AI School 6기 5주차

파이썬 기초 – 웹 크롤링

CNN 기초2

CNN 모델을 활용한 객체 분류

AI School 6기 5주차

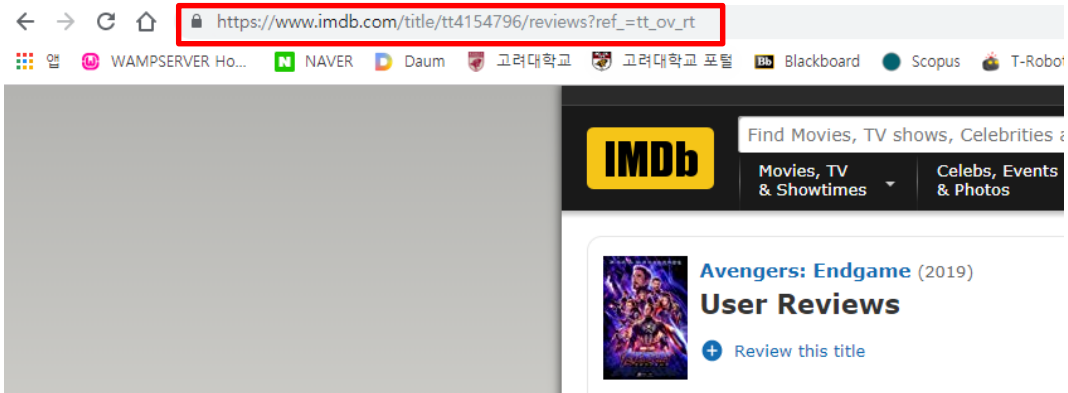
파이썬 기초 – 웹 크롤링

웹 크롤링

- BeuatifulSoup

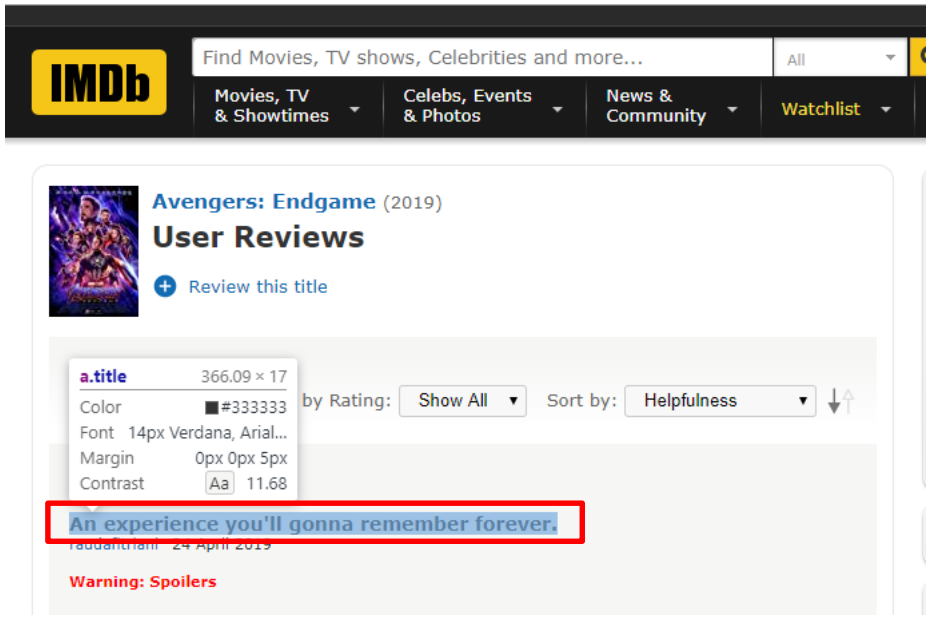
Package	Version	Latest
Markdown	2.6.8	3.1.1
PyMySQL	0.9.3	0.9.3
Werkzeug	0.12.2	0.15.4
backports.weakref	1.0rc1	1.0.post1
beautifulsoup4	4.7.1	4.7.1
bleach	1.5.0	3.1.0
bs4	0.0.1	0.0.1
html5lib	0.9999999	1.0.1
lxml	4.3.3	4.3.3

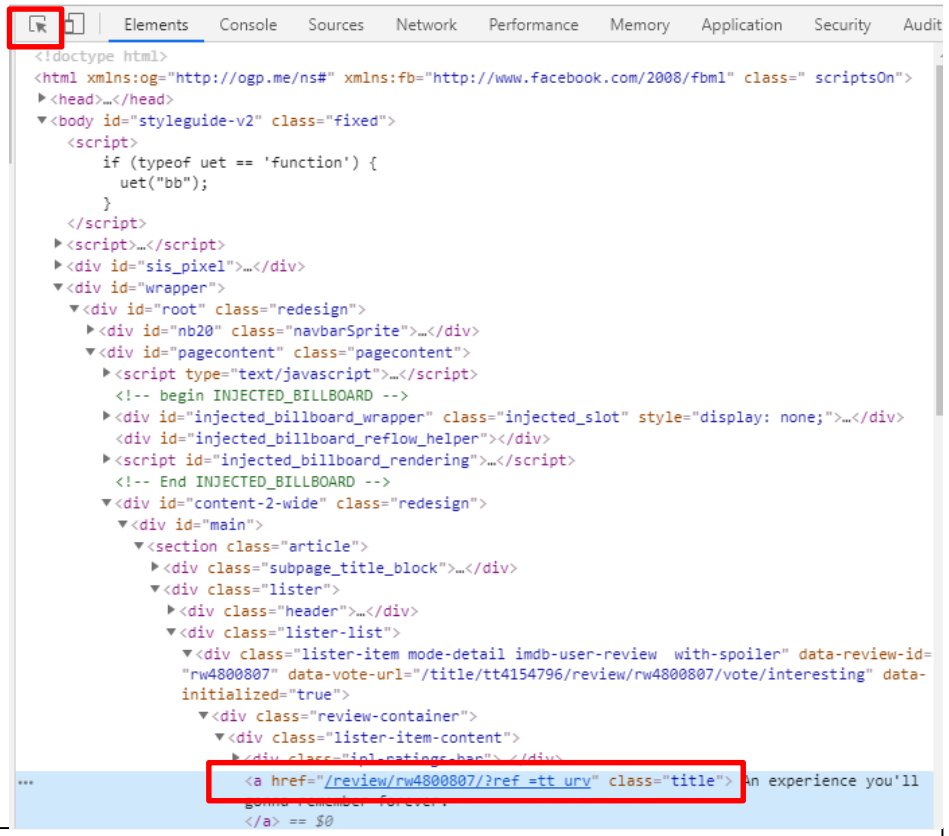
```
from bs4 import BeautifulSoup
import urllib.request

url = "https://www.imdb.com/title/tt4154796/reviews?ref_=tt_ov_rt"
htmlData = urllib.request.urlopen(url)
bs = BeautifulSoup(htmlData, 'lxml')
print(BeautifulSoup.prettify(bs))
```



