

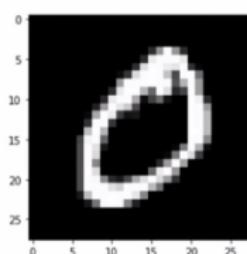
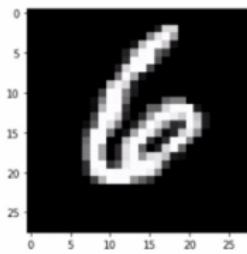
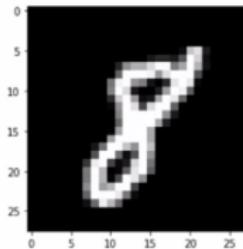
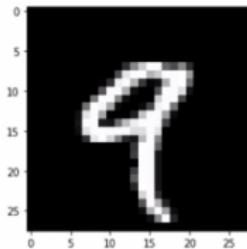
## 제4장: 세가지 기본 신경망II: CNN 기초

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# Convolution



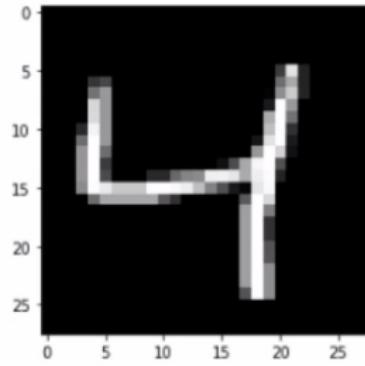
숫자	전체	위	아래
9	1	1	0
8	2	1	1
6	1	0	1
0	1	0	0

특정한 패턴의 특징이  
어디서 나타나는지를 확인하는 도구

# Convolution

특정한 패턴의 특징이  
어디서 나타나는지를 확인하는 도구

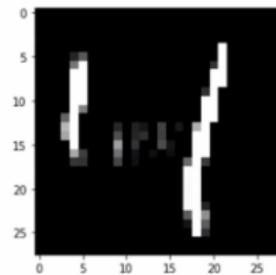
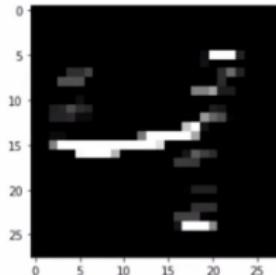
**“Convolution”**



