



# The "Hello World" of machine learning

JEJ

- Intro to Machine Learning
- Basic computer vision with ML
- Introducing convolutional neural networks
- Build an image classifier for rock,paper,scissors



# Intro to Machine Learning





사람이 인식하는 것과 같은  
방식으로 컴퓨터가 인식하면 어떨까?

[illegible][illegible]

사람이 인식하는 것과 같은 방식으로  
컴퓨터가 인식하면 어떨까?



머신러닝의 핵심, 인공지능이 추구하는 것



# Intro to Machine Learning

동그란것=묵..  
손가락 세개 또는 두개=가위..  
등등..



Rules

Data

Traditional  
Programming

Answers

묵,찌,빠



# Intro to Machine Learning

동그란것= 목..  
손가락 세개 또는 두개=가위..  
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Rules

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목,찌,빠



Answers

Data

Machine  
Learning

Rules

동그란것= 목..  
손가락 세개 또는 두개=가위..  
등등..



# Intro to Machine Learning

$X = -1, 0, 1, 2, 3, 4$

$Y = -2, 1, 4, 7, 10, 13$





# Intro to Machine Learning

$X = -1, 0, 1, 2, 3, 4$

$Y = -2, 1, 4, 7, 10, 13$

$$Y = 3X + 1$$

# Intro to Machine Learning

모델 정의

```
model = tf.keras.Sequential([keras.layers.Dense(units=1, input_shape=[1])])  
# 한개의 레이어가 있고, 그 레이어는 한개의 뉴런을 지녔고, 인풋쉐입은 1이다.
```

모델 컴파일

```
model.compile(optimizer='sgd', loss='mean_squared_error')  
# 추측 -> 평가 -> 추측 -> 평가...
```

x 데이터

```
xs = np.array([-1.0, 0.0, 1.0, 2.0, 3.0, 4.0], dtype=float)
```

y 데이터

```
ys = np.array([-2.0, 1.0, 4.0, 7.0, 10.0, 13.0], dtype=float)
```

```
model.fit(xs, ys, epochs=500)  
# x값을 y의 값에다가 피팅
```

```
print(model.predict([10.0]))  
# x에 10이 들어가면 뭐가 나올까?
```

모델이란? 학습된 뉴런 네트워크

# Intro to Machine Learning

```
model = tf.keras.Sequential([keras.layers.Dense(units=1, input_shape=[1])])  
# 한개의 레이어가 있고, 그 레이어는 한개의 뉴런을 지녔고, 인풋쉐입은 1이다.  
  
model.compile(optimizer='sgd', loss='mean_squared_error')  
# 추측 -> 평가 -> 추측 -> 평가....  
xs = np.array([-1.0, 0.0, 1.0, 2.0, 3.0, 4.0], dtype=float)  
ys = np.array([-2.0, 1.0, 4.0, 7.0, 10.0, 13.0], dtype=float)  
  
model.fit(xs, ys, epochs=500)  
# x값을 y의 값에다가 찢시킴  
  
print(model.predict([10.0]))  
# x에 10이 들어가면 뭐가 나올까?
```

31가 나와야되는데 실제로는  
31.00321가 나온다 왜일까요?

