

20220614_Wirtschaftsinformatik_FAT2

June 14, 2022

```
[1]: !pip install keras
```

Requirement already satisfied: keras in
/Users/h4/opt/anaconda3/lib/python3.8/site-packages (2.6.0)

```
[3]: import keras  
from keras.datasets import mnist  
mnist.load_data()
```

```
[3]: ((array([[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, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0]], dtype=uint8),
array([5, 0, 4, ..., 5, 6, 8], dtype=uint8)),
(array([[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],
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...,
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0]],

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[[0, 0, 0, ..., 0, 0, 0],
 [0, 0, 0, ..., 0, 0, 0],
 [0, 0, 0, ..., 0, 0, 0],
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 [0, 0, 0, ..., 0, 0, 0],
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 [0, 0, 0, ..., 0, 0, 0]],

[[0, 0, 0, ..., 0, 0, 0],
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 ...,
 [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]]], dtype=uint8),
array([7, 2, 1, ..., 4, 5, 6], dtype=uint8)))

```

```
[4]: # imports von packages
```

```

import numpy as np
import matplotlib
import matplotlib.pyplot as plt

from keras.models import Sequential, load_model
from keras.layers.core import Dense, Activation
from keras.utils import np_utils

```

```
[5]: # SPLIT Funktion
```

```
(X_train, y_train), (X_test, y_test)=mnist.load_data()
```

```
[8]: # Umwandlung der Daten
```

```

X_train =X_train.reshape(60000, 784) # die 60000 Bilder 28x28 vom Train werden
→umgewandelt in 60000 Bilder 1x784
X_test=X_test.reshape(10000, 784) # bei Test gibt es 10000 28x28 Bilder

X_train=X_train.astype('float32')
X_test=X_test.astype('float32')

```

```
# Normalisierung (Graustufen sind zw 0 und 255). Also wir wollen die Daten zw 0 und 1 haben. Deshalb teilen wir durch 255.
```

```
X_train /=255  
X_test /=255
```

```
[9]: # nicht Prüfungsrelevant
```

```
# Die zehn Kategorien werden umgewandelt in Zahlenkategorien.  
# Dies macht man mit "One-hot Encoding"
```

```
n_classes = 10
```

```
Y_train = np_utils.to_categorical(y_train, n_classes)  
Y_test = np_utils.to_categorical(y_test, n_classes)
```

```
[10]: # Deep Learning
```

```
model = Sequential()  
  
model.add(Dense(512, input_shape=(784,)))  
model.add(Activation('relu'))  
model.add(Dense(512))  
model.add(Activation('relu'))  
model.add(Dense(10))  
model.add(Activation('softmax'))
```

```
[11]: # nicht prüfungsrelevant  
# Compilation vom Model
```

```
model.compile(loss='categorical_crossentropy', metrics=['accuracy'],  
optimizer='adam')
```

```
[12]: # training vom model
```

```
history = model.fit(X_train, Y_train, batch_size=128, epochs=20,  
validation_data=(X_test, Y_test))
```

Epoch 1/20

469/469 [=====] - 3s 5ms/step - loss: 0.8066 -
accuracy: 0.7606 - val_loss: 0.3843 - val_accuracy: 0.8913

Epoch 2/20

469/469 [=====] - 2s 5ms/step - loss: 0.3525 -
accuracy: 0.8969 - val_loss: 0.3100 - val_accuracy: 0.9109

Epoch 3/20

469/469 [=====] - 2s 5ms/step - loss: 0.2893 -

accuracy: 0.9165 - val_loss: 0.2645 - val_accuracy: 0.9215
 Epoch 4/20
 469/469 [=====] - 2s 5ms/step - loss: 0.2491 -
 accuracy: 0.9262 - val_loss: 0.2371 - val_accuracy: 0.9286
 Epoch 5/20
 469/469 [=====] - 2s 5ms/step - loss: 0.2128 -
 accuracy: 0.9375 - val_loss: 0.1931 - val_accuracy: 0.9424
 Epoch 6/20
 469/469 [=====] - 2s 5ms/step - loss: 0.1818 -
 accuracy: 0.9466 - val_loss: 0.1776 - val_accuracy: 0.9454
 Epoch 7/20
 469/469 [=====] - 3s 5ms/step - loss: 0.1581 -
 accuracy: 0.9530 - val_loss: 0.1536 - val_accuracy: 0.9529
 Epoch 8/20
 469/469 [=====] - 2s 5ms/step - loss: 0.1371 -
 accuracy: 0.9594 - val_loss: 0.1355 - val_accuracy: 0.9584
 Epoch 9/20
 469/469 [=====] - 2s 5ms/step - loss: 0.1201 -
 accuracy: 0.9639 - val_loss: 0.1249 - val_accuracy: 0.9602
 Epoch 10/20
 469/469 [=====] - 2s 5ms/step - loss: 0.1062 -
 accuracy: 0.9685 - val_loss: 0.1167 - val_accuracy: 0.9648
 Epoch 11/20
 469/469 [=====] - 2s 5ms/step - loss: 0.0943 -
 accuracy: 0.9718 - val_loss: 0.1064 - val_accuracy: 0.9664
 Epoch 12/20
 469/469 [=====] - 2s 5ms/step - loss: 0.0845 -
 accuracy: 0.9744 - val_loss: 0.0978 - val_accuracy: 0.9684
 Epoch 13/20
 469/469 [=====] - 3s 5ms/step - loss: 0.0768 -
 accuracy: 0.9772 - val_loss: 0.0928 - val_accuracy: 0.9707
 Epoch 14/20
 469/469 [=====] - 2s 5ms/step - loss: 0.0683 -
 accuracy: 0.9794 - val_loss: 0.0908 - val_accuracy: 0.9722
 Epoch 15/20
 469/469 [=====] - 2s 5ms/step - loss: 0.0624 -
 accuracy: 0.9810 - val_loss: 0.0857 - val_accuracy: 0.9738
 Epoch 16/20
 469/469 [=====] - 2s 5ms/step - loss: 0.0558 -
 accuracy: 0.9835 - val_loss: 0.0847 - val_accuracy: 0.9743
 Epoch 17/20
 469/469 [=====] - 3s 5ms/step - loss: 0.0512 -
 accuracy: 0.9847 - val_loss: 0.0769 - val_accuracy: 0.9760
 Epoch 18/20
 469/469 [=====] - 2s 5ms/step - loss: 0.0453 -
 accuracy: 0.9865 - val_loss: 0.0814 - val_accuracy: 0.9742
 Epoch 19/20
 469/469 [=====] - 2s 5ms/step - loss: 0.0426 -

```
accuracy: 0.9871 - val_loss: 0.0812 - val_accuracy: 0.9745  
Epoch 20/20  
469/469 [=====] - 2s 5ms/step - loss: 0.0373 -  
accuracy: 0.9888 - val_loss: 0.0716 - val_accuracy: 0.9780
```

[]: