

Deep Learning
Deep Neural Network

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Deep Learning

CNN(2) Convolutional Neural Networks

- CNN(1): Background of Convolutional Neural Networks
 - CNN, Convolution Neural Network
 - Case study
- CNN(2): CNN examples

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5. CNN examples

• 5.1 CNN 구성 예

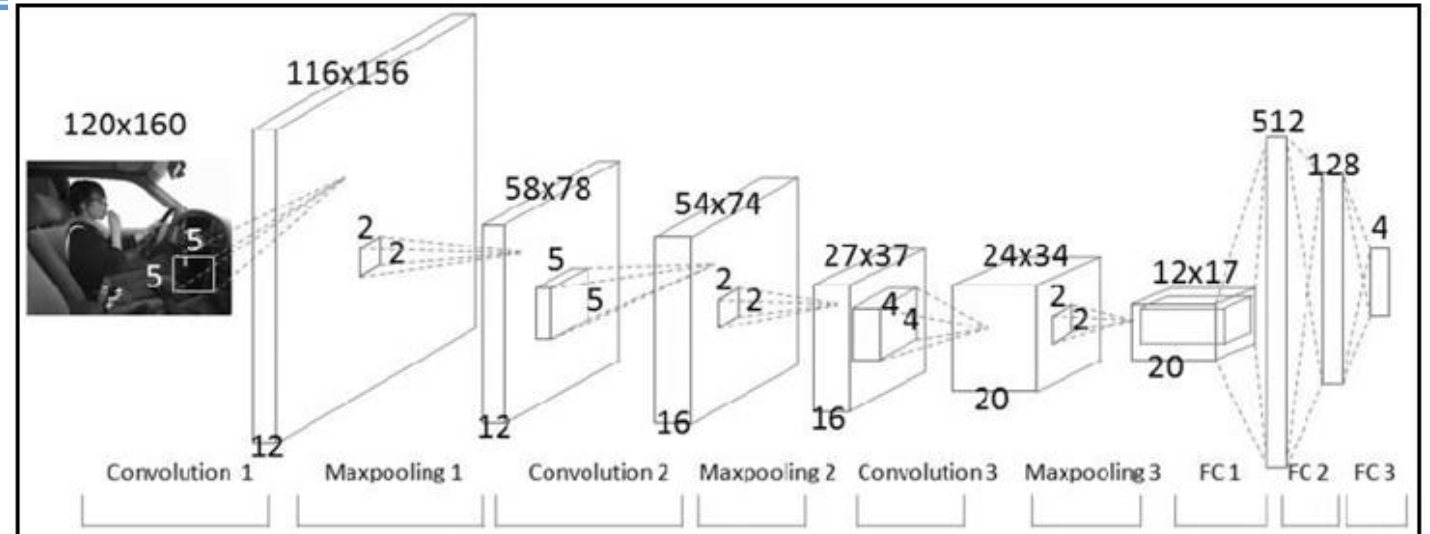
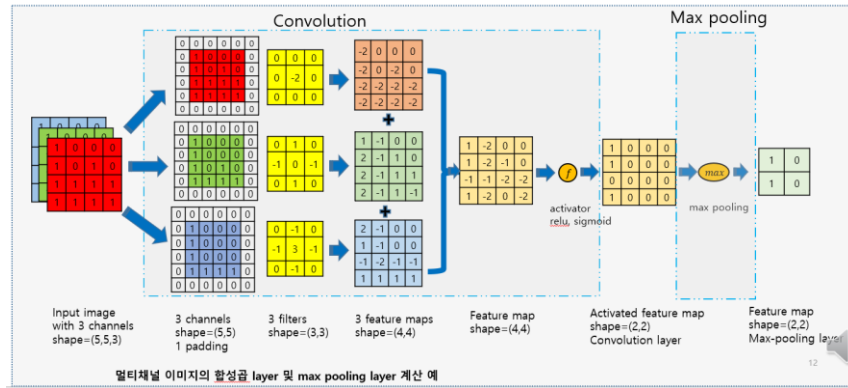


