HW10

Date: 2021. 12. 06

Student ID: 21600635

Name: Bomoon JUNG

1. Homework 10

1) show one example of Hand-written Number

```
import tensorflow as tf
from tensorflow import keras
import numpy as np
import matplotlib.pyplot as plt

[x_train, y_train], (x_test, y_test) = keras.datasets.mnist.load_data()
# x_train.shape
# type(x_train[0,0,0])

[5] # my input image
x_train = x_train.astype('float32') / 255.
n = 1
plt.imshow(x_train[n], cmap = 'Greys', interpolation = 'nearest')
plt.show()
```

2) show training result

```
[6] x_train = x_train.reshape( x_train.shape[0], 28, 28, 1 )
    # x_train.shape
    x_test = x_test.reshape( x_test.shape[0], 28, 28, 1 )

input_shape = ( 28, 28, 1)

# input_shape

[7] y_train[0:10]
    num_classes = 10
    y_train = keras.utils.to_categorical( y_train, num_classes )
    y_test = keras.utils.to_categorical(y_test, num_classes )
    y_train[0:10]

array([[0., 0., 0., 0., 0., 0., 1., 0., 0., 0., 0.],
        [1., 0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 1., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 1., 0., 0., 0., 0., 0.],
        [0., 1., 0., 0., 0., 0., 0., 0., 0.],
        [0., 1., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.],
        [0., 0., 0., 0., 0., 0., 0., 0.]],
        [0., 0., 0., 0., 0., 0., 0., 0.]],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.]],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.]],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.]],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.]],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.]],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.]],
        [0., 0., 0., 0., 0., 0., 0., 0., 0.]],
        [0., 0., 0., 0., 0., 0., 0., 0.]]
```

3) modeling

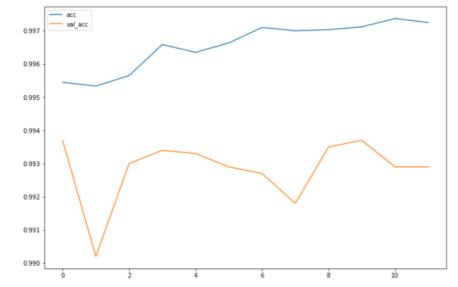
```
[8] import sys
      import tensorflow as tf
import keras
       from keras.models import Sequential from keras.layers import Dense, Dropout, Flatten
       from keras.layers.convolutional import Conv2D, MaxPooling2D
       import numpy as np
      np.random.seed(7)
[12] model = Sequential()
      model.add( Conv2D( 32, kernel_size = (5, 5), strides = (1, 1), padding = 'same', model.add( MaxPooling2D( pool_size = (2, 2), strides = (2, 2) ) ) model.add( MaxPooling2D( pool_size = (2, 2) ) )
      model.add( Dropout(0.25) )
      model.add( Flatten() )
      model.add( Dense( 1000, activation = 'relu' ) )
       model.add( Dropout(0.5) )
      model.add( Dense( num_classes, activation = 'softmax' ) )
       model = Sequential()
      model.add( Conv2D( 32, kernel_size = ( 5, 5 ), strides = (1, 1), padding = 'same', activation = 'relu', input_shape = input_shape ) )
model.add( MaxPooling2D( pool_size = ( 2, 2 ), strides = (2, 2) ) )
model.add( Conv2D( 64, (2, 2), activation = 'relu', padding = 'same' ) )
model.add( MaxPooling2D( pool_size = (2, 2) ) )
      model.add( Dropout(0.25) )
      model.add( Flatten() )
       model.add( Dense( 1000, activation = 'relu' ) )
      model.add( Dropout(0.5) )
      model.add( Dense( num_classes, activation = 'softmax' ) )
```

4) learning

```
batch_size = 128
   epochs = 12
     odel.compile( loss = 'categorical_crossentropy', optimizer = 'adam', metrics = ['accuracy'])
   hist = model.fit( x_train, y_train, batch_size = batch_size, epochs = epochs, verbose = 1, validation_data = (x_test, y_test) )
   Epoch 1/12
   469/469 [=
                     =========] - 92s 194ms/step - loss: 0.0136 - accuracy: 0.9955 - val_loss: 5.2190 - val_accuracy: 0.9937
   Epoch 2/12
                          469/469 [==
   Epoch 3/12
469/469 [==
Epoch 4/12
                            =======] - 90s 192ms/step - loss: 0.0123 - accuracy: 0.9956 - val_loss: 5.3364 - val_accuracy: 0.9930
   469/469 [=
Epoch 5/12
                         ========] - 90s 192ms/step - loss: 0.0107 - accuracy: 0.9966 - val_loss: 5.7794 - val_accuracy: 0.9934
                         ======== | - 90s 193ms/step - loss: 0.0107 - accuracy: 0.9963 - val loss: 6.3672 - val accuracy: 0.9933
   469/469 [==
   Epoch 6/12
469/469 [=
                            =======] - 90s 193ms/step - loss: 0.0104 - accuracy: 0.9966 - val_loss: 5.7960 - val_accuracy: 0.9929
   Epoch 7/12
   469/469 [
                          ========] - 91s 194ms/step - loss: 0.0087 - accuracy: 0.9971 - val_loss: 7.2519 - val_accuracy: 0.9927
   Epoch 8/12
   469/469 [=
Epoch 9/12
                        469/469 [=
   Epoch 10/12
   469/469 [==
                         ========] - 90s 192ms/step - loss: 0.0083 - accuracy: 0.9971 - val_loss: 6.3400 - val_accuracy: 0.9937
   Epoch 11/12
   469/469 [==
Epoch 12/12
                        ========] - 90s 192ms/step - loss: 0.0079 - accuracy: 0.9974 - val_loss: 8.4755 - val_accuracy: 0.9929
                        ========= ] - 90s 192ms/step - loss: 0.0080 - accuracy: 0.9973 - val loss: 7.5949 - val accuracy: 0.9929
   469/469 [==
```

5) graph of accuracy and validate accuracy

```
[18] plt.figure( figsize = (12, 8) )
    # plt.plot( hist.history['loss'] )
    # plt.plot( hist.history['val_loss'] )
    plt.plot( hist.history['accuracy'] )
    plt.plot( hist.history['val_accuracy'] )
    # plt.legend( ['loss', 'val_loss', 'acc', 'val_acc'] )
    plt.legend( ['acc', 'val_acc'] )
    plt.show()
```



6) results of error case