웹 크롤링

- 개발자 도구 (F12)

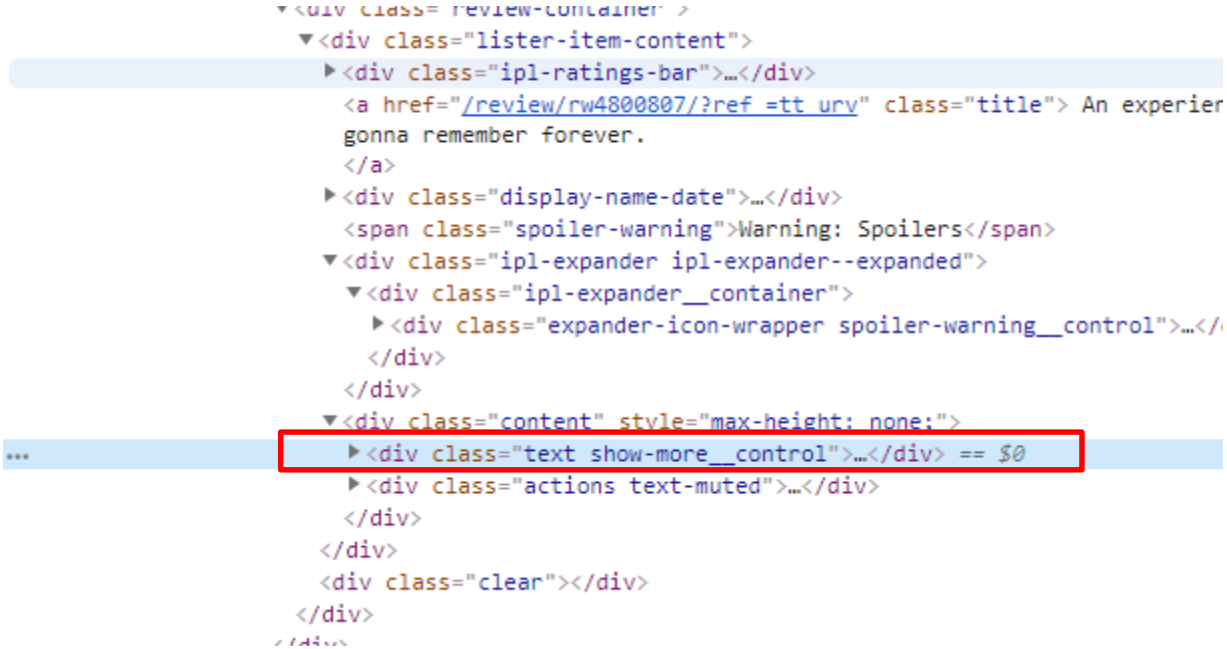





```
title_list = bs.findAll('a', 'title')

for title in title_list:
    print(title.getText())
```

웹 크롤링

- bs.findAll([tag], [class명])




```
review_list = bs.findAll('div', 'text show-more__control')

for content in review_list:
    print(content.getText()+"₩n")
```

웹 크롤링

- bs.findAll([tag], [class명])


```
score_list = bs.findAll('span', 'rating-other-user-rating')

for score in score_list:
    print(score.span.getText())
```

전처리

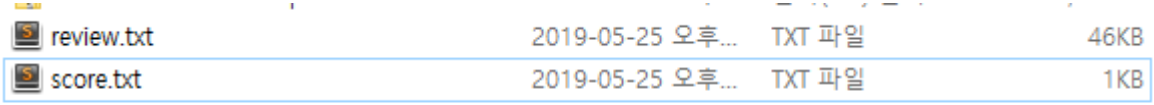
- 알파벳 외 문자 제거, 특수 문자 등 분리, 소문자 변환

```
import re

def clean_str(string):
    string = re.sub(r"[^A-Za-z0-9(),!?₩'₩`]", " ", string)
    string = re.sub(r"₩'s", " ₩'s", string)
    string = re.sub(r"₩'ve", " ₩'ve", string)
    string = re.sub(r"n₩'t", " n₩'t", string)
    string = re.sub(r"₩'re", " ₩'re", string)
    string = re.sub(r"₩'d", " ₩'d", string)
    string = re.sub(r"₩'ll", " ₩'ll", string)
    string = re.sub(r",", " , ", string)
    string = re.sub(r"!", " ! ", string)
    string = re.sub(r"₩(", " ₩( ", string)
    string = re.sub(r"₩)", " ₩) ", string)
    string = re.sub(r"₩?", " ₩? ", string)
    string = re.sub(r"₩s{2,}", " ", string)
    return string.strip().lower()
```

File 출력

- f = open("파일 경로", "읽기/쓰기")

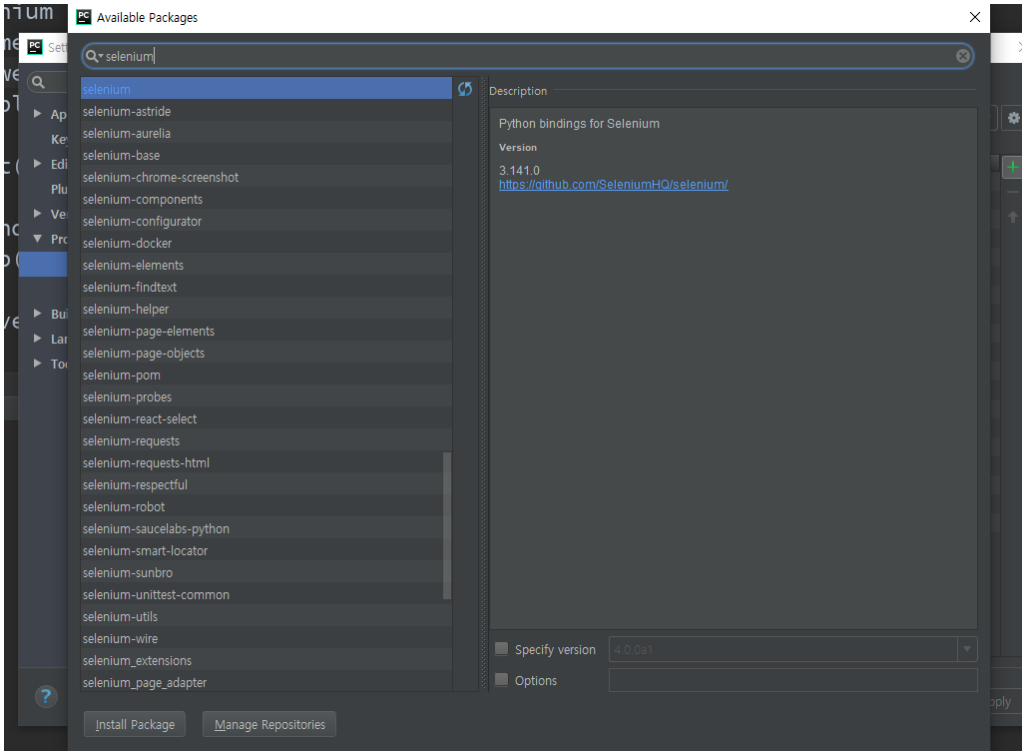
```
print(len(title_list))
f = open("./data/review.txt", "w", encoding='UTF8')
for i in range(len(title_list)):
    f.write(clean_str(title_list[i].getText())+" "+clean_str(review_list[i].getText())+"₩n")
f.close()

f = open("./data/score.txt", "w", encoding='UTF8')
for i in range(len(score_list)):
    f.write(score_list[i].span.getText()+"₩n")
f.close()
```