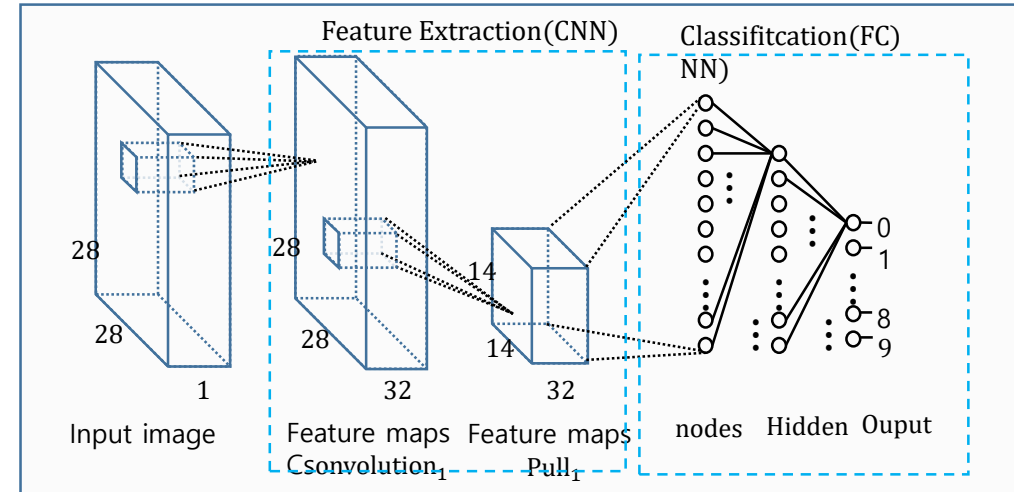
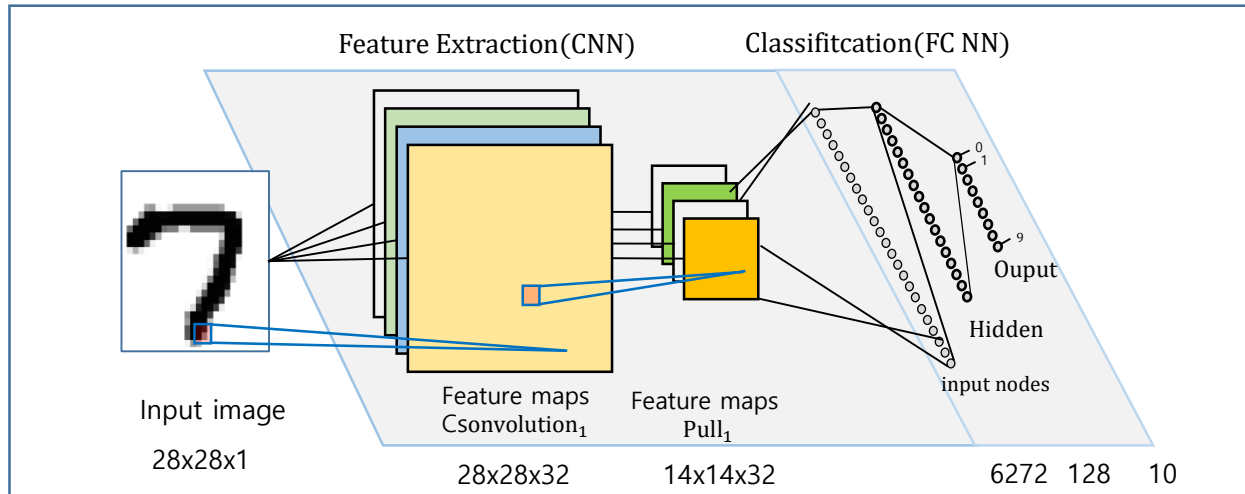
그림 8. 전형적인 CNN, 출처: https://www.researchgate.net/figure/Architecture-of-our-unsupervised-CNN-Network-contains-three-stages-each-of-which_283433254

```

model = Sequential()
model.add(Conv2D(12, kernel_size=(5, 5), activation='relu', input_shape=(120, 160, 1))) #116x156
model.add(MaxPooling2D(pool_size=(2, 2))) # 58x78
model.add(Conv2D(16, kernel_size=(5, 5), activation='relu')) #54x74
model.add(MaxPooling2D(pool_size=(2, 2))) # 27x37
model.add(Conv2D(20, kernel_size=(4, 4), activation='relu')) # 24x34
model.add(MaxPooling2D(pool_size=(2, 2))) # 12x17
model.add(Flatten()) #12x17=204
model.add(Dense(128, activation='relu'))
model.add(Dense(4, activation='softmax'))
    
```