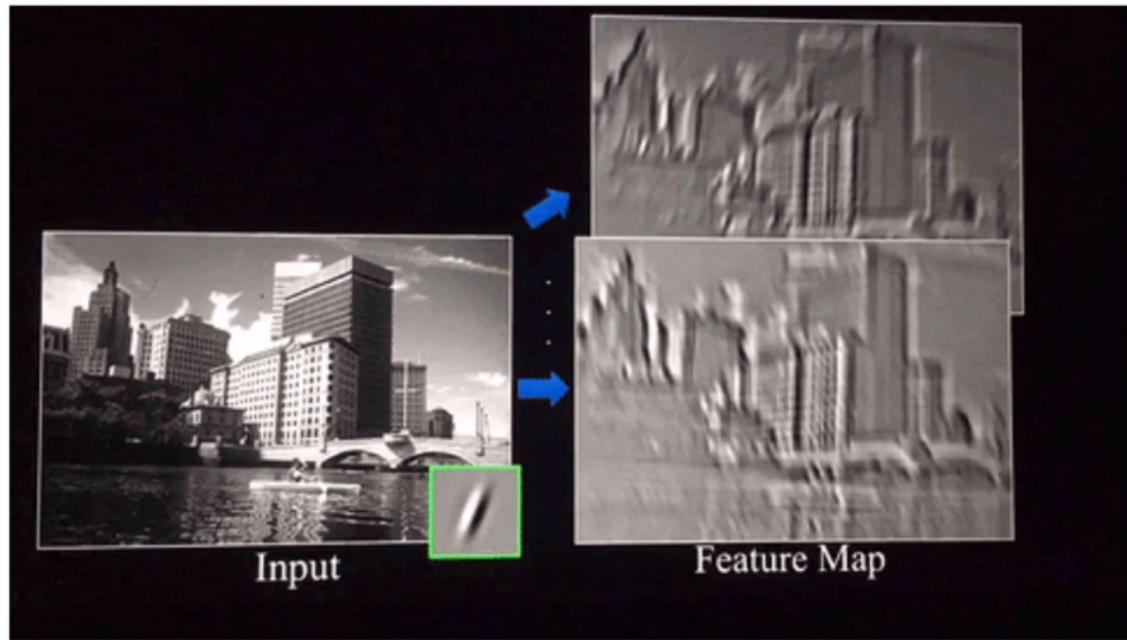
필터



특징맵 feature map



# Convolution



# Convolution

## Convolution Layer

```
[ ] import tensorflow as tf
import pandas as pd
```

```
[ ] # 데이터를 준비하고
(독립, 종속), _ = tf.keras.datasets.mnist.load_data()
print(독립.shape, 종속.shape)

독립 = 독립.reshape(60000, 28, 28, 1)
종속 = pd.get_dummies(종속)
print(독립.shape, 종속.shape)
```

```
[ ] # 모델을 만들고
X = tf.keras.layers.Input(shape=[28, 28, 1])
H = tf.keras.layers.Conv2D(3, kernel_size=5, activation='swish')(X)
H = tf.keras.layers.Conv2D(6, kernel_size=5, activation='swish')(H)
H = tf.keras.layers.Flatten()(H)
H = tf.keras.layers.Dense(84, activation='swish')(H)
Y = tf.keras.layers.Dense(10, activation='softmax')(H)
model = tf.keras.models.Model(X, Y)
model.compile(loss='categorical_crossentropy', metrics='accuracy')
```

```
[ ] # 모델을 학습하고
model.fit(독립, 종속, epochs=10)
```

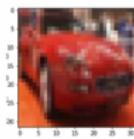
# Filters

## 필터의 이해

1. 필터셋은 3차원 형태로 된 가중치의 모음
2. 필터셋 하나는 앞선 레이어의 결과인 “특징맵” 전체를 본다.
3. 필터셋 개수 만큼 특징맵을 만든다.

`Conv2D(3, kernel_size=5, activation='swish')`

`Conv2D(6, kernel_size=5, activation='swish')`



(32, 32, 3)



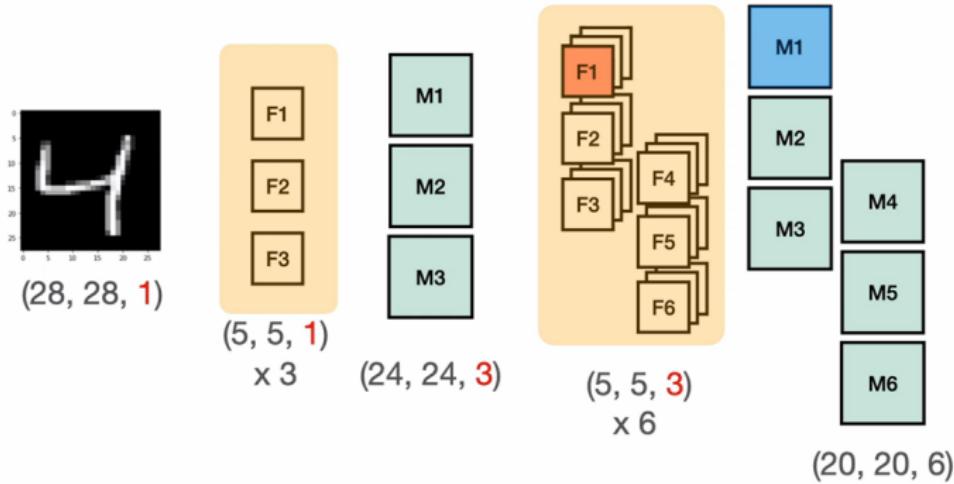
(3, 5, 5, ?)



(6, 5, 5, ?)

# Filters

`Conv2D(3, kernel_size=5, activation='swish')`  
`Conv2D(6, kernel_size=5, activation='swish')`



# Convolution 연산

(8, 8, 1)

x1	x2	x3	x4	x5	x6	x7	x8
x9	x10	x11	x12	x13	x14	x15	x16
x17	x18	x19	...				

