

## ▼ Import Data

```

1 import warnings
2 warnings.filterwarnings('ignore')

1 from keras.datasets import mnist

1 # the data, split between train and test sets
2 (x_train, y_train), (x_test, y_test) = mnist.load_data()

1 from keras import backend as K
2 K.image_data_format()

[ ]> 'channels_last'

1 # 'channels_last' data format
2 x_train = x_train.reshape(x_train.shape[0], 28, 28, 1)
3 x_test = x_test.reshape(x_test.shape[0], 28, 28, 1)

1 input_shape = (28, 28, 1)

1 x_train = x_train.astype('float32')
2 x_test = x_test.astype('float32')
3 x_train /= 255
4 x_test /= 255

1 # convert class vectors to binary class matrices
2 from keras.utils import to_categorical
3 y_train = to_categorical(y_train, 10)
4 y_test = to_categorical(y_test, 10)

```

## ▼ Create CNN Model

Saved successfully!

```

1 from keras.models import Sequential
2 from keras.layers import Dense, Dropout, Flatten, Conv2D, MaxPooling2D

1 model = Sequential()
2 model.add(Conv2D(32, kernel_size=(3, 3),
3                 activation='relu',
4                 input_shape=(28, 28, 1)))
5 model.add(Conv2D(64, (3, 3), activation='relu'))
6 model.add(MaxPooling2D(pool_size=(2, 2)))
7 model.add(Dropout(0.25))
8 model.add(Flatten())
9 model.add(Dense(128, activation='relu'))

```

```
10 model.add(Dropout(0.5))
11 model.add(Dense(10, activation='softmax'))
```

```
1 model.summary()
```

```
Model: "sequential_5"
```

Layer (type)	Output Shape	Param #
conv2d_9 (Conv2D)	(None, 26, 26, 32)	320
conv2d_10 (Conv2D)	(None, 24, 24, 64)	18496
max_pooling2d_5 (MaxPooling2D)	(None, 12, 12, 64)	0
dropout_9 (Dropout)	(None, 12, 12, 64)	0
flatten_5 (Flatten)	(None, 9216)	0
dense_9 (Dense)	(None, 128)	1179776
dropout_10 (Dropout)	(None, 128)	0
dense_10 (Dense)	(None, 10)	1290
Total params: 1,199,882		
Trainable params: 1,199,882		
Non-trainable params: 0		

## ▼ Compile CNN Model

```
1 model.compile(loss='categorical_crossentropy',
2               optimizer='Adadelta',
3               metrics=['accuracy'])
```

## ▼ Train CNN Model

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```
3         epochs=20,
4         verbose=1,
5         validation_split=0.2)
```