5.2 Mnist digit classifier with 1 Convolution layer

- 1 Convolution NN



```

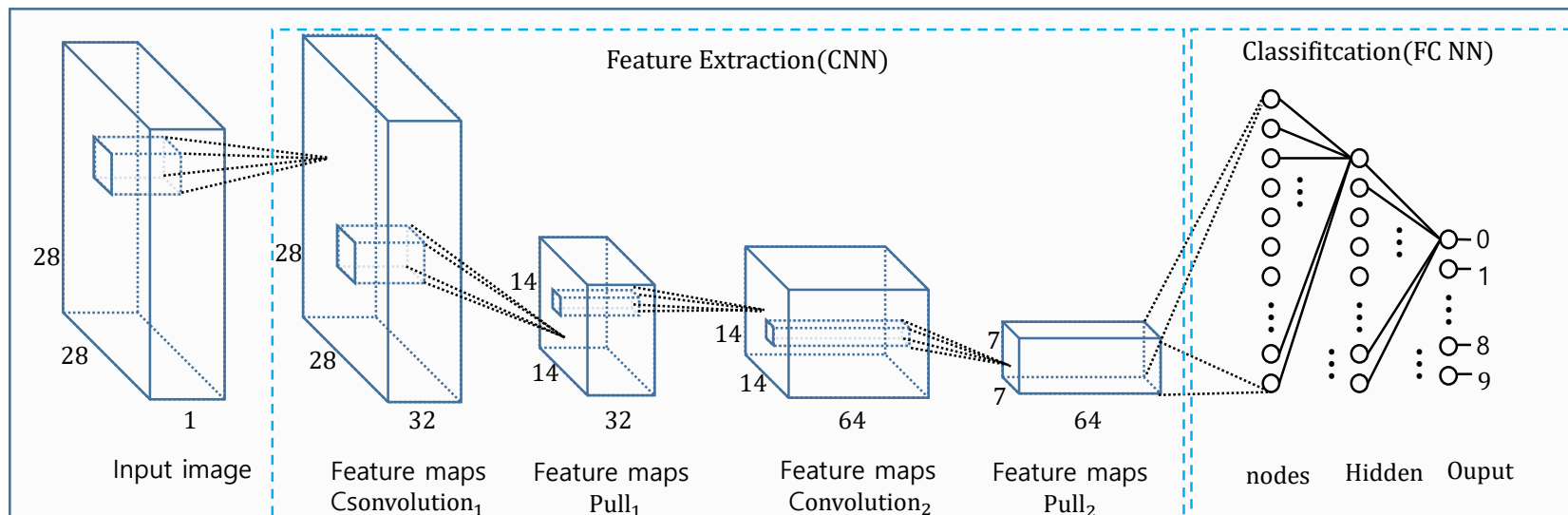
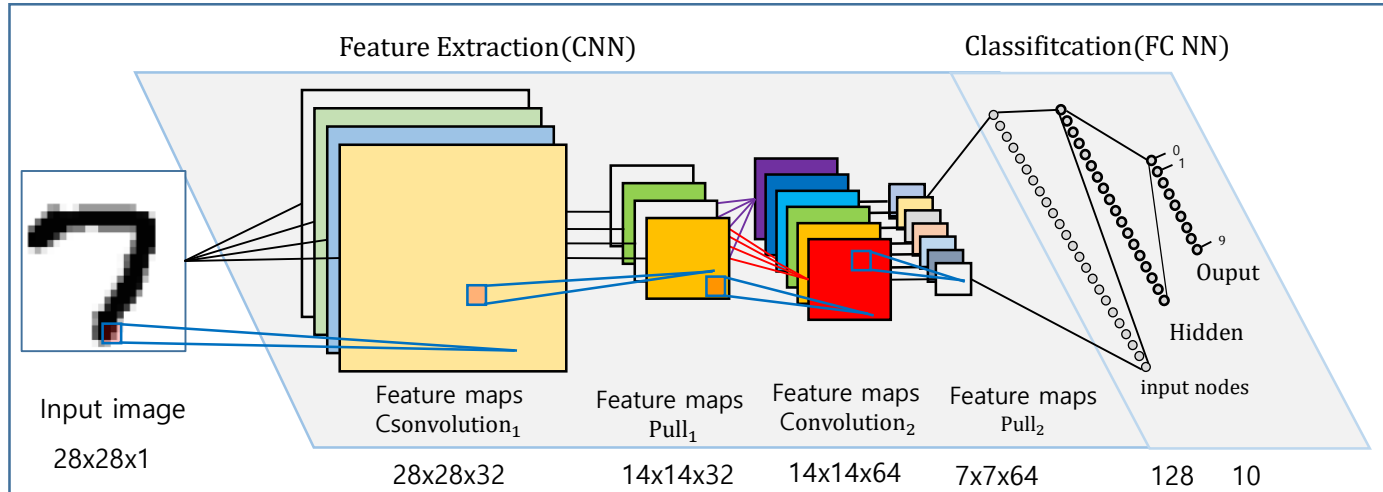
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3), activation='relu',input_shape=input_shape)) #26,26,32
model.add(MaxPooling2D(pool_size=(2, 2))) #14,14,32
model.add(Dropout(0.25))
model.add(Flatten()) #6272 = 14*14*32
model.add(Dense(128, activation='relu')) #128
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax')) #10
    
```

```

Epoch 10/12
60000/60000 [=====] - 4s 73us/step - loss: 0.0459 - acc: 0.9852 - val_loss: 0.0402 - val_acc: 0.9865
Epoch 11/12
60000/60000 [=====] - 5s 78us/step - loss: 0.0431 - acc: 0.9859 - val_loss: 0.0373 - val_acc: 0.9881
Epoch 12/12
60000/60000 [=====] - 5s 75us/step - loss: 0.0389 - acc: 0.9873 - val_loss: 0.0357 - val_acc: 0.9880
Test loss: 0.035724134841622436
Test accuracy: 0.988
    
```