# Intro to Machine Learning

```
model = tf.keras.Sequential([keras.layers.Dense(units=1, input_shape=[1])])  
# 한개의 레이어가 있고, 그 레이어는 한개의 뉴런을 지녔고, 인풋쉐입은 1이다.  
  
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# 추측 -> 평가 -> 추측 -> 평가....  
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```

훈련데이터가 6쌍 뿐이기 때문  
두 관계가 일직선처럼 보이지만  
그 밖의 값은 일직선이 아닐수도 있는 확률이 존재





# Intro to Machine Learning

Hello ML world

<https://colab.research.google.com/drive/1eS0laRdvRMejyh1jYxj12zsj6YEuDAG->



# Basic computer vision with ML



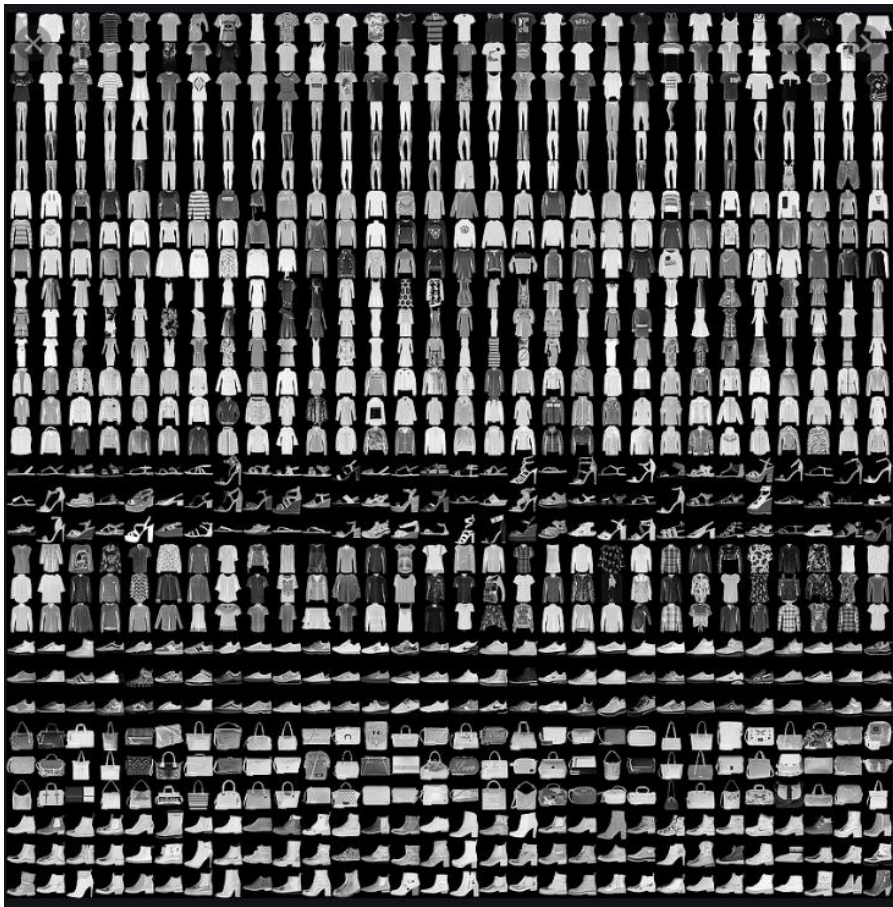
신발은 몇개일까?



# Basic computer vision with ML



# Basic computer vision with ML



- 0 T-shirt/top
- 1 Trouser
- 2 Pullover
- 3 Dress
- 4 Coat
- 5 Sandal
- 6 Shirt
- 7 Sneaker
- 8 Bag
- 9 Ankle boot

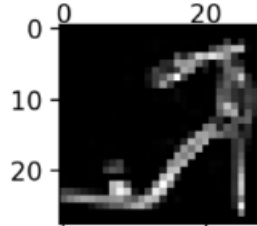
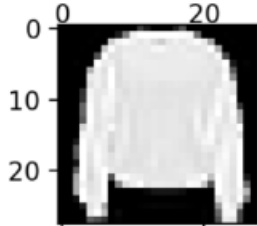
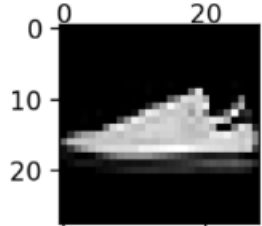
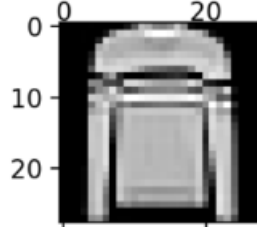
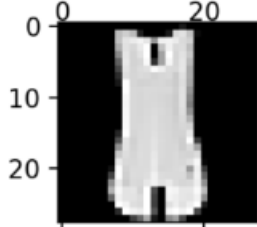
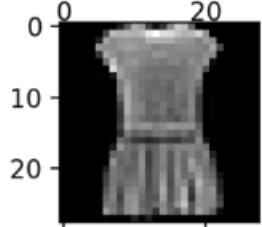
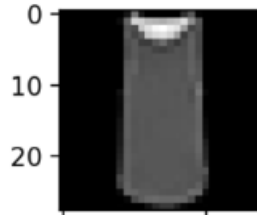
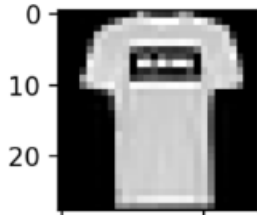
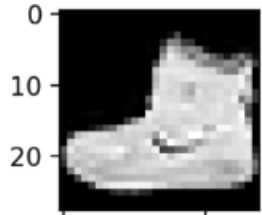
## Fashion MNIST

- 총 70000개의 이미지
- 10개의 카테고리
- 28X28 크기의 이미지

<https://github.com/zalandoresearch/fashion-mnist>



# Basic computer vision with ML



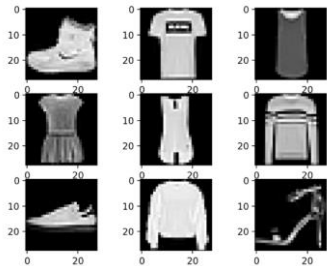
- 0 T-shirt/top
- 1 Trouser
- 2 Pullover
- 3 Dress
- 4 Coat
- 5 Sandal
- 6 Shirt
- 7 Sneaker
- 8 Bag
- 9 Ankle boot

## Fashion MNIST

- 총 70000개의 이미지
- 10개의 카테고리
- 28X28 크기의 이미지

# Basic computer vision with ML

