CAP6619 Assignment #2 Question 6

```
In [ ]: %matplotlib inline
        import matplotlib.pyplot as plt
        import pandas as pd
        import numpy as np
        from sklearn.utils import shuffle
        import tensorflow as tf
        from tensorflow import keras
        print(tf.__version__)
       2.10.0
In [ ]: # Loading data into notebook
        data = np.load("olivetti_faces.npy")
        target = np.load("olivetti_faces_target.npy")
In [ ]: data.shape
Out[]: (400, 64, 64)
In [ ]: def show_face_img_for_each_class(images, unique_ids):
            fig, ax = plt.subplots(nrows=4, ncols=10, figsize=(18,9))
            ax = ax.flatten()
            rand = np.random.randint(10)
            for unique_id in unique_ids:
                image_index = unique_id * 10 + rand
                ax[unique_id].imshow(images[image_index], cmap='gray')
                ax[unique_id].set_xticks([])
                ax[unique_id].set_yticks([])
                ax[unique_id].set_title("Class {}".format(unique_id))
            plt.suptitle("A face for each class in the Olivetti faces dataset")
In [ ]: # Show at least one face image for each class in the Olivetti face dataset
        show_face_img_for_each_class(data, np.unique(target))
```



Model: "sequential"

Layer (type)	Output Shape	Param #
Input (Flatten)	(None, 4096)	0
Hidden (Dense)	(None, 10)	40970
Output (Dense)	(None, 40)	440
Total params: 41,410 Trainable params: 41,410	=======================================	=======

Non-trainable params: 0

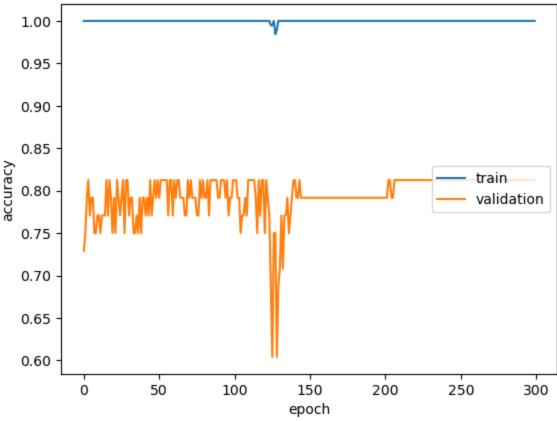
Layer (type) Output Shape Param # ______ Input (Flatten) (None, 4096) Hidden (Dense) (None, 10) 40970 Output (Dense) (None, 40) 440

Total params: 41,410 Trainable params: 41,410 Non-trainable params: 0

```
In [ ]: # Train a one-hidden layer neural network with 10 hidden nodes.
        model.compile(optimizer='adam',
                      loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),
                      metrics=['accuracy'])
        history = model.fit(X_train, y_train, validation_split=0.2, epochs=300, verbose=1)
```

```
In [ ]: plt.plot(history.history['accuracy'])
        plt.plot(history.history['val_accuracy'])
        plt.title('model accuracy')
        plt.ylabel('accuracy')
        plt.xlabel('epoch')
        plt.legend(['train', 'validation'], loc='center right')
        plt.show()
```

model accuracy



```
In [ ]: # Report the classification accuracy of the classifier on the test set
        test_loss, test_acc = model.evaluate(X_test, y_test, verbose=2)
        print('\nTest accuracy:', test_acc)
       5/5 - 0s - loss: 1.4169 - accuracy: 0.7250 - 35ms/epoch - 7ms/step
       Test accuracy: 0.7250000238418579
       Test accuracy: 0.7250000238418579
In [ ]: # Use one time 10-fold cross validation to compare the performance
        # using different neural network architectures
In [ ]: # cross-validation framework
        from sklearn.model_selection import KFold
        kf = KFold(n_splits=10, shuffle=True)
        kf.get_n_splits(data)
        Acc = []
In [ ]: from sklearn.metrics import accuracy score
In [ ]: # function to conduct cross validation of a given model
        def run_model(model):
```

for train_index, test_index in kf.split(data):

y_pred=model.predict(X_test)

