

digitbaseline

November 4, 2022

```
[1]: # Baseline MLP for MNIST dataset
      #import tensorflow as tf
      from keras.datasets import mnist
      from keras.models import Sequential
      from keras.layers import Dense
      from keras.utils import np_utils

      from sklearn.metrics import confusion_matrix
      from sklearn.metrics import classification_report

[2]: # load data
      (X_train, y_train), (X_test, y_test) = mnist.load_data()

[ ]: img = X_test[0]
      y_test[0]

[ ]: y_test.shape

[ ]: %matplotlib inline
      import matplotlib.pyplot as plt
      imgplot = plt.imshow(img)

[3]: # flatten 28*28 images to a 784 vector for each image
      num_pixels = X_train.shape[1] * X_train.shape[2]
      X_train = X_train.reshape((X_train.shape[0], num_pixels)).astype('float32')
      X_test = X_test.reshape((X_test.shape[0], num_pixels)).astype('float32')

[ ]: X_train.shape

[ ]: y_train.shape

[4]: # normalize inputs from 0-255 to 0-1
      X_train = X_train / 255
      X_test = X_test / 255

[ ]: y_train[1]
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[5]: # one hot encode outputs
y_test_org = y_test
y_train = np_utils.to_categorical(y_train)
y_test = np_utils.to_categorical(y_test)
num_classes = y_test.shape[1]

[ ]: y_test.shape

[ ]: img = X_train[30].reshape(28,28)

[ ]: y_train[30]

[ ]: %matplotlib inline
import matplotlib.pyplot as plt
imgplot = plt.imshow(img)

[6]: # define baseline model
def baseline_model():
    # create model
    model = Sequential()
    model.add(Dense(num_pixels, input_dim=num_pixels,activation='relu'))
    model.add(Dense(num_classes, activation='softmax'))
    # Compile model
    model.compile(loss='categorical_crossentropy', optimizer='adam',
↳metrics=['accuracy'])
    return model

[7]: # build the model
model = baseline_model()
# Fit the model
model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=10,
↳batch_size=200, verbose=1)
# Final evaluation of the model
```

```
Epoch 1/10
300/300 [=====] - 31s 25ms/step - loss: 0.4834 -
accuracy: 0.8629 - val_loss: 0.1387 - val_accuracy: 0.9593
Epoch 2/10
300/300 [=====] - 8s 26ms/step - loss: 0.1177 -
accuracy: 0.9667 - val_loss: 0.0964 - val_accuracy: 0.9725 0s - loss: 0.1185 -
accuracy: 0. - ETA: 0s - loss: 0.118
Epoch 3/10
300/300 [=====] - 7s 24ms/step - loss: 0.0714 -
accuracy: 0.9790 - val_loss: 0.0830 - val_accuracy: 0.9762
Epoch 4/10
300/300 [=====] - 7s 24ms/step - loss: 0.0488 -
accuracy: 0.9865 - val_loss: 0.0718 - val_accuracy: 0.9776
Epoch 5/10
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300/300 [=====] - 7s 24ms/step - loss: 0.0350 -
accuracy: 0.9913 - val_loss: 0.0686 - val_accuracy: 0.9787
Epoch 6/10
300/300 [=====] - 7s 23ms/step - loss: 0.0254 -
accuracy: 0.9932 - val_loss: 0.0692 - val_accuracy: 0.9794
Epoch 7/10
300/300 [=====] - 7s 24ms/step - loss: 0.0197 -
accuracy: 0.9949 - val_loss: 0.0588 - val_accuracy: 0.9824
Epoch 8/10
300/300 [=====] - 7s 22ms/step - loss: 0.0133 -
accuracy: 0.9975 - val_loss: 0.0598 - val_accuracy: 0.9815
Epoch 9/10
300/300 [=====] - 7s 22ms/step - loss: 0.0106 -
accuracy: 0.9978 - val_loss: 0.0641 - val_accuracy: 0.9816
Epoch 10/10
300/300 [=====] - 6s 21ms/step - loss: 0.0086 -
accuracy: 0.9983 - val_loss: 0.0629 - val_accuracy: 0.9820

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[7]: <keras.callbacks.History at 0x1f023034430>
```

```
[8]: scores, accuracy = model.evaluate(X_test, y_test, verbose=0)
# print("Baseline Error: %.2f%%" % (100-scores[1]*100))
print("Baseline Accuracy: %.2f%%" % (accuracy*100))
```

Baseline Accuracy: 98.20%

```
[9]: y_predicted = model.predict_classes(X_test)
```

```

C:\Users\Think\anaconda3\envs\tensorflowEnv\lib\site-
packages\keras\engine\sequential.py:450: UserWarning: `model.predict_classes()`
is deprecated and will be removed after 2021-01-01. Please use instead:
`np.argmax(model.predict(x), axis=-1)`, if your model does multi-class
classification (e.g. if it uses a `softmax` last-layer activation).
` (model.predict(x) > 0.5).astype("int32")`, if your model does binary
classification (e.g. if it uses a `sigmoid` last-layer activation).
warnings.warn("`model.predict_classes()` is deprecated and

```

```
[10]: confuse_matrix = confusion_matrix(y_test_org, y_predicted)
print(confuse_matrix)
```

```

[[ 972   1   1   1   1   0   1   1   2   0]
 [   0 1126   3   1   0   0   1   1   3   0]
 [   2   2 1015   0   1   0   2   3   7   0]
 [   0   0   4  994   0   3   0   1   5   3]
 [   1   0   3   0  966   0   3   1   0   8]
 [   2   0   0   9   3  867   3   1   6   1]
 [   5   3   2   1   4   3  939   0   1   0]
 [   1   5   9   1   1   0   0 1006   2   3]
 [   2   0   4   2   6   0   0   4  953   3]

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[ 2  4  0  2 13  3  0  3  0 982]]
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```
[11]: result = classification_report(y_test_org, y_predicted)
      print(result)
```

	precision	recall	f1-score	support
0	0.98	0.99	0.99	980
1	0.99	0.99	0.99	1135
2	0.98	0.98	0.98	1032
3	0.98	0.98	0.98	1010
4	0.97	0.98	0.98	982
5	0.99	0.97	0.98	892
6	0.99	0.98	0.98	958
7	0.99	0.98	0.98	1028
8	0.97	0.98	0.98	974
9	0.98	0.97	0.98	1009
accuracy			0.98	10000
macro avg	0.98	0.98	0.98	10000
weighted avg	0.98	0.98	0.98	10000

```
[ ]: y_predicted[0]
```

```
[ ]: img = X_test[0].reshape(28,28)
      %matplotlib inline
      import matplotlib.pyplot as plt
      imgplot = plt.imshow(img)
```

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[ ]:
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