(3, 3, 1)

w1	w2	w3
w4	w5	w6
w7	w8	w9

필터

(6, 6)

y1	y2	y3	y4	y5	y6
y7	y8	y9	y10	...	

$$\begin{aligned}y1 = & x1 * w1 + x2 * w2 + x3 * w3 + \\& x9 * w4 + x10 * w5 + x11 * w6 + \\& x17 * w7 + x18 * w8 + x19 * w9\end{aligned}$$

# Convolution 연산

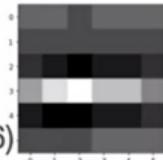
(8, 8, 1)

0	0	0	0	0	0	0	0	0
0	1	0	0	0	0	1	0	0
0	1	0	0	0	0	1	0	0
0	1	0	0	0	0	1	0	0
0	1	1	1	1	1	1	1	0
0	0	0	0	0	1	0	0	0
0	0	0	0	0	0	1	0	0
0	0	0	0	0	0	0	0	0

(3, 3, 1)

-1	-1	-1
2	2	2
-1	-1	-1

필터



$$Y = \begin{array}{l} * -1 + * -1 + * -1 + \\ * 2 + * 2 + * 2 + \\ * -1 + * -1 + * -1 \end{array}$$

1	1	0	1	1	1
0	0	0	0	0	0
-1	-2	-3	-2	-2	-1
3	5	6	4	4	2
-2	-3	-3	-2	-2	-1
0	0	0	1	1	1

# Max-pooling

```
X = tf.keras.layers.Input(shape=[28, 28])
H = tf.keras.layers.Flatten()(X)
H = tf.keras.layers.Dense(84, activation='swish')(H)
Y = tf.keras.layers.Dense(10, activation='softmax')(H)
model = tf.keras.models.Model(X, Y)
model.compile(loss='categorical_crossentropy', metrics='accuracy')
```

$$84 * (784 + 1) = 65940$$

$$10 * (84 + 1) = 850$$

`model.summary()`

Model: "model"		
Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 28, 28)]	0
flatten (Flatten)	(None, 784)	0
dense (Dense)	(None, 84)	65940
dense_1 (Dense)	(None, 10)	850

Total params: 66,790  
 Trainable params: 66,790  
 Non-trainable params: 0

# Max-pooling

```
X = tf.keras.layers.Input(shape=[28, 28, 1])
H = tf.keras.layers.Conv2D(3, kernel_size=5, activation='swish')(X)
H = tf.keras.layers.Conv2D(6, kernel_size=5, activation='swish')(H)
H = tf.keras.layers.Flatten()(H)
H = tf.keras.layers.Dense(84, activation='swish')(H)
Y = tf.keras.layers.Dense(10, activation='softmax')(H)
model = tf.keras.models.Model(X, Y)
model.compile(loss='categorical_crossentropy')
```

`model.summary()`

$$84 * (2400 + 1) = 201,684$$

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[None, 28, 28, 1]	0
conv2d (Conv2D)	(None, 24, 24, 3)	78
conv2d_1 (Conv2D)	(None, 20, 20, 6)	456
flatten (Flatten)	(None, 2400)	0
dense (Dense)	(None, 84)	201684
dense_1 (Dense)	(None, 10)	850

Total params: 203,068  
 Trainable params: 203,068  
 Non-trainable params: 0

# Pooling!!

# Max-pooling

```
X = tf.keras.layers.Input(shape=[28, 28, 1])

H = tf.keras.layers.Conv2D(3, kernel_size=5, activation='swish')(X)
H = tf.keras.layers.MaxPool2D()(H)
H = tf.keras.layers.Conv2D(6, kernel_size=5, activation='swish')(H)
H = tf.keras.layers.MaxPool2D()(H)

H = tf.keras.layers.Flatten()(H)
H = tf.keras.layers.Dense(84, activation='swis'
Y = tf.keras.layers.Dense(10, activation='soft'
model = tf.keras.models.Model(X, Y)
model.compile(loss='scategorical_crossentropy')

model.summary()
```

$$84 * (96 + 1) = 8,148$$

Model: "model_1"		
Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, 28, 28, 1)]	0
conv2d_2 (Conv2D)	(None, 24, 24, 3)	78
max_pooling2d (MaxPooling2D)	(None, 12, 12, 3)	0
conv2d_3 (Conv2D)	(None, 8, 8, 6)	456
max_pooling2d_1 (MaxPooling2D)	(None, 4, 4, 6)	0
flatten_1 (Flatten)	(None, 96)	0
dense_2 (Dense)	(None, 84)	8148
dense_3 (Dense)	(None, 10)	850

Total params: 9,532  
Trainable params: 9,532  
Non-trainable params: 0

# Max-pooling

나는 특징맵

10	7	2	42	30	25
1	16	33	8	7	6
9	22	11	45	10	22
13	15	31	27	12	41
7	20	19	30	23	6
3	24	15	27	14	40