동적 크롤링

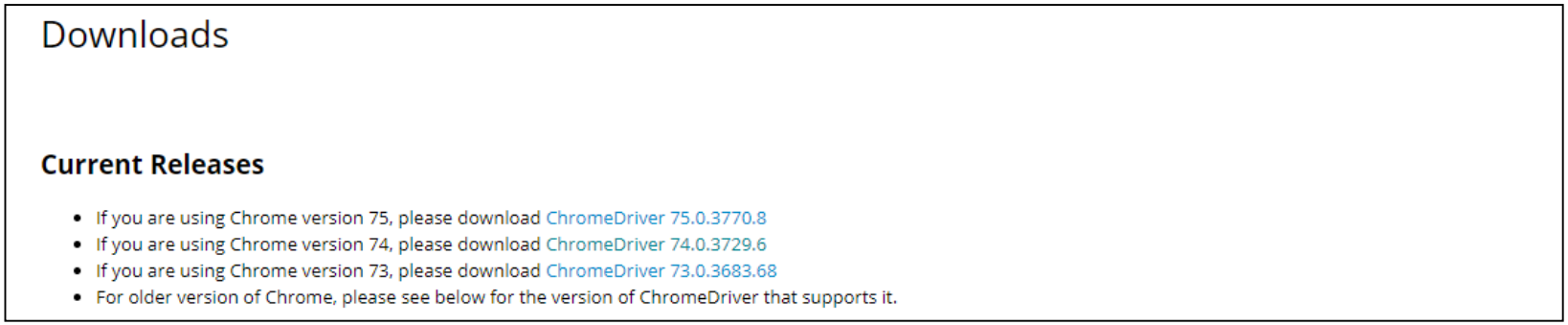
- Selenium
  - 웹앱을 테스트하는데 이용하는 프레임 워크.
  - Webdriver라는 API를 통해 운영체제에 설치된 Chrome등의 브라우저를 제어

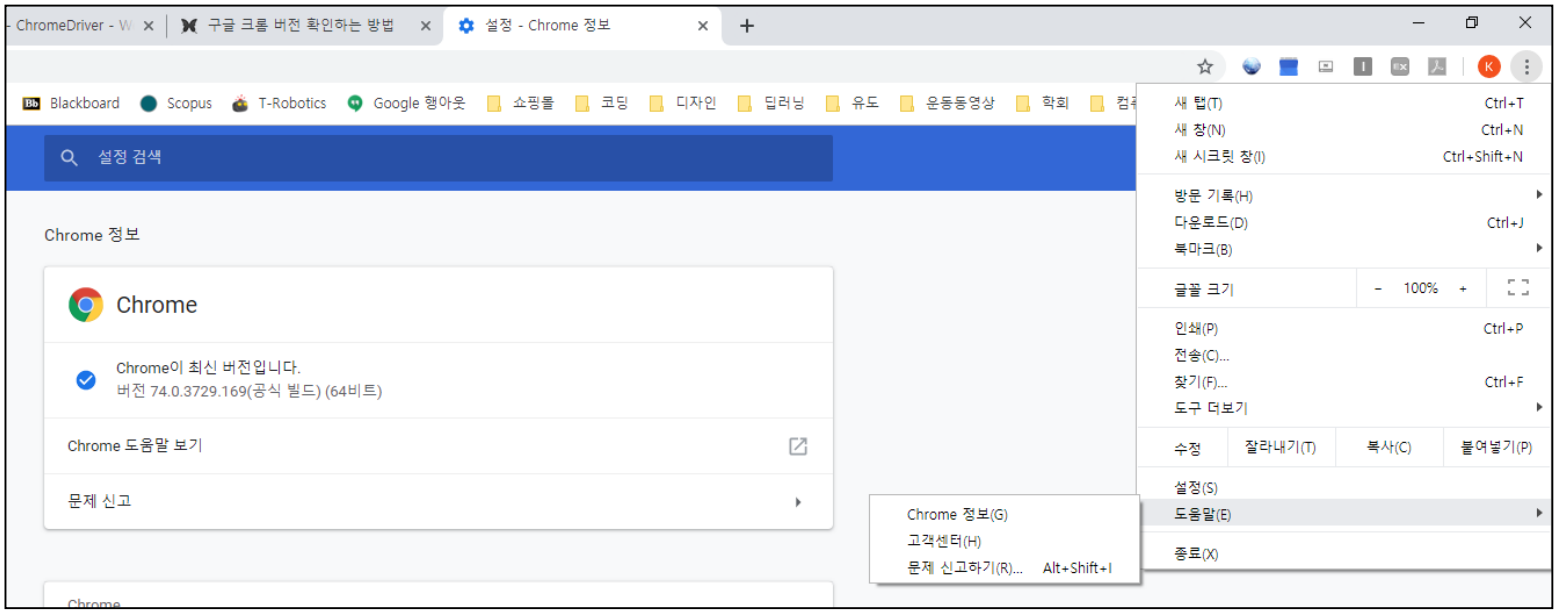
```
pip install selenium
```



동적 크롤링

- Chrome WebDriver
  - https://sites.google.com/a/chromium.org/chromedriver/downloads
  - 본인의 Chrome 버전 확인 후 버전에 맞는 webdriver 다운로드



