

Flatten()은 1줄로 만들어서 Nueral Network에 1차원으로 데이터를 넣을 수 있게끔 하는데, 이러한 과정을 거치게 되면, 입력 데이터의 형상이 무너지는 부작용이 있다. 이는 단점으로써 작용할 수 있다.

```
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D, MaxPooling2D
from keras import backend as K
```

↳ Using TensorFlow backend.

```
K.image_data_format()
```

↳ 'channels\_last'

```
# the data, split between train and test sets
(x_train, y_train), (x_test, y_test) = mnist.load_data()
x_train = x_train / 255.0
x_test = x_test / 255.0
```

↳ Downloading data from <https://s3.amazonaws.com/img-datasets/mnist.npz>  
11493376/11490434 [=====] - 1s 0us/step

```
print(x_train.shape)
```

↳ (60000, 28, 28)

```
print(x_test.shape)
```

↳ (10000, 28, 28)

## ▼ Reshape for CNN

```
x_train = x_train.reshape(-1, 28, 28, 1)
print(x_train.shape)
```

↳ (60000, 28, 28, 1)

```
x_test = x_test.reshape(-1, 28, 28, 1)
print(x_test.shape)
```

↳ (10000, 28, 28, 1)

```
input_shape1 = (28, 28, 1)
```

## ▼ Hyper parameters

```
nb_epoch = 5
num_classes = 10
batch_size = 128
```

## ▼ Model definition

```
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3),
                 activation='relu',
                 input_shape=input_shape1))
model.add(Conv2D(64, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
#model.add(Dropout(0.25)) # 중간 레이어에서 조정하는 과정.
model.add(Flatten())
model.add(Dense(128, activation='relu'))
#model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))
```

☞ WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/framework  
Instructions for updating:  
Colocations handled automatically by placer.

```
model.summary()
```