MaxPooling

16	42	30
22	45	41
24	30	40

값이 크다 = 필터로 찾으려는 특징이 많이 나타난 부분

# Max-pooling

# 1. 과거의 데이터를 준비합니다.

```
(독립, 종속), _ = tf.keras.datasets.mnist.load_data()
독립 = 독립.reshape(60000, 28, 28, 1)
종속 = pd.get_dummies(종속)
print(독립.shape, 종속.shape)
```

# 2. 모델의 구조를 만듭니다

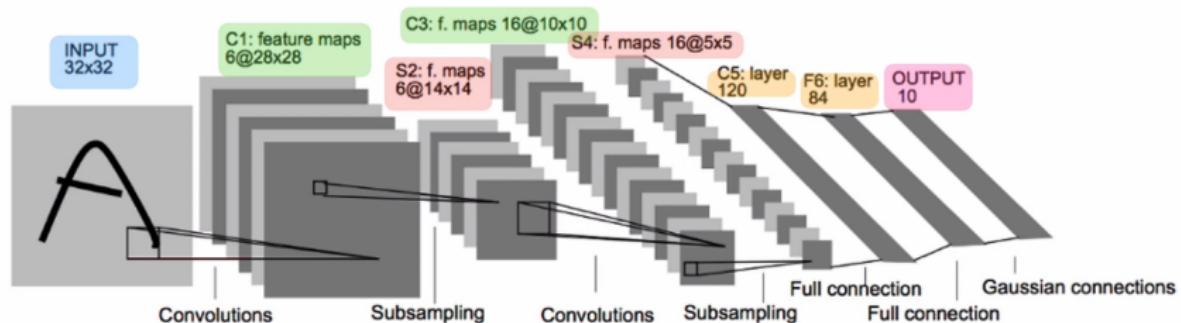
```
X = tf.keras.layers.Input(shape=[28, 28, 1])

H = tf.keras.layers.Conv2D(3, kernel_size=5, activation='swish')(X)
H = tf.keras.layers.MaxPool2D()(H)
H = tf.keras.layers.Conv2D(6, kernel_size=5, activation='swish')(H)
H = tf.keras.layers.MaxPool2D()(H)

H = tf.keras.layers.Flatten()(H)
H = tf.keras.layers.Dense(84, activation='swish')(H)
Y = tf.keras.layers.Dense(10, activation='softmax')(H)

model = tf.keras.models.Model(X, Y)
model.compile(loss='categorical_crossentropy', metrics='accuracy')
```

# LeNet 5



CNN called LeNet by Yann LeCun (1998)

# LeNet 5

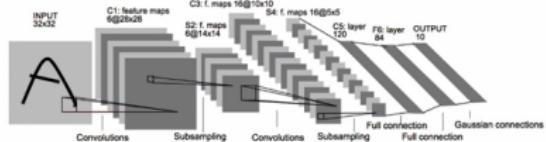
## # 1. 과거의 데이터를 준비합니다.

```
(독립, 종속), _ = tf.keras.datasets.mnist.load_data()
독립 = 독립.reshape(60000, 28, 28, 1)
종속 = pd.get_dummies(종속)
print(독립.shape, 종속.shape)
```

