

Week-7: consider the network and dataset from week 6 visualize the hidden layers features. Compute the confusion matrix.

Code:

```
import numpy as np
import tensorflow as tf
from tensorflow.keras import layers, models
from sklearn.model_selection import KFold
from tensorflow.keras.datasets import mnist
from sklearn.metrics import accuracy_score, confusion_matrix
import matplotlib.pyplot as plt
```

Load the MNIST dataset

```
(x_train, y_train), (x_test, y_test) = mnist.load_data()
```

Normalize pixel values to be between 0 and 1

```
x_train, x_test = x_train / 255.0, x_test / 255.0
```

Expand dimensions to add a channel dimension

```
x_train = np.expand_dims(x_train, axis=-1)
```

```
x_test = np.expand_dims(x_test, axis=-1)
```

Define the CNN model with 4 layers

```
def create_model():
```

```
    model = models.Sequential()
```

```
model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28,
28, 1)))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.Flatten())
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(10, activation='softmax'))
model.compile(optimizer='sgd',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])
return model
```

Print model summary

```
model = create_model()
model.summary()
```

Perform K-fold cross-validation

```
k_folds = 10
kf = KFold(n_splits=k_folds, shuffle=True, random_state=42)
```

```
accuracies = []
```

```
conf_matrices = []
```

```
for train_index, val_index in kf.split(x_train):
```

```
    x_train_fold, x_val_fold = x_train[train_index], x_train[val_index]
```

```
    y_train_fold, y_val_fold = y_train[train_index], y_train[val_index]
```

```
model = create_model()
```

Train the model

```
model.fit(x_train_fold, y_train_fold, epochs=5, batch_size=64, verbose=0)
```

Evaluate on the validation set

```
val_predictions = np.argmax(model.predict(x_val_fold), axis=-1)
```

```
accuracy = accuracy_score(y_val_fold, val_predictions)
```

```
accuracies.append(accuracy)
```

Compute confusion matrix

```
conf_matrix = confusion_matrix(y_val_fold, val_predictions)
```

```
conf_matrices.append(conf_matrix)
```

Print the accuracy over the K fold

```
print('Accuracy is:', accuracy)
```

Print the confusion matrix

```
for fold, conf_matrix in enumerate(conf_matrices):
```

```
    print(f'Confusion Matrix for Fold {fold + 1}:')
```

```
    print(conf_matrix)
```

Print the average accuracy over the K folds

```
print(f'Average Accuracy: {np.mean(accuracies)}')
```

Visualize hidden layer features

```
layer_outputs = [layer.output for layer in model.layers[:4]] # Extracting  
the outputs of the top 4 layers
```

```
activation_model = tf.keras.models.Model(inputs=model.input,  
outputs=layer_outputs) # Creates a model that will return these outputs,  
given the model input
```

```
sample_img = x_train[0].reshape(1, 28, 28, 1) # Reshape an input  
sample
```

```
activations = activation_model.predict(sample_img) # Predictions of  
the activations
```

Plot the activations of each layer

```
for layer_activation, layer_name in zip(activations, ['conv1', 'maxpool1',  
'conv2', 'flatten']):
```

```
    plt.figure()
```

```
    plt.matshow(layer_activation[0, :, :, 0], cmap='viridis')
```

```
    plt.title(layer_name)
```

```
    plt.colorbar()
```

```
plt.show()
```

Output:

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 26, 26, 32)	320
max_pooling2d (MaxPooling2D)	(None, 13, 13, 32)	0
conv2d_1 (Conv2D)	(None, 11, 11, 64)	18496
flatten (Flatten)	(None, 7744)	0
dense (Dense)	(None, 64)	495680
dense_1 (Dense)	(None, 10)	650

Total params: 515146 (1.97 MB)
Trainable params: 515146 (1.97 MB)
Non-trainable params: 0 (0.00 Byte)

188/188 [=====] - 4s 17ms/step
Accuracy is: 0.9588333333333333
188/188 [=====] - 4s 16ms/step
Accuracy is: 0.9598333333333333
188/188 [=====] - 4s 16ms/step
Accuracy is: 0.9616666666666667
188/188 [=====] - 3s 14ms/step
Accuracy is: 0.9663333333333334
188/188 [=====] - 3s 13ms/step
Accuracy is: 0.9595
188/188 [=====] - 4s 16ms/step
Accuracy is: 0.972
188/188 [=====] - 3s 15ms/step
Accuracy is: 0.96
188/188 [=====] - 3s 13ms/step
Accuracy is: 0.9571666666666667
188/188 [=====] - 3s 13ms/step
Accuracy is: 0.9611666666666666
188/188 [=====] - 5s 19ms/step
Accuracy is: 0.9651666666666666

```

Confusion Matrix for Fold 1:
[[604  0  3  2  0  2  3  1  8  1]
 [  0 645  2  0  0  0  0  0  6  1]
 [  0  2 559  2  2  0  0  2  4  1]
 [  2  1  7 557  0 10  0  1 11  0]
 [  1  3  3  2 562  0  1  1  4  3]
 [  2  2  2  9  3 520  2  0 11  0]
 [  2  0  1  0  2  4 568  0  3  0]
 [  1  6 12  3  3  3  0 589 10  6]
 [  1  4  1  3  1  1  4  0 570  0]
 [  1  2  1  9 19  3  0  3 15 579]]

```

```

Confusion Matrix for Fold 2:
[[545  0  1  0  2  1  0  0  0  2]
 [  0 659  4  2  1  0  0  0  1  1]
 [  0  5 584  2  1  0  0  6  3  1]
 [  1  1  5 602  0  9  0  3  4  5]
 [  0  0  0  0 583  0  1  0  0 12]
 [  5  1  1 12  1 524  3  0  1  5]
 [  0  3  2  0  5  8 576  0  3  0]
 [  1  2  4  4  6  1  0 634  0 14]
 [  2  6  4 23  4  8  2  3 512 11]
 [  2  2  0  1  8  3  0  6  0 540]]

```

```

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-
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```

```

Confusion Matrix for Fold 10:
[[578  0  1  0  1  2  5  1  1  0]
 [  0 663  3  0  0  1  1  2  4  0]
 [  1  1 573  0  5  0  0  4  4  1]
 [  0  3  9 539  0 15  1  4  7  0]
 [  1  0  1  0 566  0  4  2  0 10]
 [  0  1  0  3  2 518  3  0  1  2]
 [  7  4  1  1  0  7 586  0  3  0]
 [  0  1  3  0  1  1  1 609  1  6]
 [  1  3  4  1  1  7  4  1 599  4]
 [  0  1  2  4  5  9  0 13  5 560]]

```

```

Average Accuracy: 0.9621666666666666
1/1 [=====] - 1s 772ms/step

```