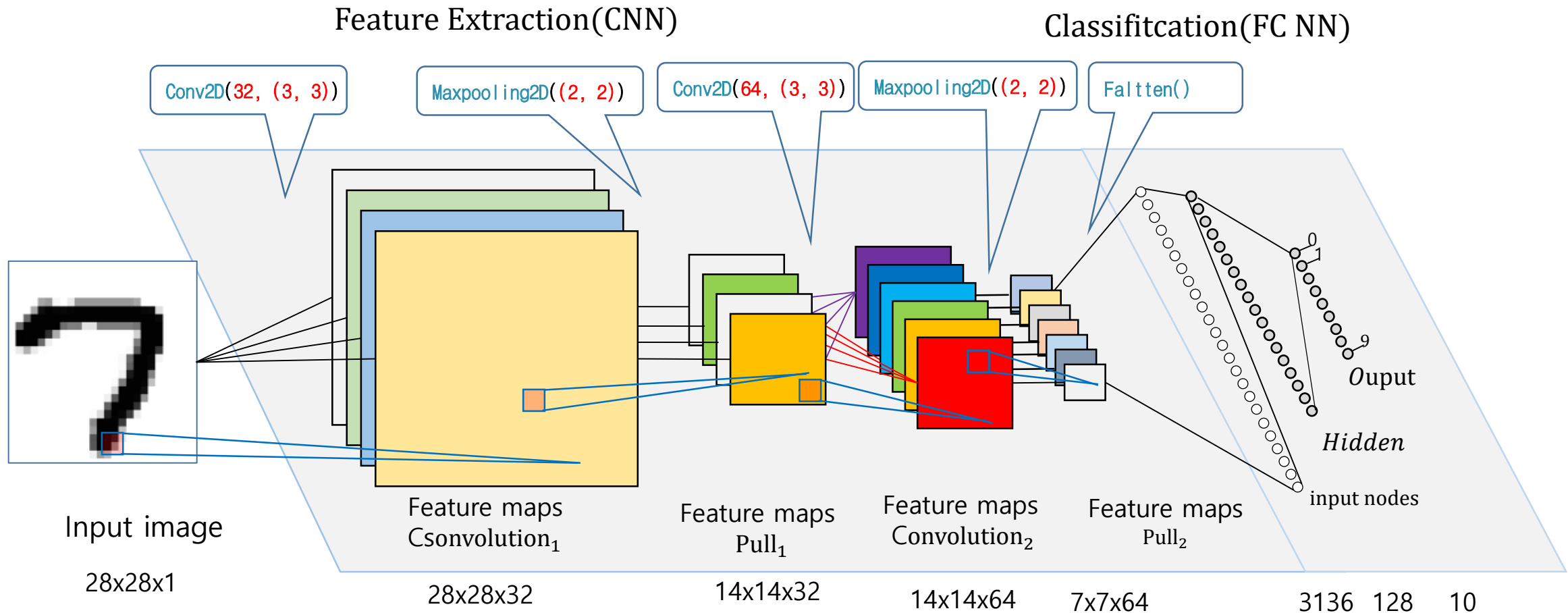
5.2 Mnist digit classifier with deep Convolution layers

- 2 Convolution NN



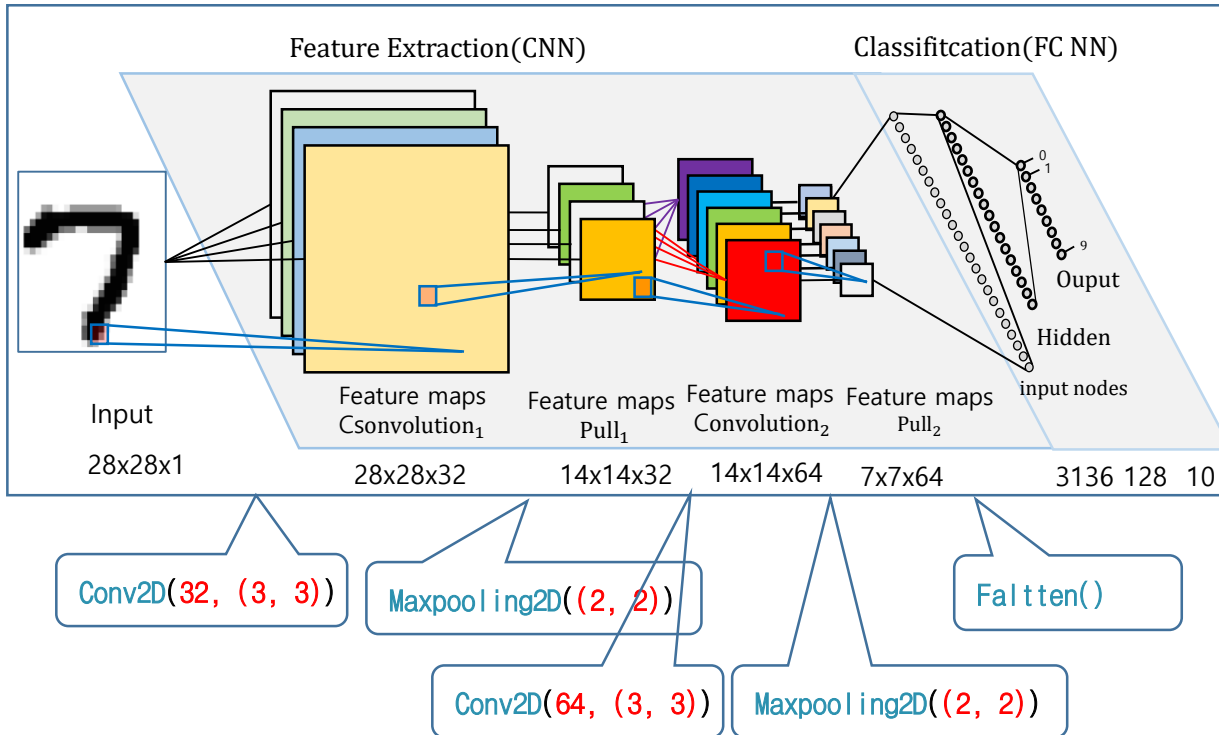
5.2 Mnist digit classifier with deep Convolution layers