```
import tensorflow as tf
from tensorflow import keras
# fashion_mnist 데이터는 tensorflow에 내장되어있어서
fashion_mnist = keras.datasets.fashion_mnist
(train_images, train_labels), (test_images, test_labels) = fashion_mnist.load_data()
# 쉽게 불러올 수 있다
```

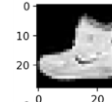


0 T-shirt/top  
1 Trouser  
2 Pullover  
3 Dress  
4 Coat  
...

훈련 이미지는 60,000개

테스트 이미지는 10,000개

# Basic computer vision with ML



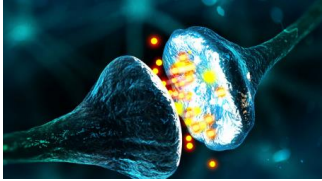
```
model = tf.keras.models.Sequential([tf.keras.layers.Flatten(input_shape=(28,28)),  
                                     tf.keras.layers.Dense(128, activation='relu'),  
                                     tf.keras.layers.Dense(10, activation='softmax')])
```



0 T-shirt/top  
1 Trouser  
2 Pullover  
3 Dress  
4 Coat  
...  
10

# Basic computer vision with ML

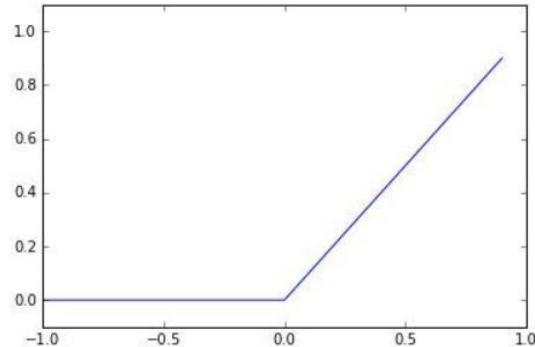
```
model = tf.keras.models.Sequential([tf.keras.layers.Flatten(input_shape=(28,28)),  
                                     tf.keras.layers.Dense(128, activation='relu'),  
                                     tf.keras.layers.Dense(10, activation='softmax')])
```



Firing of a neuron

<ReLU>