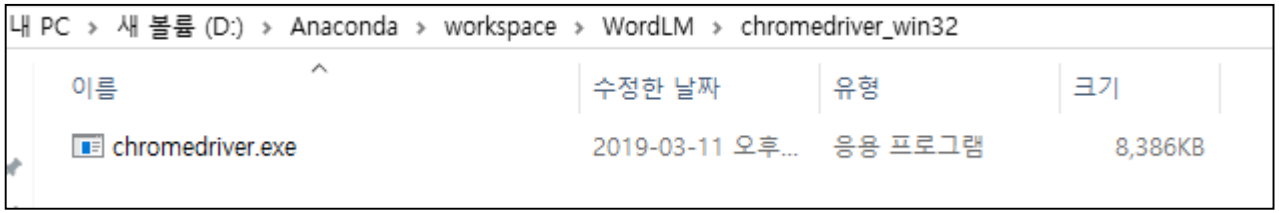


동적 크롤링

- Chrome WebDriver
  - 운영체제에 맞는 webdriver 다운로드
  - 원하는 위치에 압축 풀기

Index of /74.0.3729.6/

Name	Last modified	Size	ETag
Parent Directory		-	
chromedriver_linux64.zip	2019-03-12 19:25:26	4.83MB	3cd9e67808926bfba9a3f5946e2a994d
chromedriver_mac64.zip	2019-03-12 19:25:27	6.69MB	de2aa78283af413100cddc2a4dee3ebc
chromedriver_win32.zip	2019-03-12 19:25:29	4.41MB	9780b9b586e74253df9b58928b959861
notes.txt	2019-03-14 18:17:49	0.00MB	d6180d1b525cf857b030077a525a9f47



동적 크롤링

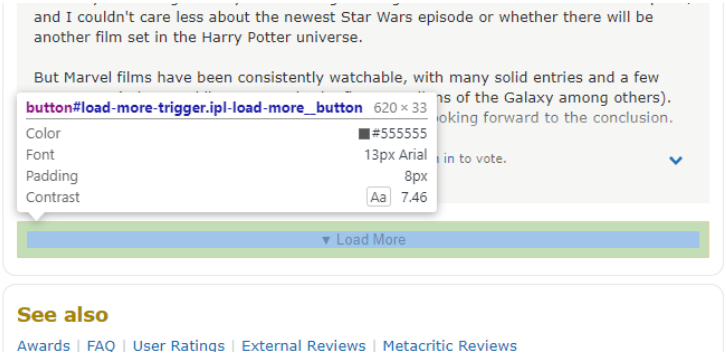
- Chrome WebDriver
  - https://sites.google.com/a/chromium.org/chromedriver/downloads
  - 본인의 Chrome 버전 확인 후 버전에 맞는 webdriver 다운로드

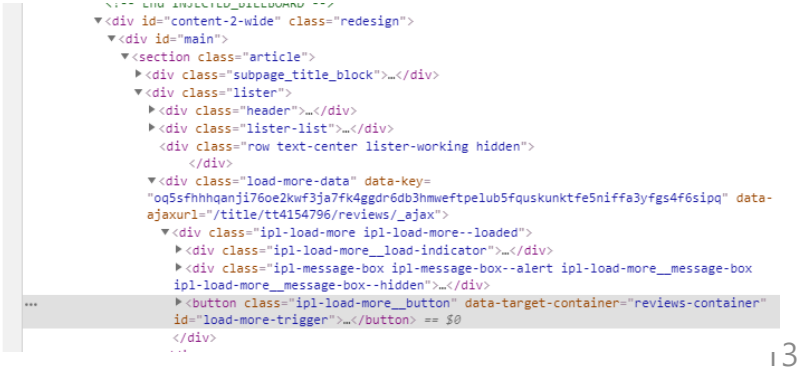
```
import selenium
from selenium import webdriver
import time
driver = webdriver.Chrome("D:/Anaconda/workspace/WordLM/chromedriver_win32/chromedriver")
driver.implicitly_wait(3)

driver.get('https://www.imdb.com/title/tt4154796/reviews?ref_=tt_ov_rt')
```

동적 크롤링

- 버튼 클릭 및 html source 받아오기

```
driver.find_element_by_xpath('//*[@id="load-more-trigger"]').click()
time.sleep(10)

click_list = driver.find_elements_by_xpath("//div[@class='expander-icon-wrapper show-more__control']")
for click in click_list:
    if click.is_displayed():
        click.click()

req = driver.page_source

bs=BeautifulSoup(req, 'lxml')
```





동적 크롤링

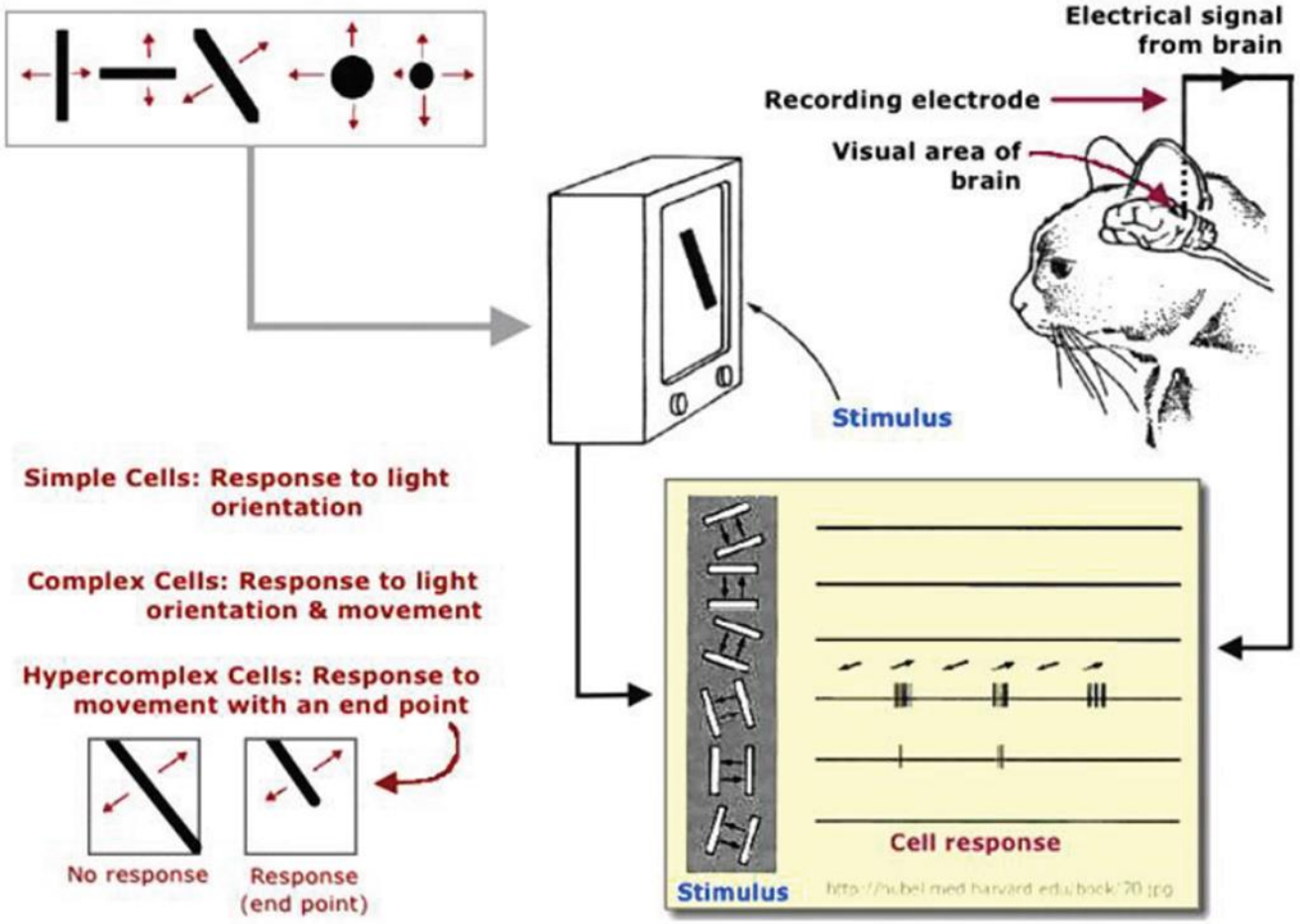
- 뒤 부분 동일

```
title_list = bs.findAll('a', 'title')
review_list = bs.findAll('div', 'text show-more__control')
score_list = bs.findAll('span', 'rating-other-user-rating')
.
.
.
f = open("./data/score.txt", "w", encoding='UTF8')
for i in range(len(score_list)):
    f.write(score_list[i].span.getText()+"₩n")
f.close()
```