- 2 Convolution NN



5.2 Mnist digit classifier with deep Convolution layers

- 2 Convolution NN



```
model = Sequential()
#convolution layer
model.add(Conv2D(32, kernel_size=(3, 3), activation='relu',
                 input_shape=input_shape)) #28,28,32
model.add(MaxPooling2D(pool_size=(2, 2))) #14,14,32
model.add(Dropout(0.25))
model.add(Conv2D(64, kernel_size=(3, 3), activation='relu')) #14,14,64
model.add(MaxPooling2D(pool_size=(2, 2))) #7,7,64
model.add(Dropout(0.25))

model.add(Flatten()) #3136 = 7*7*64
model.add(Dense(128, activation='relu')) #128
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax')) #10
```


5.2 Mnist digit classifier with deep Convolution layers

```
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.utils import to_categorical as ohe

batch_size = 128; num_classes = 10; epochs = 12

# input image dimensions
img_rows, img_cols = 28, 28
input_shape=(img_rows, img_cols,1)

# load mnist image and train and test datasets
(x_train, y_train), (x_test, y_test) = mnist.load_data()
x_train = x_train.reshape(x_train.shape[0], img_rows, img_cols,
                          1).astype(float)/255
x_test = x_test.reshape(x_test.shape[0], img_rows, img_cols,
                        1).astype(float)/255
print('x_train shape: {} y_train shape: {}'.format(x_train.shape, y_train.shape))
print('x_test shape : {} y_test shape : {}'.format(x_test.shape, y_test.shape))

# convert label to one_hot_encoding(label,10)
y_train_ohe = ohe(y_train, num_classes)
y_test_ohe = ohe(y_test, num_classes)
```

```
model = Sequential()
#convolution layer
model.add(Conv2D(32, kernel_size=(3, 3), activation='relu',
                 input_shape=input_shape)) #26,26,32
model.add(MaxPooling2D(pool_size=(2, 2))) #14,14,32
model.add(Dropout(0.25))

model.add(Conv2D(64, kernel_size=(3, 3), activation='relu')) #14,14,64
model.add(MaxPooling2D(pool_size=(2, 2))) #7,7,64
model.add(Dropout(0.25))

model.add(Flatten()) #3136 = 7*7*64
model.add(Dense(128, activation='relu')) #128
model.add(Dropout(0.25))
model.add(Dense(num_classes, activation='softmax')) #10

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

model.fit(x_train, y_train_ohe, validation_data=(x_test, y_test_ohe),
         batch_size=batch_size, epochs=epochs, verbose=1)

score = model.evaluate(x_test, y_test_ohe, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

5.2 Mnist digit classifier with deep Convolution layers

```
from keras.data_loader import DataGenerator from keras.models import Sequential from keras.layers import Conv2D, MaxPooling2D, Flatten, Dense from keras.optimizers import Adam from keras.utils import Progbar

batch_size = 128

# input image
img_rows, img_cols = 28, 28
input_shape = (img_rows, img_cols, 1)

# load mnist
(x_train, y_train), (x_test, y_test) = mnist.load_data()

x_train = x_train.reshape(x_train.shape[0], x_train.shape[1], x_train.shape[2], 1)
x_test = x_test.reshape(x_test.shape[0], x_test.shape[1], x_test.shape[2], 1)

print('x_train shape:', x_train.shape)
print('x_test shape:', x_test.shape)