```
if (x>0){  
    return x;  
}  
else{  
    return 0;  
}
```



Activation function(활성화 함수)  
Input data -> {activation function} -> output data

```
import numpy as np  
def relu(x):  
    return np.maximum(0, x)
```

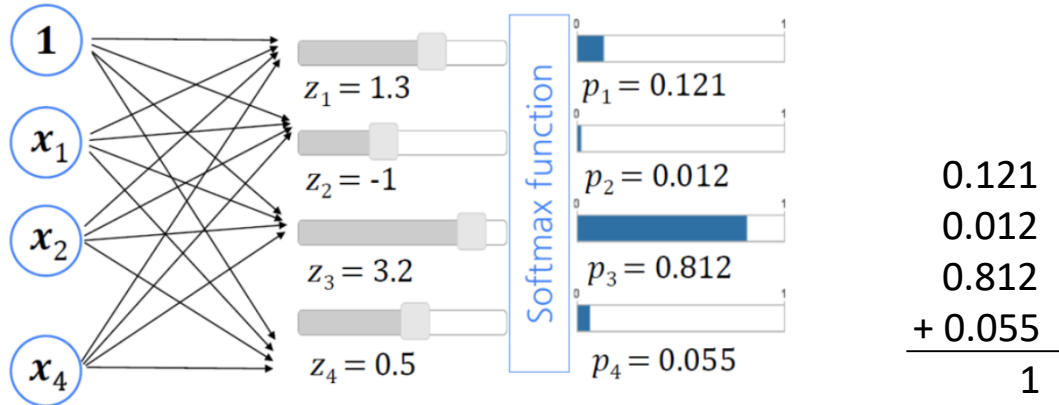


# Basic computer vision with ML

```
model = tf.keras.models.Sequential([tf.keras.layers.Flatten(input_shape=(28,28)),  
                                     tf.keras.layers.Dense(128, activation='relu'),  
                                     tf.keras.layers.Dense(10, activation='softmax')])
```



