0.28206831 C:\Users\Azlan\anaconda3\envs\tensorflow_env\lib\site-packages\sklearn\neural_network_multilayer_perceptron.py:702: ConvergenceWarning: Stochastic Optimiz</pre>
Maximum iterations (10) reached and the optimization hasn't converged yet. warnings.warn(Out[25]: MLPClassifier MLPClassifier(max_iter=10, random_state=1, solver='sgd', tol=1e-05, verbose=True)
<pre>In [26]: print("Training set score: %f" % mlp.score(X_train, y_train)) print("Test set score: %f" % mlp.score(X_test, y_test)) Training set score: 0.923921 Test set score: 0.915143</pre>
6. Prediction on Test Data In [27]: yfit = mlp.predict(x_test)
<pre>In [28]: y_test.shape Out[28]: (7000, 1) In [29]: tn = ['0','1','2','3','4','5','6','7','8','9'] print(metrics.classification_report(y_test, yfit,</pre>
target_names=tn)) precision recall f1-score support 0 0.96 0.96 0.96 690 1 0.98 0.97 0.97 820 2 0.91 0.92 0.91 695
3 0.87 0.88 0.88 679 4 0.92 0.93 0.92 685 5 0.93 0.78 0.85 624 6 0.96 0.95 0.95 664 7 0.95 0.94 0.95 739 8 0.77 0.94 0.84 673
9 0.93 0.87 0.90 731 accuracy 0.92 7000 macro avg 0.92 0.91 0.91 7000 weighted avg 0.92 0.92 0.92 7000
<pre>In [30]: mat = metrics.confusion_matrix(y_test, yfit) fig, ax = plt.subplots(figsize=(8,8)) sns.heatmap(mat.T, square=True, annot=True, fmt='d', cbar=False, cmap='viridis',</pre>
plt.xlabel('true label'); plt.ylabel('predicted label'); Confusion matrix
O - 661 0 6 3 1 3 5 1 2 5 H - 0 798 1 3 2 1 2 1 6 4
N - 0 4 636 21 8 9 2 12 7 0 N - 0 5 9 599 0 47 0 5 7 14
Dedicted label of the label of
φ - 5 1 3 1 2 11 629 0 4 1
c - 0 2 11 3 3 1 0 695 1 15 ω - 16 8 25 34 9 59 12 4 632 25
σ - 1 0 0 4 25 1 0 16 2 635 0 1 2 3 4 5 6 7 8 9 true label
<pre>In [31]: from sklearn.metrics import accuracy_score print("Accuracy", metrics.accuracy_score(y_test, yfit)*100)</pre>
<pre>import matplotlib.pyplot as plt fig, axes = plt.subplots(10, 10, figsize=(8, 8),</pre>
<pre>for i, ax in enumerate(axes.flat):</pre>
<pre>ax.imshow(X_test[i].reshape((28,28)), cmap='binary', interpolation='nearest') # actual class ax.text(0.05, 0.05, str(int(y_test[i])),</pre>
<pre>ax.imshow(X_test[i].reshape((28,28)), cmap='binary', interpolation='nearest') # actual class ax.text(0.05, 0.05, str(int(y_test[i])),</pre>
<pre>ax.imshow(X_test[i].reshape((28,28)), cmap='binary', interpolation='nearest') # actual class ax.text(0.05, 0.05, str(int(y_test[i])),</pre>
<pre>ax.imshow(X_test[i].reshape((28,28)), cmap='binary', interpolation='nearest') # actual class ax.text(0.05, 0.05, str(int(y_test[i])),</pre>
<pre>ax.imshow(X_test[i].reshape((28,28)), cmap='binary', interpolation='nearest') # actual class ax.text(0.05, 0.05, str(int(y_test[i])),</pre>
<pre>ax.imshow(X_test[i].reshape((28,28)), cmap='binary', interpolation='nearest') # actual class ax.text(0.05, 0.05, str(int(y_test[i])),</pre>
<pre>ax.imshow(X_test[i].reshape((28,28)), cmap='binary', interpolation='nearest') # actual class ax.text(0.05, 0.05, str(int(y_test[i])),</pre>