```
Model:
```

Train on 48000 samples, validate on 12000 samples

```
Epoch 1/20
48000/48000 [=====] - 8s 168us/step - loss: 0.3058 - acc: 0.
Epoch 2/20
48000/48000 [=====] - 7s 150us/step - loss: 0.0991 - acc: 0.
Epoch 3/20
48000/48000 [=====] - 7s 151us/step - loss: 0.0702 - acc: 0.
Epoch 4/20
48000/48000 [=====] - 7s 151us/step - loss: 0.0569 - acc: 0.
Epoch 5/20
48000/48000 [=====] - 7s 151us/step - loss: 0.0498 - acc: 0.
Epoch 6/20
48000/48000 [=====] - 7s 150us/step - loss: 0.0427 - acc: 0.
Epoch 7/20
48000/48000 [=====] - 7s 150us/step - loss: 0.0376 - acc: 0.
Epoch 8/20
48000/48000 [=====] - 7s 150us/step - loss: 0.0347 - acc: 0.
Epoch 9/20
48000/48000 [=====] - 7s 150us/step - loss: 0.0318 - acc: 0.
Epoch 10/20
48000/48000 [=====] - 7s 150us/step - loss: 0.0285 - acc: 0.
Epoch 11/20
48000/48000 [=====] - 7s 151us/step - loss: 0.0279 - acc: 0.
Epoch 12/20
48000/48000 [=====] - 7s 151us/step - loss: 0.0258 - acc: 0.
Epoch 13/20
48000/48000 [=====] - 7s 155us/step - loss: 0.0243 - acc: 0.
Epoch 14/20
48000/48000 [=====] - 7s 152us/step - loss: 0.0223 - acc: 0.
Epoch 15/20
48000/48000 [=====] - 7s 151us/step - loss: 0.0231 - acc: 0.
Epoch 16/20
48000/48000 [=====] - 7s 150us/step - loss: 0.0225 - acc: 0.
Epoch 17/20
48000/48000 [=====] - 7s 149us/step - loss: 0.0208 - acc: 0.
Epoch 18/20
48000/48000 [=====] - 7s 150us/step - loss: 0.0198 - acc: 0.
Epoch 19/20
48000/48000 [=====] - 7s 150us/step - loss: 0.0197 - acc: 0.
Epoch 20/20
48000/48000 [=====] - 7s 149us/step - loss: 0.0185 - acc: 0.
```

Export to ONNX Model

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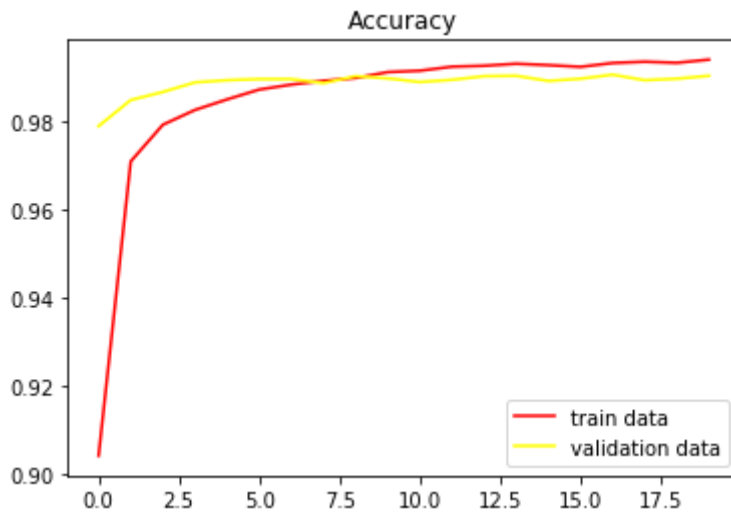


```
1 model.evaluate(x_test, y_test, verbose=0)
```

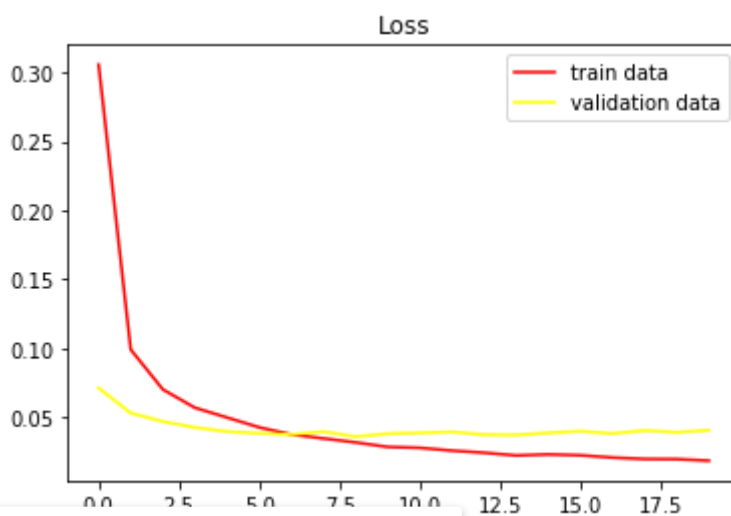
```
[0.025763151499420656, 0.9918]
```

```
1 from matplotlib import pyplot as plt
2 epochs = history.epoch
3 acc = history.history['acc']
4 val_acc = history.history['val_acc']
5 plt.plot(epochs, acc, color='red', label='train data')
6 plt.plot(epochs, val_acc, color='yellow', label='validation data')
7 plt.title('Accuracy')
```

```
8 plt.legend()
9 plt.show()
```



```
1 epochs = history.epoch
2 loss = history.history['loss']
3 val_loss = history.history['val_loss']
4 plt.plot(epochs, loss, color='red', label='train data')
5 plt.plot(epochs, val_loss, color='yellow', label='validation data')
6 plt.title('Loss')
7 plt.legend()
8 plt.show()
```



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## ▼ DoubledMNIST?

```
1 # For practise
```