AI School 5기 5주차

CNN 기초2

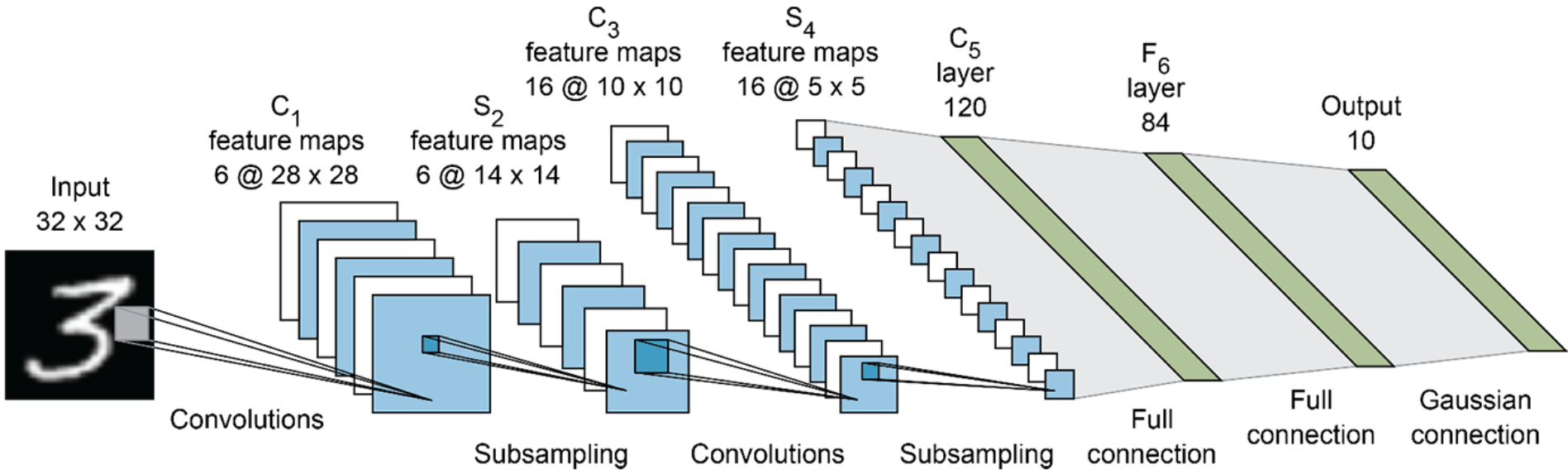
Convolutional Neural Networks



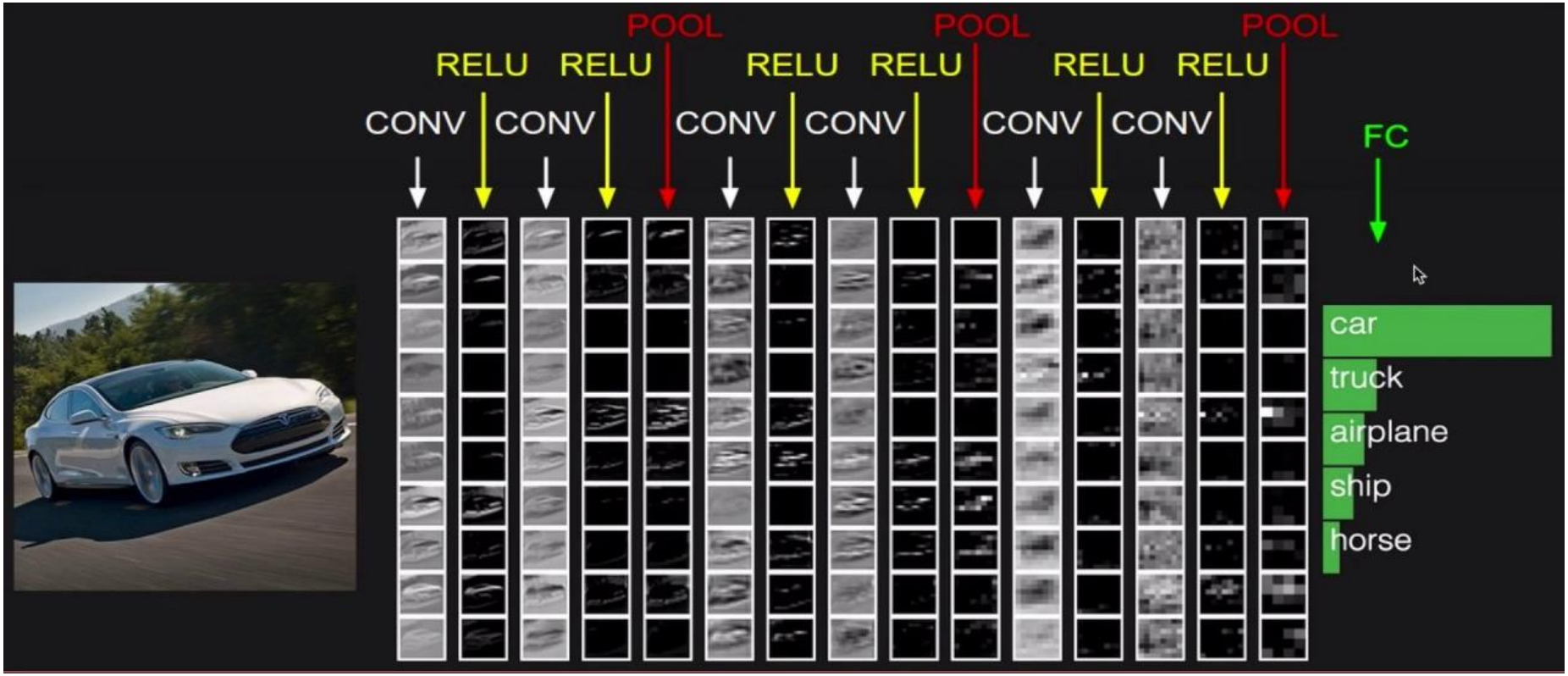
Hubel & Wiesel, 1959

Convolutional Neural Networks

- Convolution과 Pooling을 반복하여 상위 Feature를 구성
- Convolution은 Local영역에서의 특정 Feature를 얻는 과정
- Pooling은 Dimension을 줄이면서도, Translation-invariant 한 Feature를 얻는 과정