```
import numpy as np  
def softmax(arr):  
    arr = np.exp(arr)  
    return arr / np.sum(arr)
```



편차란? 관측값과 평균의 차이

Softmax는 최종 출력 단계에서 N가지 범주로 분류하는 Multi-class classification 에 쓰임



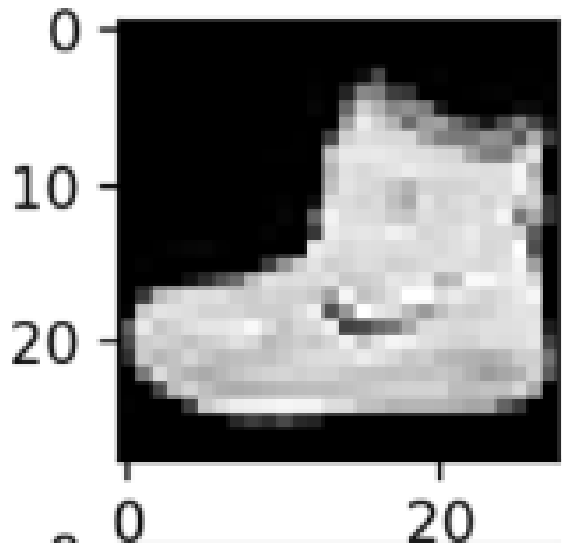
# Basic computer vision with ML

2\_Computer\_Vision\_FashionMNIST\_end

<https://colab.research.google.com/drive/1eRRtGqB0396cqkrdCCWBukVJ433tVLWJ>

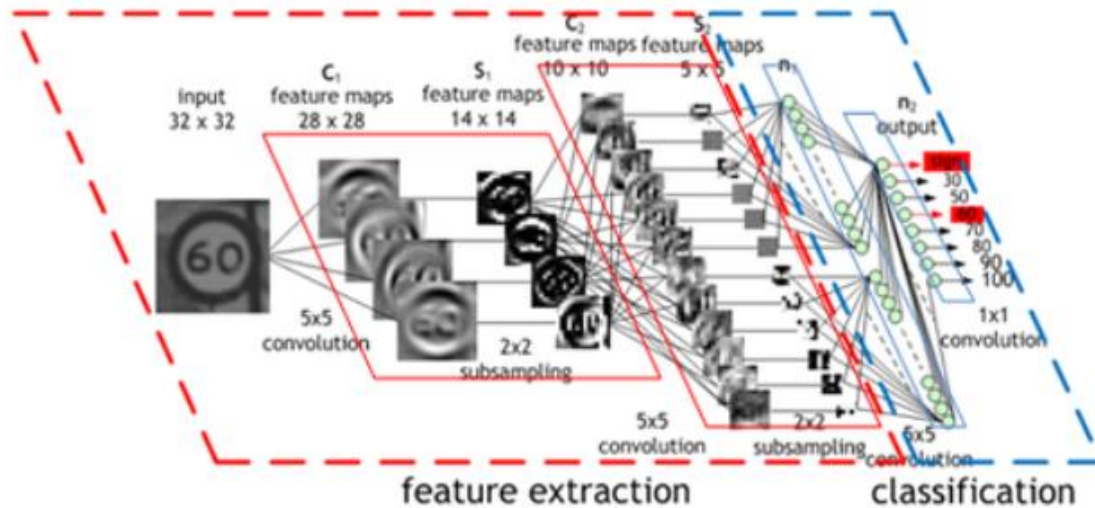


# convolutional neural networks



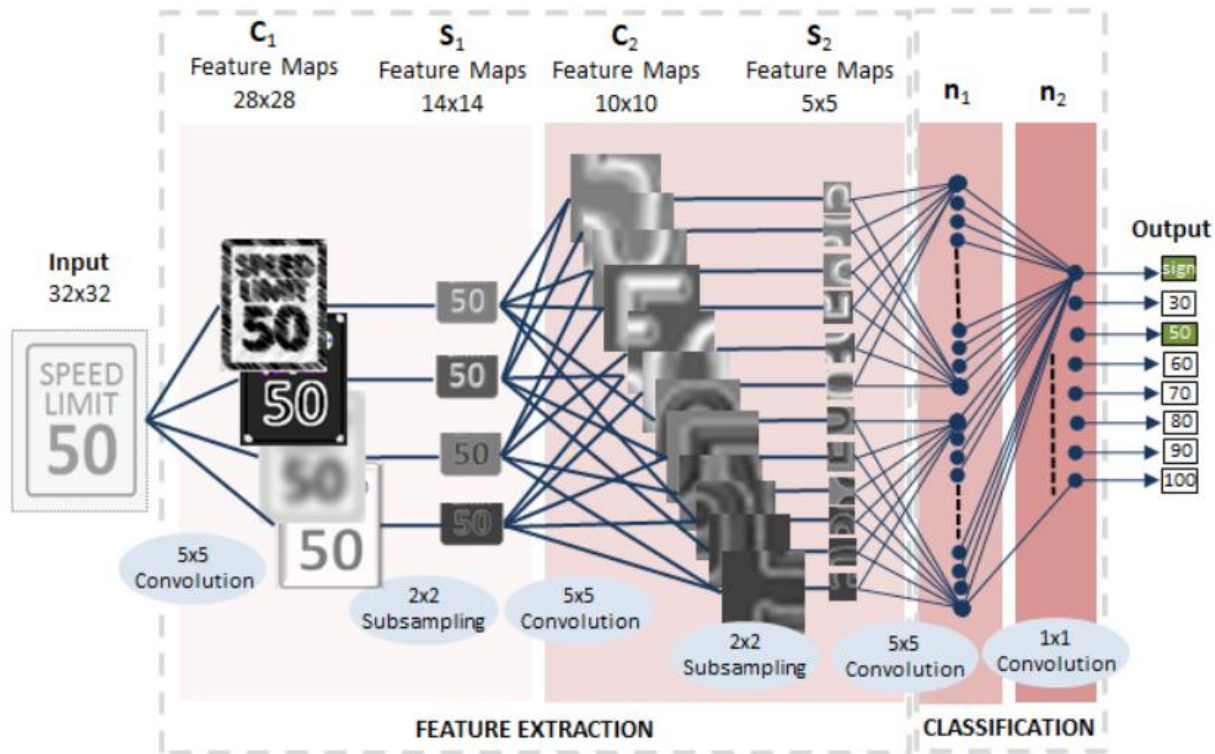


# convolutional neural networks





# convolutional neural networks







# convolutional neural networks

