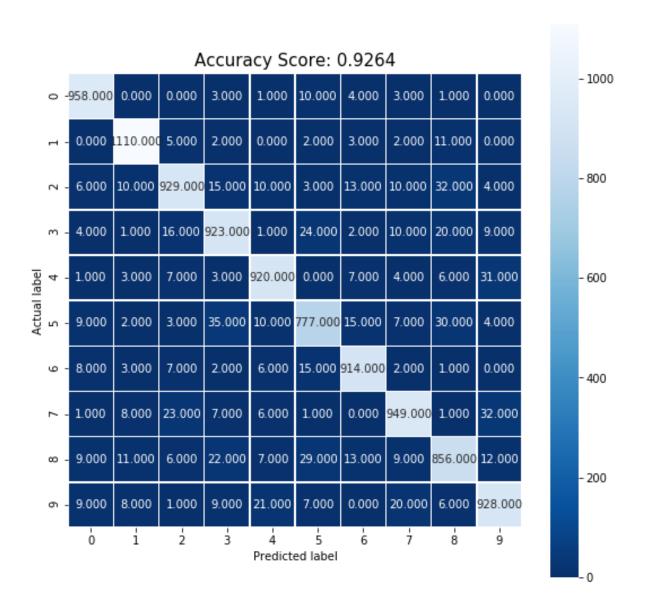
COEN 240 Machine Learning

Homework #3

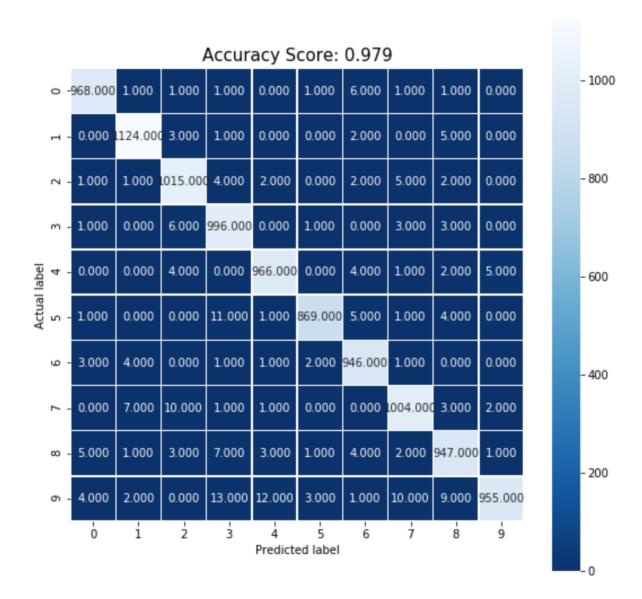
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Problem 1



Comment: The logistic regression confusion matrix shows a fairly accurate recognition score of around 92% with a fairly consistent distribution of miss labeled predictions.

Problem 2



Comment: The neural network confusion matrix shows a much better recognition accuracy than the logistic regression of around 97% and a much smaller distribution of miss labeled predictions.

Problem 3

3.	$\delta_{k} = \frac{\partial E_{n}}{\partial \alpha_{k}}$ $E_{n} = \frac{1}{2} \frac{k}{2} \left(y_{nn} - t_{nn} \right)^{2}$
	YNK= YK(Xn) YK= 1+ CE-ak, I syrind 8 h = 3En 3 ynk - 2 (Ynk - tnk) 3 ynk 3 ynk 3ak ker Ank 3ak
	$\frac{\partial y_{nn}}{\partial a_{n}} = \sigma(a_{n})(1 - \sigma(a_{n}))$ $\frac{\partial y_{nn}}{\partial a_{n}} = \sigma(a_{n})(1 - \sigma(a_{n}))$ $\frac{\partial y_{nn}}{\partial a_{n}} = \sigma(a_{n})(1 - \sigma(a_{n}))$