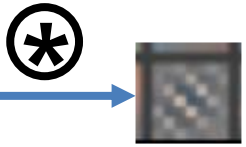


Convolutional Neural Networks



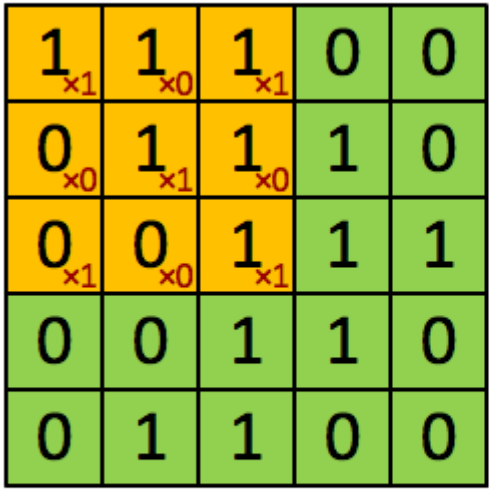
Convolution, Filter

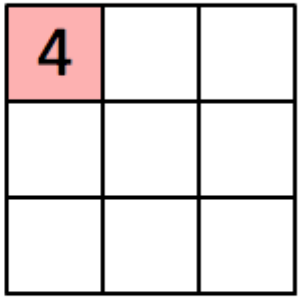
Filter 1

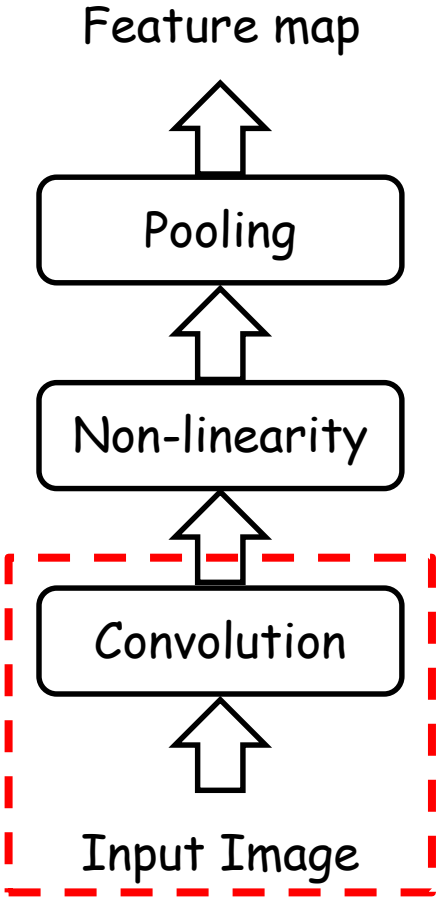


Filter 2

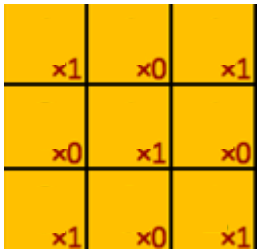
Convolution, Filter



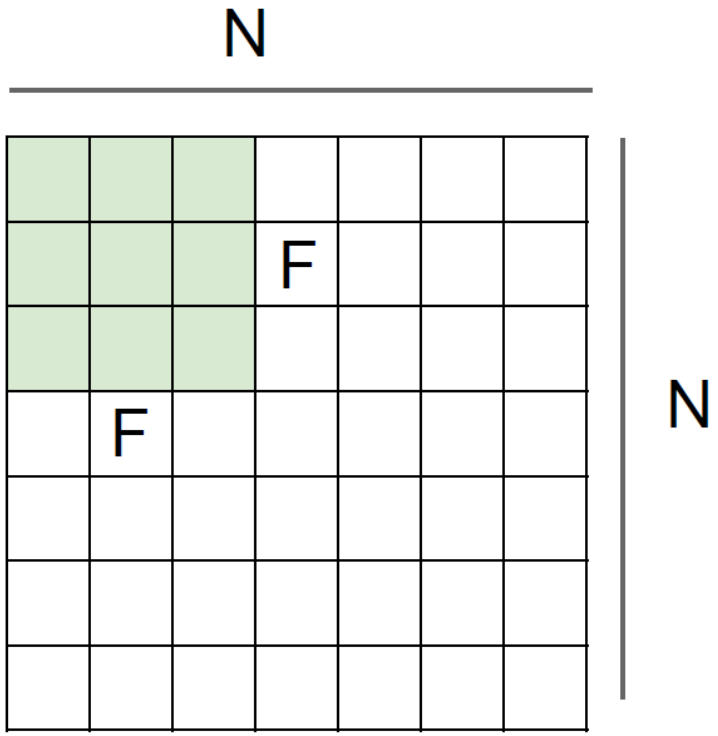