```
model = tf.keras.models.Sequential([
    tf.keras.layers.Conv2D(64,(3,3), activation='relu',
                           input_shape=(28,28,1)),
    tf.keras.layers.MaxPooling2D(2,2),

    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(128, activation='relu'),
    tf.keras.layers.Dense(10, activation='softmax')])
```



# convolutional neural networks

## 풀링 (Sub sampling or Pooling)

컨볼루셔널 레이어를 통해 어느정도 특징이 추출되었으면 이 모든 특징을 가지고 판단을 할 필요가 없다. (고해상도 저해상도 사진)

그래서, 인위로 줄이는 작업을 하는데, 이 작업을 sub sampling 또는 pooling 이라고 한다.

[Pooling의 종류]

- Max pooling
- Average pooling,

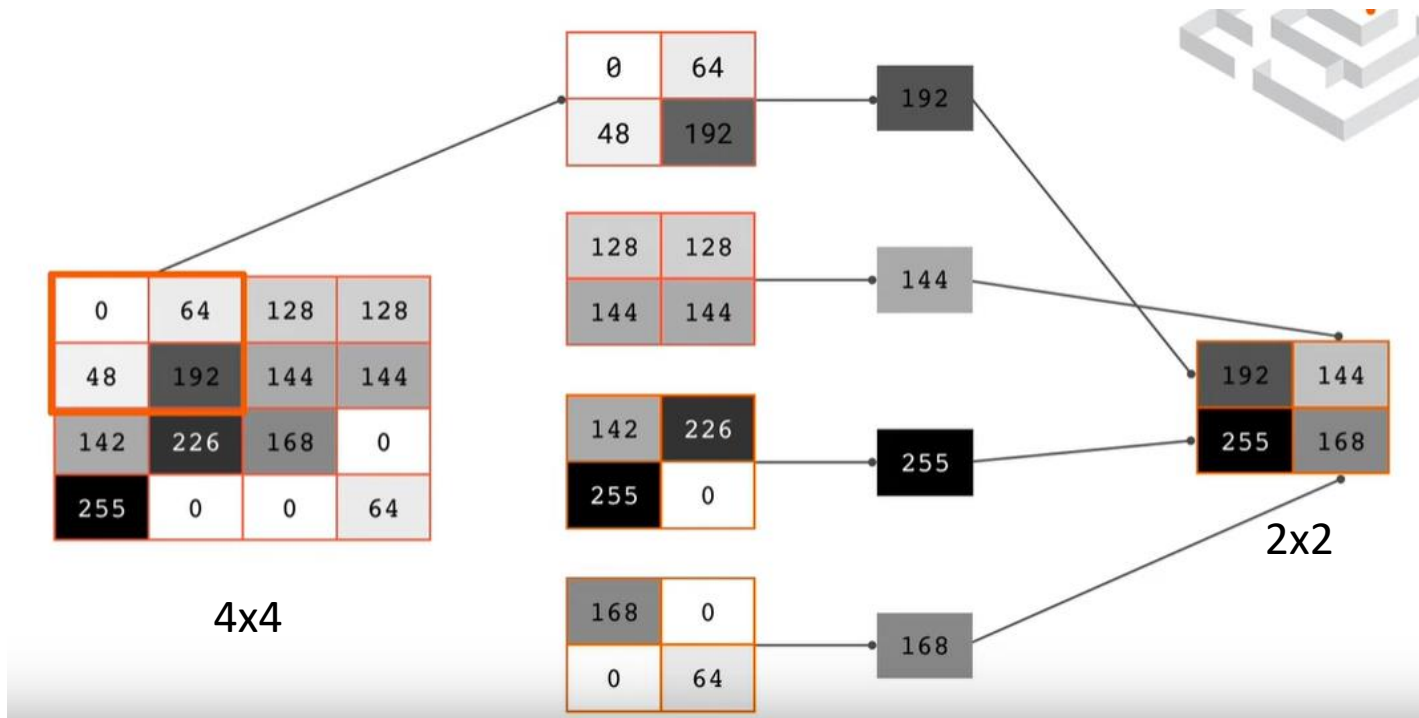
등이 있고, 그중에서 max pooling 이라는 기법이 많이 사용된다.

이미지의 크기는 줄어들지만 특징들은 강조되게 된다.



# convolutional neural networks

Max pooling(맥스풀링) : 가장 큰 값을 뽑아냄





# convolutional neural networks

Average pooling(에버리지풀링) : 평균 값을 뽑아냄

Max Pooling

29	15	28	184
0	100	70	38
12	12	7	2
12	12	45	6

2 x 2  
pool size

100	184
12	45