## # 2. 모델의 구조를 만듭니다

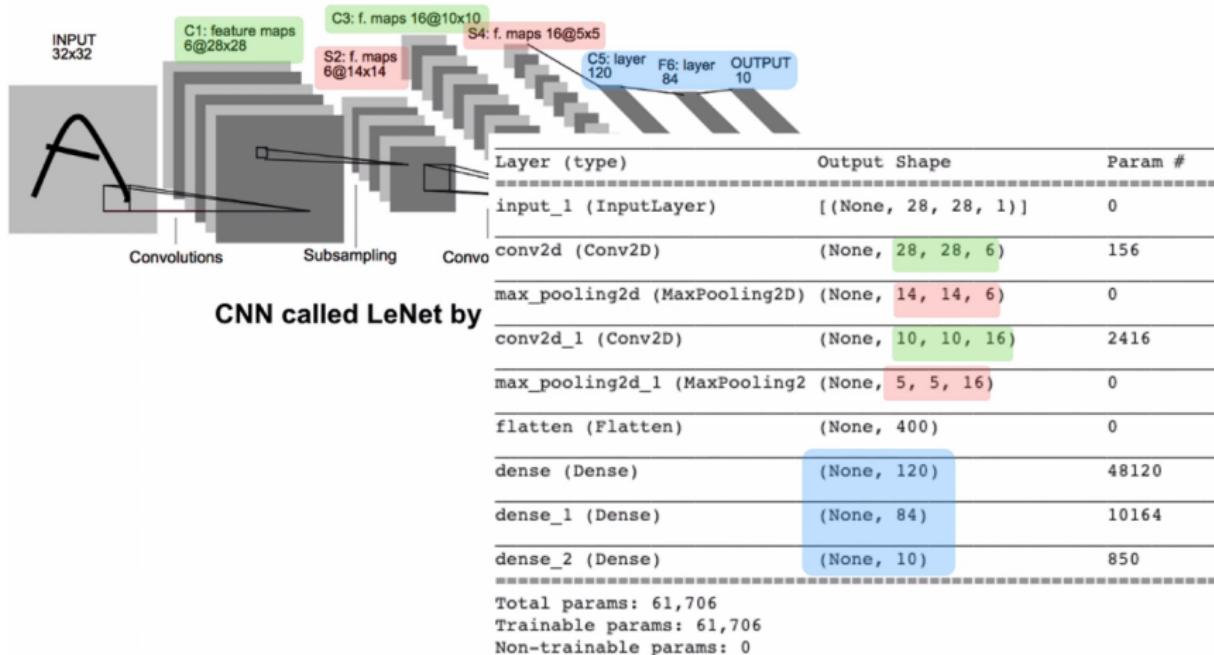
```
X = tf.keras.layers.Input(shape=[28, 28, 1])
H = tf.keras.layers.Conv2D(6, kernel_size=5, padding='same', activation='swish')(X)
H = tf.keras.layers.MaxPool2D()(H)
H = tf.keras.layers.Conv2D(16, kernel_size=5, activation='swish')(H)
H = tf.keras.layers.MaxPool2D()(H)

H = tf.keras.layers.Flatten()(H)
H = tf.keras.layers.Dense(120, activation='swish')(H)
H = tf.keras.layers.Dense(84, activation='swish')(H)
Y = tf.keras.layers.Dense(10, activation='softmax')(H)
model = tf.keras.models.Model(X, Y)
model.compile(loss='categorical_crossentropy', metrics='accuracy')
```



CNN called LeNet by Yann LeCun (1998)

# LeNet 5



# Python 실습

내 이미지 사용



```

2a  wget -q https://raw.githubusercontent.com/blackdew/tensorflow1/master/csv/notMNIST_small.tar.gz
      !tar -xzf notMNIST_small.tar.gz

[ ] # 이미지 읽어서 데이터 준비하기
import glob
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

paths = glob.glob('./notMNIST_small/*/*.png')
paths = np.random.permutation(paths)
독립 = np.array([plt.imread(paths[i]) for i in range(len(paths))])
종속 = np.array([paths[i].split('/')[-2] for i in range(len(paths))])
print(독립.shape, 종속.shape)

```

# Python 실습

```
[ ] 종속[0:10]

[ ] plt.imshow(독립[0], cmap='gray')

[ ] 독립 = 독립.reshape(18724, 28, 28, 1)
종속 = pd.get_dummies(종속)
print(독립.shape, 종속.shape)

❸ import tensorflow as tf

# 모델을 완성합니다.
X = tf.keras.layers.Input(shape=[28, 28, 1])

H = tf.keras.layers.Conv2D(6, kernel_size=5, padding='same', activation='swish')(X)
H = tf.keras.layers.MaxPool2D()(H)

H = tf.keras.layers.Conv2D(16, kernel_size=5, activation='swish')(H)
H = tf.keras.layers.MaxPool2D()(H)

H = tf.keras.layers.Flatten()(H)
H = tf.keras.layers.Dense(120, activation='swish')(H)
H = tf.keras.layers.Dense(84, activation='swish')(H)
Y = tf.keras.layers.Dense(10, activation='softmax')(H)

model = tf.keras.models.Model(X, Y)
model.compile(loss='categorical_crossentropy', metrics='accuracy')

[ ] model.fit(독립, 종속, epochs=10)
```

# Reference

## Tensorflow 102 - 이미지 분류(CNN)

Tensorflow 102 - 이미지 분류(CNN)  공동공부 (81명)

[커버 페이지](#)

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- [다섯번째 딥러닝 완성 - LeNet](#)
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