# convert labels to one-hot
y_train = one_hot(y_train, num_classes=10)
y_test = one_hot(y_test, num_classes=10)

score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

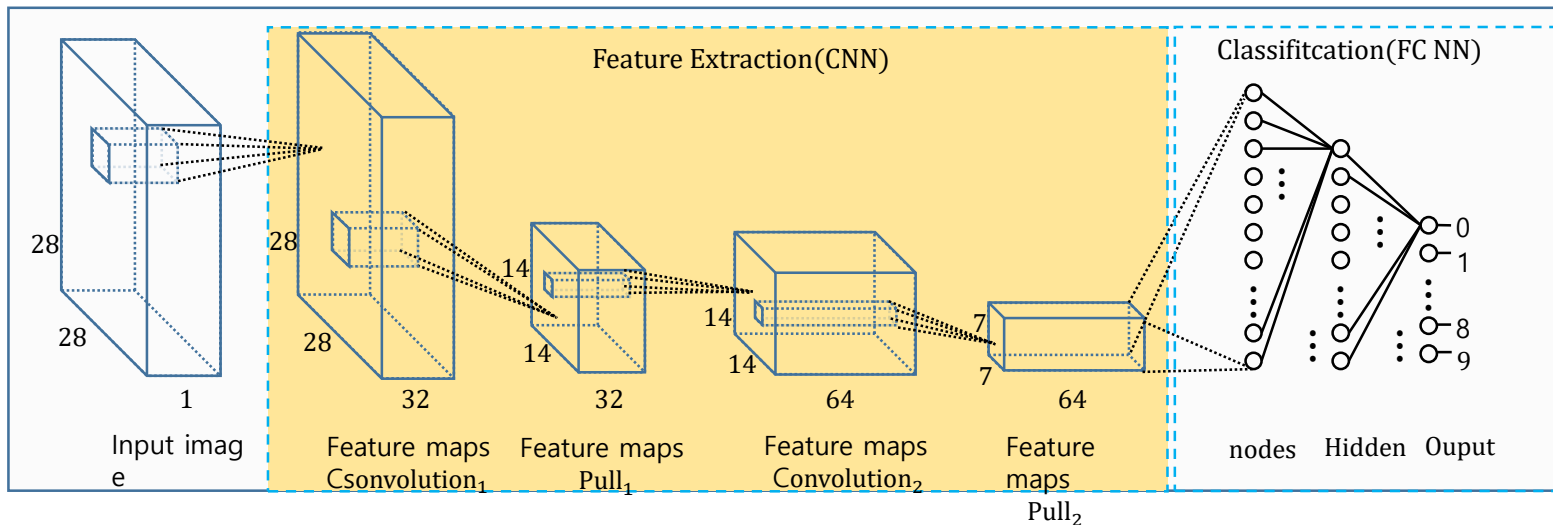
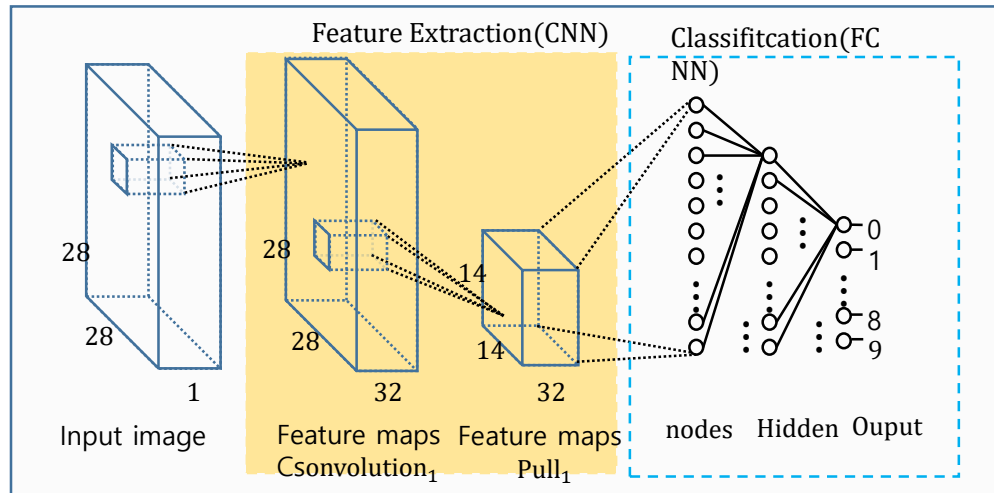
C:\Program Files (x86)\Microsoft Visual Studio\Shared\Python36_64\python.exe

Using TensorFlow backend.
x_train shape: (60000, 28, 28, 1) y_train shape: (60000, 10)
x_test shape: (10000, 28, 28, 1) y_test shape: (10000, 10)
Train on 60000 samples, validate on 10000 samples

Epoch	60000/60000	loss	acc	val_loss	val_acc
1/12	[=====]	0.2960	0.9077	0.0724	0.9766
2/12	[=====]	0.0897	0.9727	0.0442	0.9859
3/12	[=====]	0.0655	0.9799	0.0331	0.9882
4/12	[=====]	0.0526	0.9835	0.0307	0.9896
5/12	[=====]	0.0474	0.9855	0.0278	0.9906
6/12	[=====]	0.0409	0.9871	0.0245	0.9912
7/12	[=====]	0.0359	0.9886	0.0249	0.9916
8/12	[=====]	0.0340	0.9889	0.0250	0.9912
9/12	[=====]	0.0299	0.9904	0.0235	0.9917
10/12	[=====]	0.0271	0.9914	0.0232	0.9924
11/12	[=====]	0.0254	0.9915	0.0207	0.9932
12/12	[=====]	0.0255	0.9917	0.0203	0.9929

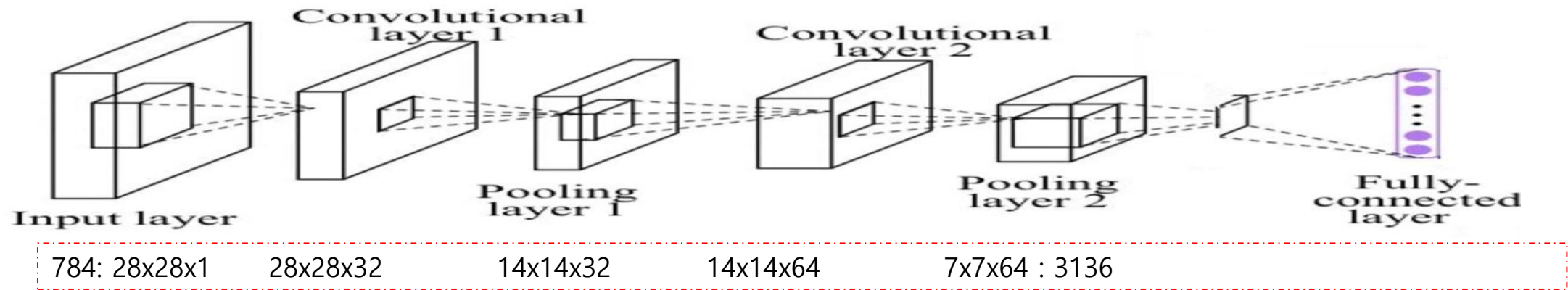
Test loss: 0.02031740538996528
Test accuracy: 0.9929

5.2 Mnist digit classifier with deep Convolution layers



5.3 exercise

- 다음 CNN 구조로 mnist image 인식 시스템을 구현하여 99.3% 이상의 인식률을 얻을 수 있음을 확인 하시오.