3.6.
$$\delta_j = \frac{\partial E_n}{\partial a_j}$$
 $= \omega_{kj} h'(a_j)$

$$= \frac{\partial E_n}{\partial a_k} \frac{\partial a_{kk}}{\partial a_j} = \delta_j = h'(a_j) \underbrace{\sum_{k} \omega_{kj} \delta_k}_{k}$$

$$h(a_j) = \tanh(a_j) = \underbrace{e^{a_j} - e^{-a_j}}_{e^{a_j} + e^{-a_j}}$$

$$h'(a_j) = 1 - h^2(a_j) = 1 - \tanh^2(a_j)$$

$$\delta_j = 1 - \tanh^2(a_j) \underbrace{\sum_{k} \omega_{kj} \left(y_{nk} - t_{nk} \right) \sigma(a_k) \left(1 - \sigma(a_k) \right)}_{k}$$

Attachment

Problem 1 Code (in zip file):

```
import tensorflow as tf
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.linear model import LogisticRegression
from sklearn.model selection import train test split
from sklearn.metrics import confusion matrix
# uploading data from mnist dataset
mnist = tf.keras.datasets.mnist
(x traino, y train),(x testo, y test) = mnist.load data()
# reshaping matrices and init the regression solver
x train = np.reshape(x traino, (60000, 28*28))
x \text{ test} = \text{np.reshape}(x \text{ testo}, (10000, 28*28))
x train, x test = x train / 255.0, x test / 255.0
logreg = LogisticRegression(solver='saga', multi class='multinomial', max iter
= 100, verbose=2)
# train the algorithm
logreg.fit(x train, y train)
# predict all values from dataset
predictions = logreg.predict(x test)
# create confusion matrix and accuracy score
predictions cm = confusion matrix(y test, predictions)
recognition accuracy rate = accuracy score(y test, predictions)
# plot using seaborn to make it look nice
plt.figure(figsize=(9,9))
sns.heatmap(predictions cm, annot=True, fmt=".3f", linewidths=.5, square =
True, cmap = 'Blues r');
plt.ylabel('Actual label');
plt.xlabel('Predicted label');
all_sample_title = 'Accuracy Score: {0}'.format(recognition_accuracy rate)
plt.title(all sample title, size = 15)
```

Problem 2 Code (in zip file):

```
import tensorflow as tf
import numpy as np
import seaborn as sns
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
import matplotlib.pyplot as plt
from sklearn.linear model import LogisticRegression
from sklearn.model selection import train test split
from sklearn.metrics import confusion matrix
from sklearn.metrics import accuracy score
# uploading data from mnist dataset
mnist = tf.keras.datasets.mnist
(x traino, y train), (x testo, y test) = mnist.load data()
# reshaping matrices and init the regression solver
x train = np.reshape(x traino, (60000, 28*28))
x \text{ test} = \text{np.reshape}(x \text{ testo,} (10000, 28*28))
x train, x test = x train / 255.0, x test / 255.0
# fix random seed for reproducibility
numpy.random.seed(7)
# create the model
# model.add(Flatten(input shape(28,28))
model = Sequential()
model.add(Dense(512, input dim=784, activation='relu')) # input 784
#model.add(Dense(512, activation='relu')) # hidden 512, reLu
model.add(Dense(10, activation='softmax')) # output 10 nodes, soft max
# compile the model
model.compile(loss='sparse categorical crossentropy', optimizer='adam',
metrics=['accuracy'])
# fit the model
model.fit(x train, y train, epochs=5, batch size=64, verbose=2)
# calculate predictions
predictions net = model.predict(x test)
# finding values from prob percentages of each category
prediction = np.argmax(predictions net, axis=1)
# create confusion matrix and accuracy score
```

```
predictions_cm = confusion_matrix(y_test, prediction)

recognition_accuracy_rate_net = accuracy_score(y_test, prediction)

# plot using seaborn to make it look nice

plt.figure(figsize=(9,9))

sns.heatmap(predictions_cm, annot=True, fmt=".3f", linewidths=.5, square =
True, cmap = 'Blues_r');

plt.ylabel('Actual label');

plt.xlabel('Predicted label');

all_sample_title = 'Accuracy Score:
{0}'.format(recognition_accuracy_rate_net)

plt.title(all_sample_title, size = 15)
```