*filter



Convolution, Filter

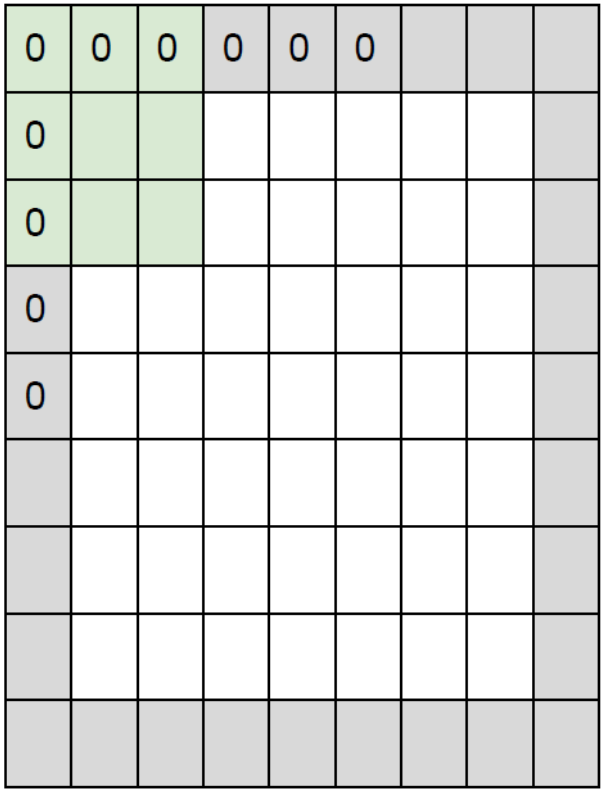


Output size:
(N - F) / stride + 1

e.g. N = 7, F = 3:
stride 1 => (7 - 3)/1 + 1 = 5
stride 2 => (7 - 3)/2 + 1 = 3
stride 3 => (7 - 3)/3 + 1 = 2.33 :\

Convolution, Filter

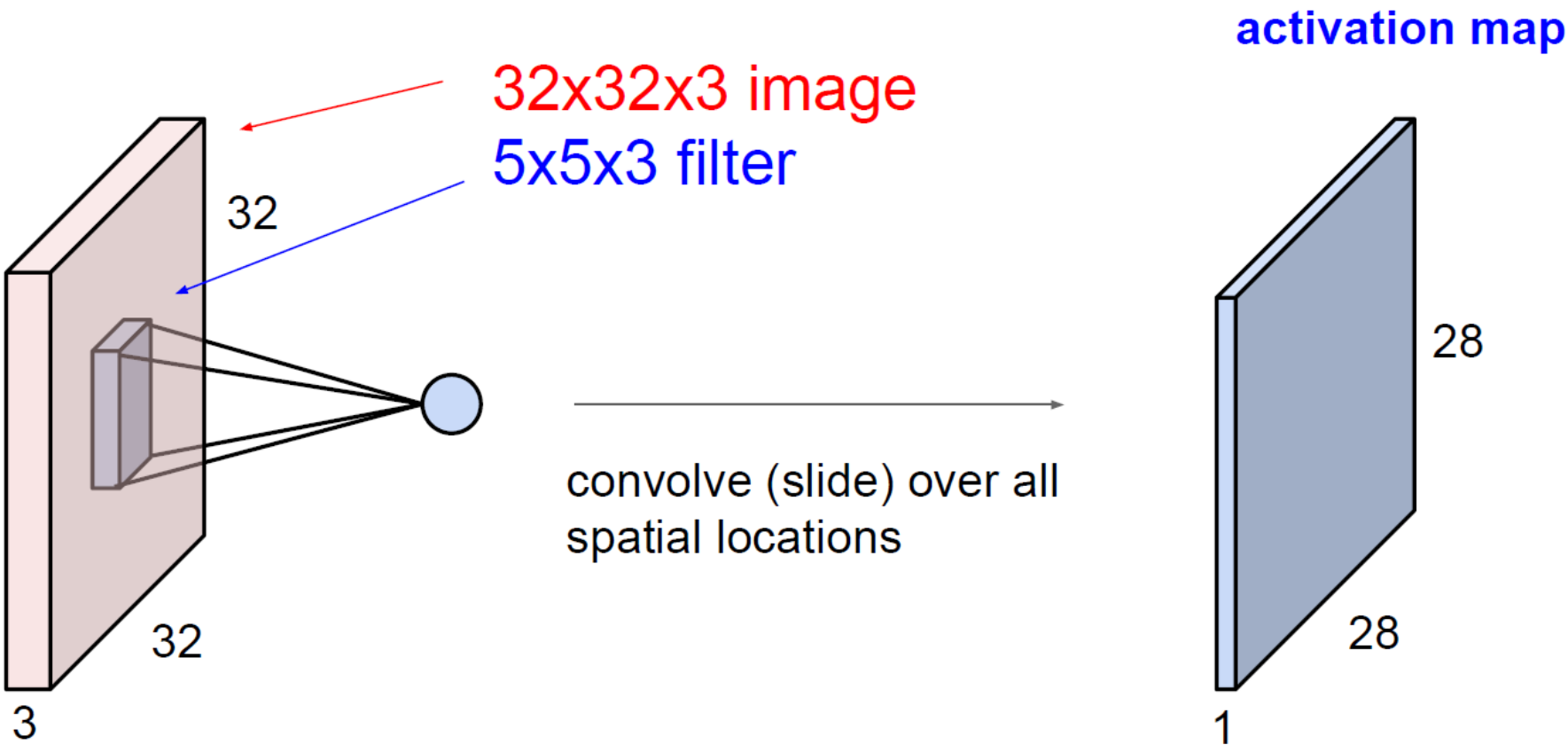
In practice: Common to zero pad the border