Average Pooling

31	15	28	184
0	100	70	38
12	12	7	2
12	12	45	6

2 x 2  
pool size

36	80
12	15

$$(31+15+0+100)/4 = 36$$
$$(28+184+70+38)/4 = 80$$





# convolutional neural networks



Max pooling 2x2



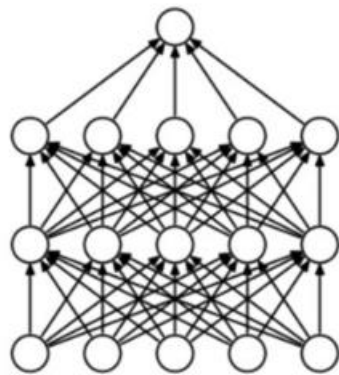


# convolutional neural networks

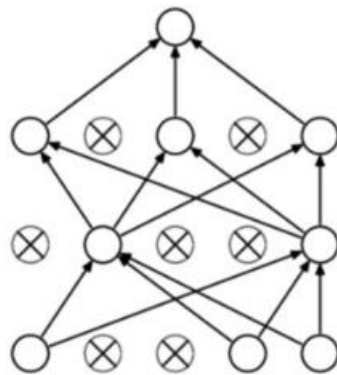


[https://colab.research.google.com/drive/1yw9rwtDNo45mJ1XcVT\\_8W2hG--9Gt2vt](https://colab.research.google.com/drive/1yw9rwtDNo45mJ1XcVT_8W2hG--9Gt2vt)

Drop out : 드롭 아웃은 오버피팅(over-fit)을 막기 위한 방법으로 뉴럴 네트워크가 학습중일때, 랜덤하게 뉴런을 꺼서 학습을 방해함으로써, 학습이 학습용 데이터에 치우치는 현상을 막아준다.



(a) Standard Neural Net



(b) After applying dropout.

Hyperparameters:

f : filter size

$$f=2, s=2$$

s : stride

$$f=3, s=2$$

Max or average pooling

→ p: padding.

$$n_H \times n_W \times \underline{n_C}$$

↓

$$\left\lfloor \frac{n_H - f}{s} + 1 \right\rfloor \times \left\lfloor \frac{n_W - f}{s} + 1 \right\rfloor$$

$$\times \underline{n_C}$$