model = model

X_train, X_test = data[train_index], data[test_index]
y_train, y_test = target[train_index], target[test_index]

```
y_pred=np.argmax(y_pred,axis=1)
            Acc.append(accuracy_score(y_test, y_pred))
In [ ]: # 1) one-hidden Layer NN with 10 hidden nodes
      Acc = []
      run_model(model)
      print(np.mean(Acc), np.std(Acc))
     2/2 [=======] - 0s 2ms/step
     2/2 [======= ] - 0s 2ms/step
     2/2 [=======] - 0s 2ms/step
     2/2 [======= ] - 0s 3ms/step
     2/2 [======] - 0s 3ms/step
     2/2 [======== ] - 0s 2ms/step
     2/2 [======= ] - 0s 2ms/step
     2/2 [=======] - 0s 41ms/step
     2/2 [======= ] - 0s 4ms/step
     0.8675 0.03716517186829629
In [ ]: # 2) one-hidden Layer NN with 50 hidden nodes
      model50 = keras.Sequential([
         keras.layers.Flatten(input_shape=(64,64,1), name="Input"),
         keras.layers.Dense(50, activation='sigmoid', name='Hidden'),
         keras.layers.Dense(40, name="Output"),
      model50.compile(optimizer='adam',
                loss=tf.keras.losses.SparseCategoricalCrossentropy(from logits=True),
                metrics=['accuracy'])
      model50.fit(X_train, y_train, validation_split=0.2, epochs=300,verbose=0)
      Acc = []
      run model(model50)
      print(np.mean(Acc), np.std(Acc))
     2/2 [=======] - 0s 4ms/step
     2/2 [=======] - Os 4ms/step
     2/2 [======= ] - 0s 3ms/step
     2/2 [======= ] - 0s 43ms/step
     2/2 [=======] - 0s 12ms/step
     2/2 [======== ] - 0s 3ms/step
     2/2 [=======] - 0s 6ms/step
     2/2 [======== ] - 0s 2ms/step
     2/2 [======= ] - 0s 4ms/step
     2/2 [=======] - 0s 39ms/step
     2/2 [=======] - 0s 9ms/step
     0.919999999999999 0.04153311931459037
In []: # 3) one-hidden Layer NN with 500 hidden nodes
      model500 = keras.Sequential([
         keras.layers.Flatten(input_shape=(64,64,1), name="Input"),
         keras.layers.Dense(500, activation='sigmoid', name='Hidden'),
         keras.layers.Dense(40, name="Output"),
      model500.compile(optimizer='adam',
                loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),
                 metrics=['accuracy'])
```

```
model500.fit(X_train, y_train, validation_split=0.2, epochs=300,verbose=0)
      Acc = []
      run model(model500)
      print(np.mean(Acc), np.std(Acc))
     2/2 [======] - 0s 5ms/step
     2/2 [======= ] - 0s 5ms/step
     2/2 [======= ] - 0s 5ms/step
     2/2 [=======] - 0s 5ms/step
     2/2 [======= ] - 0s 4ms/step
     2/2 [======= ] - 0s 5ms/step
     2/2 [======= ] - 0s 3ms/step
     2/2 [======= ] - 0s 5ms/step
     2/2 [======= ] - 0s 3ms/step
     2/2 [======= ] - 0s 5ms/step
     2/2 [======= ] - 0s 4ms/step
     0.9475000000000000 0.02610076627227636
In [ ]: # 4) two-hidden layer NN with 50 hidden nodes (1st) and 10 hidden nodes (2nd)
      model2hl = keras.Sequential([
         keras.layers.Flatten(input_shape=(64,64,1), name="Input"),
         keras.layers.Dense(50, activation='sigmoid', name='HiddenLayer1'),
         keras.layers.Dense(10, activation='sigmoid', name='HiddenLayer2'),
         keras.layers.Dense(40, name="Output"),
      ])
      model2hl.compile(optimizer='adam',
                loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),
                metrics=['accuracy'])
      model2hl.fit(X_train, y_train, validation_split=0.2, epochs=300,verbose=0)
      Acc = []
      run model(model2hl)
      print(np.mean(Acc), np.std(Acc))
     2/2 [=======] - 0s 3ms/step
     2/2 [=======] - 0s 3ms/step
     2/2 [======= ] - 0s 3ms/step
     2/2 [=======] - Os 3ms/step
     2/2 [======= ] - 0s 3ms/step
     2/2 [======== ] - 0s 4ms/step
     2/2 [======= ] - 0s 3ms/step
     2/2 [======= ] - 0s 5ms/step
     2/2 [======= ] - 0s 3ms/step
     2/2 [======= ] - 0s 3ms/step
     2/2 [======== ] - 0s 4ms/step
     0.31 0.04769696007084728
```

Report and compare the cross-validation accuracy of the four neural networks

NN #1 accuracy: 0.8675
NN #2 accuracy: 0.9199
NN #3 accuracy: 0.9475
NN #4 accuracy: 0.31

From the results of these cross-validation, NN #3 (one-hidden layer with 500 hidden nodes)

seems to be the most accurate in classifying the faces to the correct classes. It seems to be a general trend to have slightly improved accuracy as the number of hidden nodes increased within a one-hidden layer NN. NN #4's results suggest that the two-hidden layer NN has a risk of overfitting, although that may be able to remedied with other solutions or techniques.