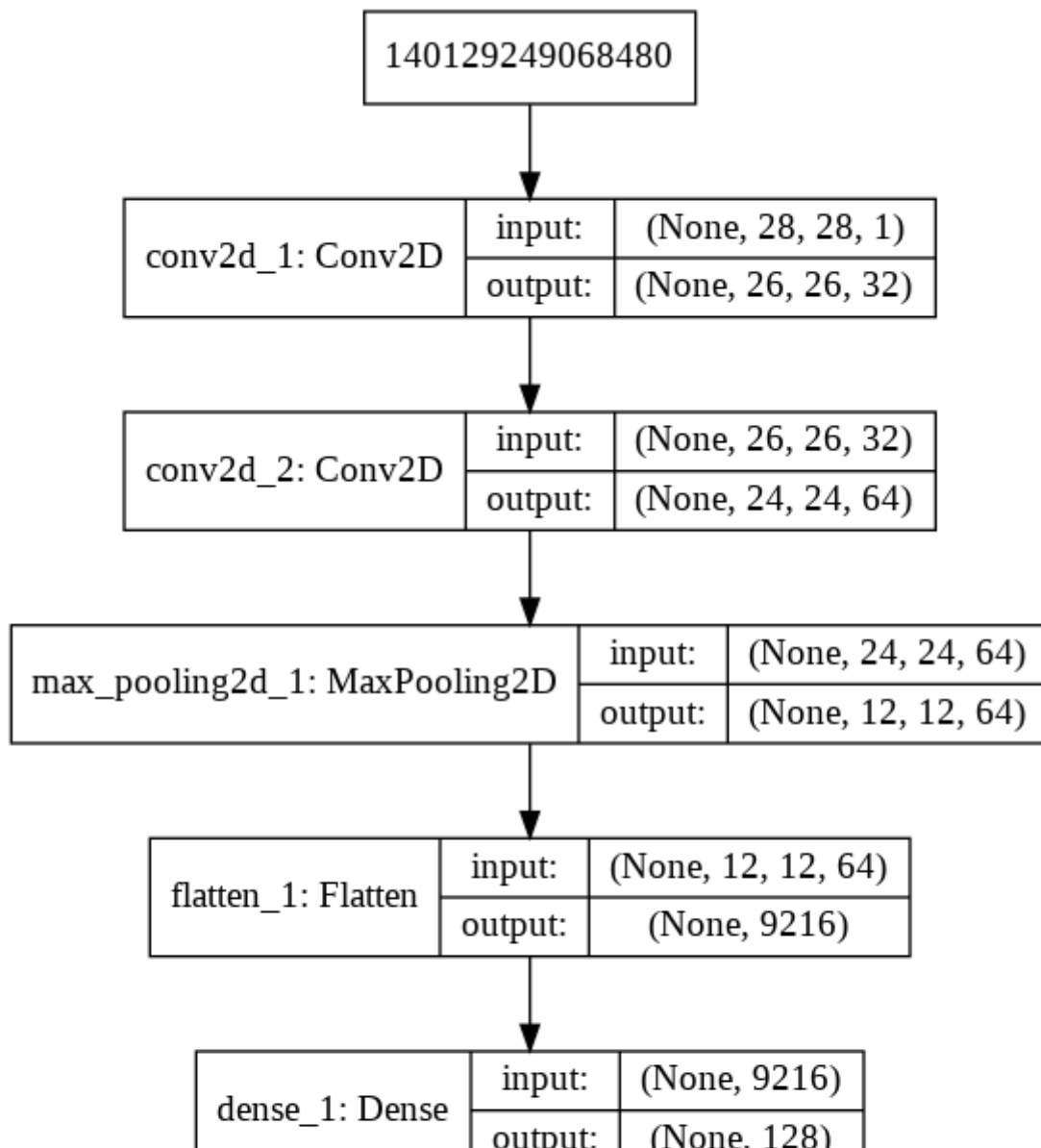
☞

Layer (type)	Output Shape	Param #
=====		
conv2d_1 (Conv2D)	(None, 26, 26, 32)	320
conv2d_2 (Conv2D)	(None, 24, 24, 64)	18496
max_pooling2d_1 (MaxPooling2	(None, 12, 12, 64)	0
flatten_1 (Flatten)	(None, 9216)	0
dense_1 (Dense)	(None, 128)	1179776
dense_2 (Dense)	(None, 10)	1290
=====		
Total params: 1,199,882		
Trainable params: 1,199,882		
Non-trainable params: 0		

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

```
from keras.utils import plot_model
plot_model(model, to_file='modelCNN.png', show_shapes=True)
from IPython.display import Image
Image("modelCNN.png")
```

☞



```
model.fit(x_train, y_train, epochs=nb_epoch)
```

```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/ops/math_
Instructions for updating:
Use tf.cast instead.
Epoch 1/5
60000/60000 [=====] - 178s 3ms/step - loss: 0.1112 - acc: 0.9655
Epoch 2/5
60000/60000 [=====] - 178s 3ms/step - loss: 0.0371 - acc: 0.9888
Epoch 3/5
60000/60000 [=====] - 178s 3ms/step - loss: 0.0224 - acc: 0.9928
Epoch 4/5
60000/60000 [=====] - 178s 3ms/step - loss: 0.0158 - acc: 0.9949
Epoch 5/5
60000/60000 [=====] - 177s 3ms/step - loss: 0.0127 - acc: 0.9954
<keras.callbacks.History at 0x7f7261968f28>

```

**Real World Challenge: Difference between training and testing set accuracy**

### ▼ Test accuracy

```
score = model.evaluate(x_test, y_test)
```

```
↳ 10000/10000 [=====] - 8s 754us/step
```

```
score
```

```
↳ [0.039154979102405923, 0.9896]
```

```
print('Test loss:{}'.format(score[0]))  
print('Test accuracy:{}'.format(score[1]))
```

```
↳ Test loss:0.039154979102405923  
Test accuracy:0.9896
```

### ▼ Training accuracy

```
score = model.evaluate(x_train, y_train)
```

```
↳ 60000/60000 [=====] - 44s 734us/step
```

```
print('Training loss:{}'.format(score[0]))  
print('Training accuracy:{}'.format(score[1]))
```

```
↳ Training loss:0.007022159157026423  
Training accuracy:0.9978
```

### ▼ Q.What is the difference between the training and test accuracy?

A.Model은 Training set을 통해서 학습을 진행하고 validation set을 통해서 자체 검증을 하는 절차가 없을 시 평가결과에선 굉장히 높은 적중률을 보여주지만 Test set을 통해서 평가를 진행하게 되면 학습시 보지 못했던 데이터이기 때문에 Training set보다 낮은 적중률을 보여주게 됩니다. 이 둘의 차이점은 기계가 보았던 데이터로 학습을 한건지 못 보던 데이터로 학습을 한 건지의 차이입니다.