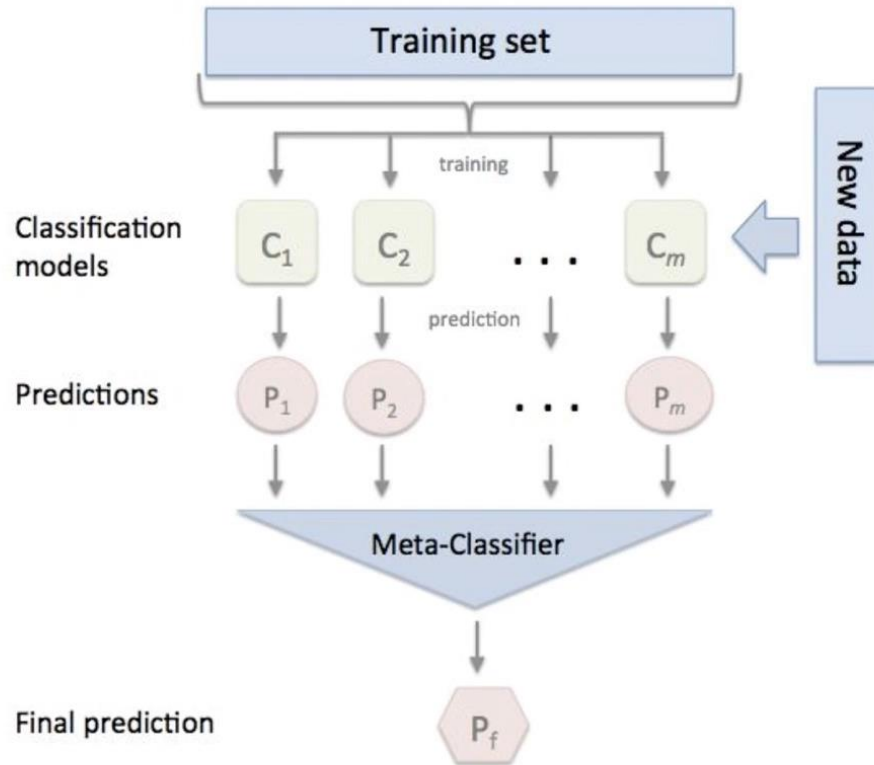


```
Epoch 12/12
60000/60000 [=====]
Test loss: 0.02079462225716561
Test accuracy: 0.9935
Press any key to continue . . .
```

5.4 CIFAR-10

- CIFAR-10 데이터세트(Canadian Institute For Advanced Research)
 - 일반적으로 머신 러닝 및 컴퓨터 비전 알고리즘을 훈련시키는 데 사용되는 이미지 모음입니다.
 - 10 가지 클래스로 구성된 60,000 개의 32x32 컬러 이미지가 포함되어 있습니다.
 - 10 가지 다른 클래스는 비행기, 자동차, 새, 고양이, 사슴, 개, 개구리, 말, 배 및 트럭을 나타냅니다. 각 클래스마다 6,000 개의 이미지가 있습니다.
- [ConvNetJS demo: training on CIFAR-10]
 - <http://cs.Stanford.edu/people/karpathy/convnetjs/demo/cifar10.html>

5.5 Ensemble

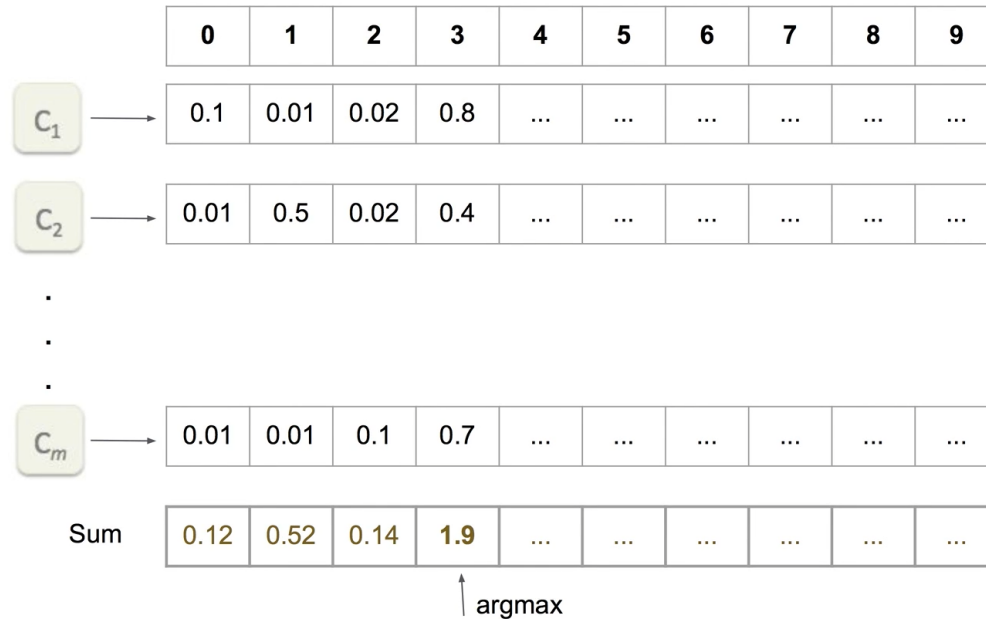


$$\begin{aligned} P_1 &= C_1.predict(y_{train}) \\ P_2 &= C_2.predict(y_{train}) \\ &\vdots \\ P_n &= C_n.predict(y_{train}) \end{aligned}$$

