

Assignment 6.1

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class: DSC650

Assignment 5.1

Using section 5.1 in Deep Learning with Python as a guide (listing 5.3 in particular), create a ConvNet model that classifies images in the MNIST digit dataset. Save the model, predictions, metrics, and validation plots in the dsc650/assignments/assignment06/results directory. If you are using JupyterHub, you can include those plots in your Jupyter notebook.

```
4 # Import Packages
from keras.datasets import mnist
from tensorflow.keras.utils import to_categorical
from keras import layers
from keras import models
import tensorflow as tf

7 # Instantiating a small covnet
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.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))

8 # Creating Summary of Model
model.summary()
```

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
max_pooling2d_1 (MaxPooling2D)	(None, 5, 5, 64)	0
conv2d_2 (Conv2D)	(None, 3, 3, 64)	36928
=====		
Total params: 55,744		
Trainable params: 55,744		

Non-trainable params: 0

9 # 5.2 Adding a classifier on top of the covnet

```
model.add(layers.Flatten())
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(10, activation='softmax'))
```

10 # Creating Summary of Model using a final layer with 10 outputs and a softmax activation

```
model.summary()
```

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
max_pooling2d_1 (MaxPooling2D)	(None, 5, 5, 64)	0
conv2d_2 (Conv2D)	(None, 3, 3, 64)	36928
flatten (Flatten)	(None, 576)	0
dense (Dense)	(None, 64)	36928
dense_1 (Dense)	(None, 10)	650
=====		
Total params: 93,322		
Trainable params: 93,322		
Non-trainable params: 0		

11 # 5.3 Training the covnet on MNIST images

```
(train_images, train_labels), (test_images, test_labels) = mnist.load_data()
```

```
train_images = train_images.reshape((60000, 28, 28, 1))
train_images = train_images.astype('float32') / 255
test_images = test_images.reshape((10000, 28, 28, 1))
test_images = test_images.astype('float32') / 255
train_labels = to_categorical(train_labels)
test_labels = to_categorical(test_labels)
model.compile(optimizer='rmsprop',
              loss='categorical_crossentropy',
              metrics=['accuracy'])
model.fit(train_images, train_labels, epochs=5, batch_size=64)
```

Epoch 1/5

938/938 [=====] - 15s 15ms/step - loss: 0.1681 - accuracy: 0.9471????????????????

Epoch 2/5

938/938 [=====] - 14s 15ms/step - loss: 0.0448 - accuracy: 0.9863????????????????

Epoch 3/5

938/938 [=====] - 14s 15ms/step - loss: 0.0307 - accuracy: 0.9905????????????????

Epoch 4/5

938/938 [=====] - 14s 15ms/step - loss: 0.0235 - accuracy: 0.9927????????????????

Epoch 5/5
938/938 [=====] - 14s 15ms/step - loss: 0.0190 - accuracy: 0.9939

11 <keras.callbacks.History at 0x174b0d27748>

12 # Evaluate the model on test data
test_loss, test_acc = model.evaluate(test_images, test_labels)
test_acc

313/313 [=====] - 1s 2ms/step - loss: 0.0318 - accuracy: 0.9908

12 0.9908000230789185

13 model_no_max_pool = models.Sequential()
model_no_max_pool.add(layers.Conv2D(32, (3, 3), activation='relu',
input_shape=(28, 28, 1)))
model_no_max_pool.add(layers.Conv2D(64, (3, 3), activation='relu'))
model_no_max_pool.add(layers.Conv2D(64, (3, 3), activation='relu'))
model_no_max_pool.summary()

Model: "sequential_1"

Layer (type)	Output Shape	Param #
conv2d_3 (Conv2D)	(None, 26, 26, 32)	320
conv2d_4 (Conv2D)	(None, 24, 24, 64)	18496
conv2d_5 (Conv2D)	(None, 22, 22, 64)	36928
Total params: 55,744		
Trainable params: 55,744		
Non-trainable params: 0		