http://rasbt.github.io/mlxtend/user_guide/classifier/StackingClassifier/

5.5 Ensemble

Ensemble prediction



```
import numpy as np
```

```
predictions=np.zeros(10,dtype=float)
```

```
for i, model in enumerate(models):
```

```
    acc=model.evaluate(X_train,Y_train_ohe)
```

```
    print(' model[{}] acc:{}'.format(i,acc))
```

```
    p=model.predict(X_train) #p=[0.1, 0.3, 0.2, .W0.5,,,]
```

```
    predictions =predictions+p #[0..9] =[0..9]+[0..9]
```

```
ensemble_predicts=np.equal(np.argmax(predictions,1),
```

```
                            np.argmax(Y_train_ohe,1))
```

```
ensemble_accuracy=ensemble_predicts.mean()
```

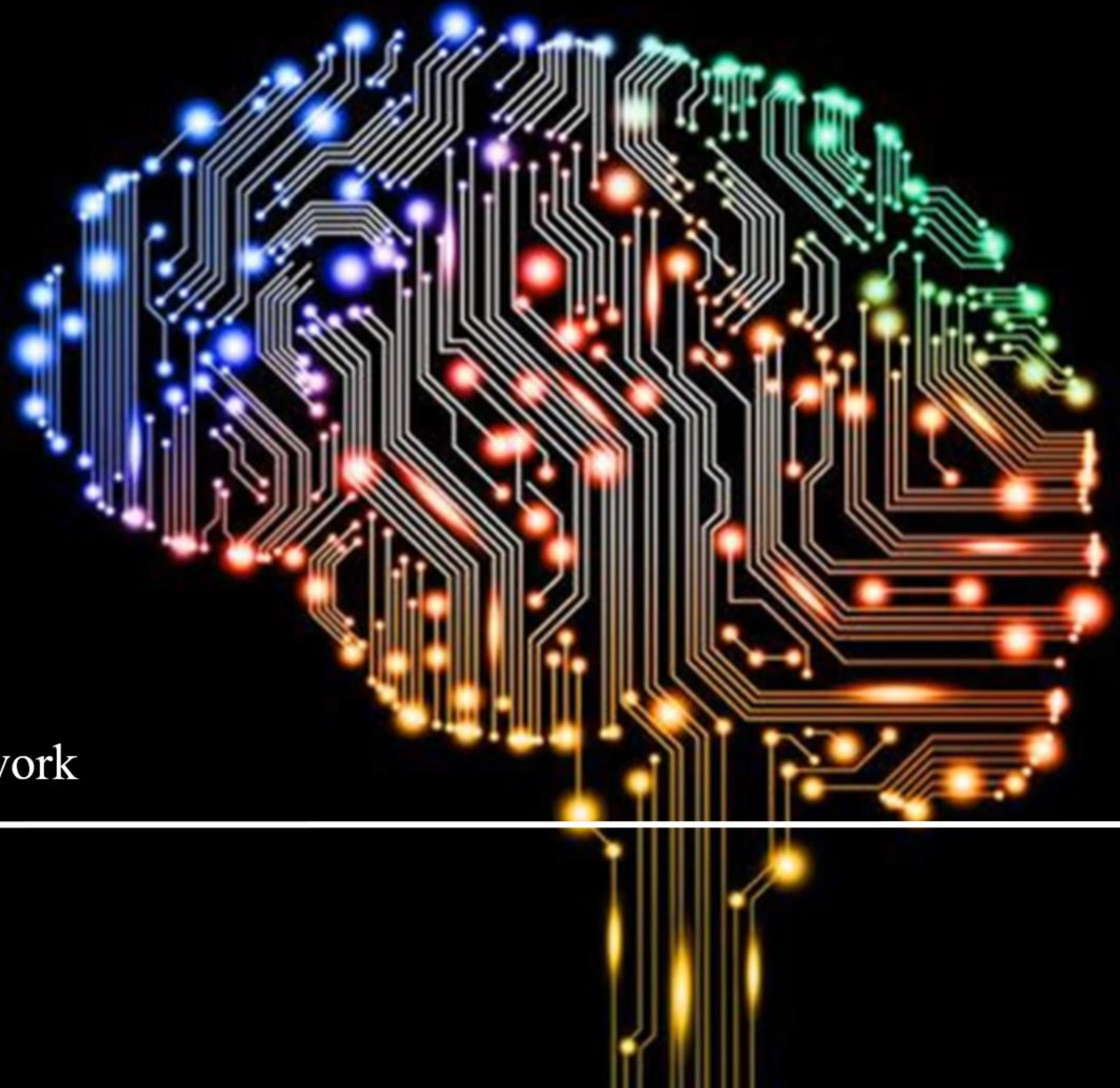
```
print('Ensemble accuracy : ',ensemble_accuracy)
```

Recap

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