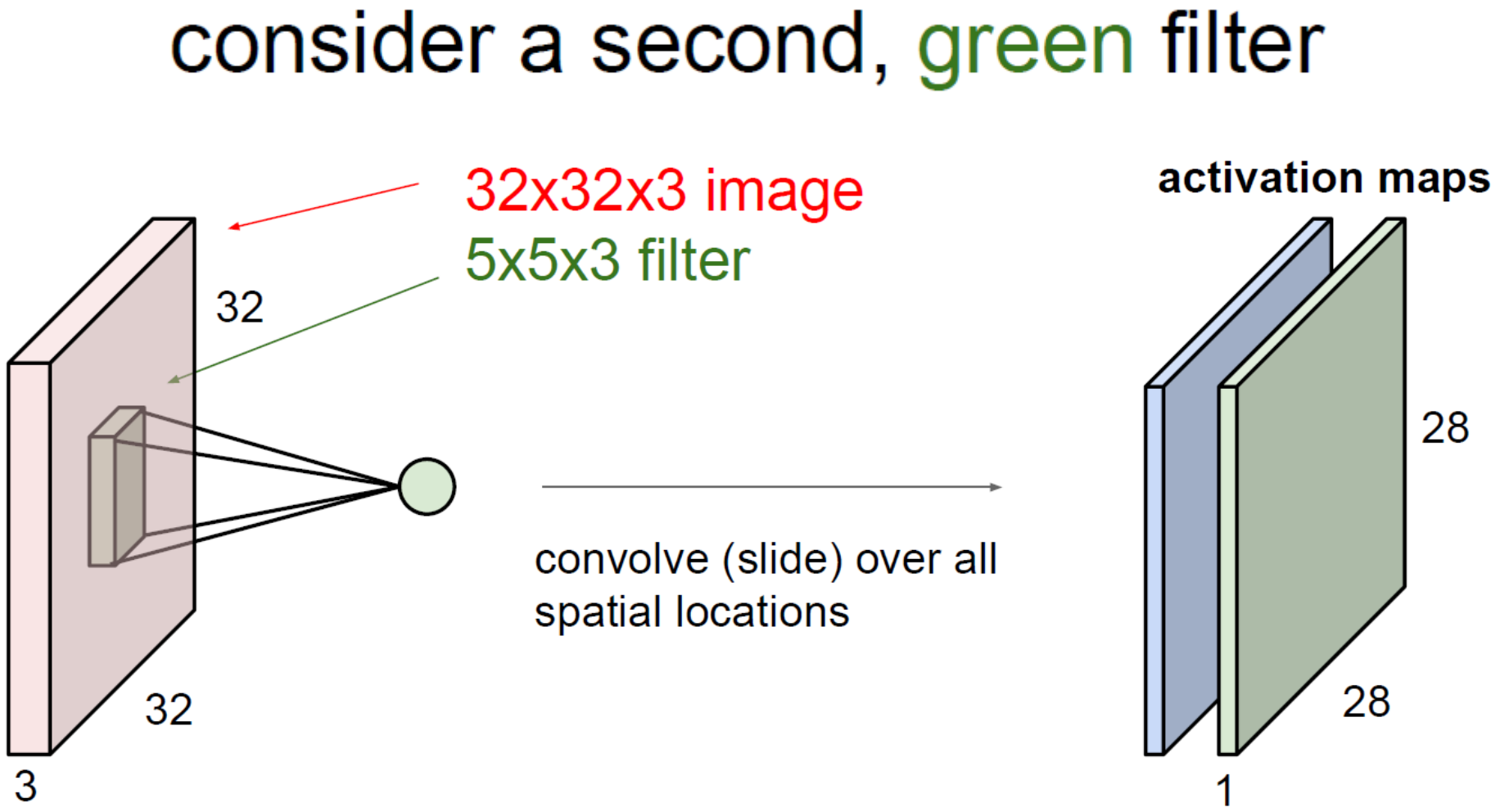
e.g. input 7x7
3x3 filter, applied with **stride 1**
pad with 1 pixel border => what is the output?

(recall:)
(N - F) / stride + 1

Convolution, Filter

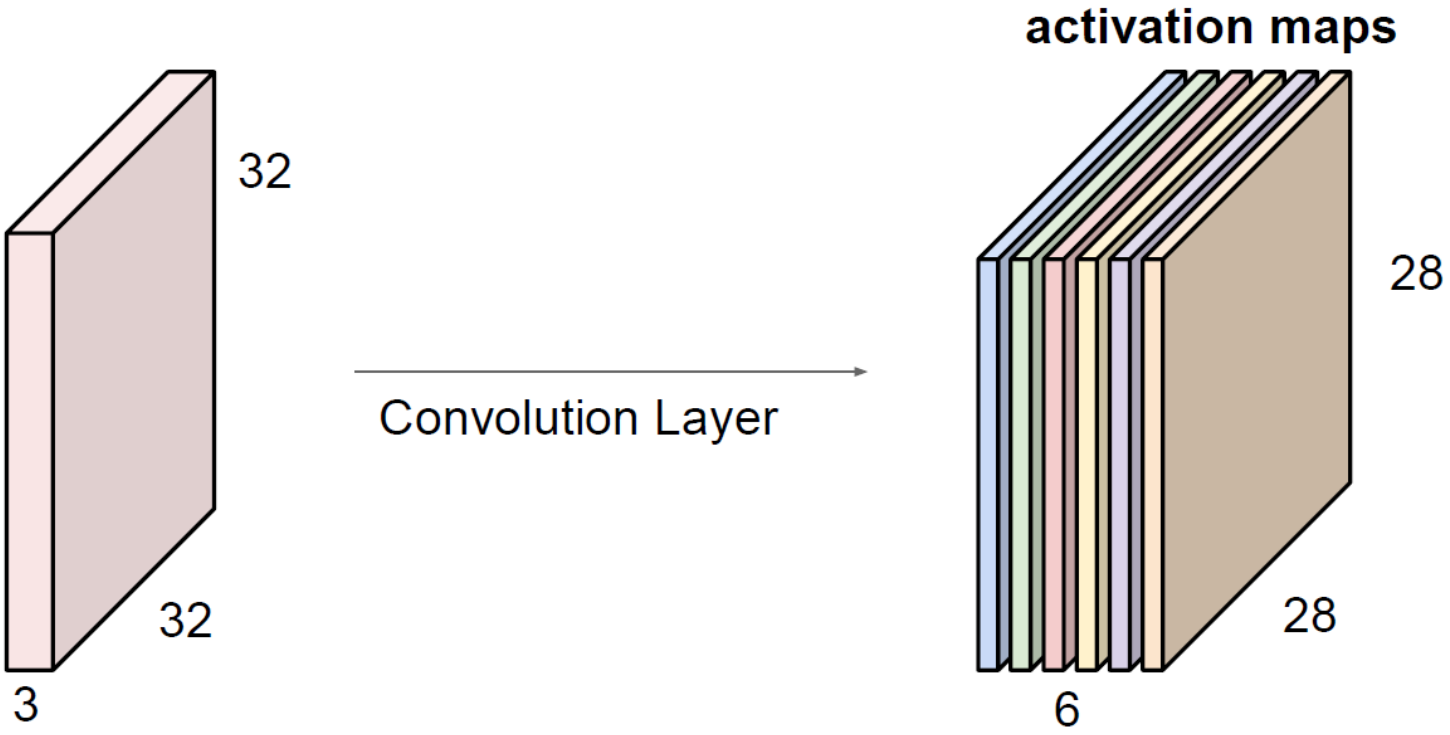


Convolution, Filter



Convolution, Filter

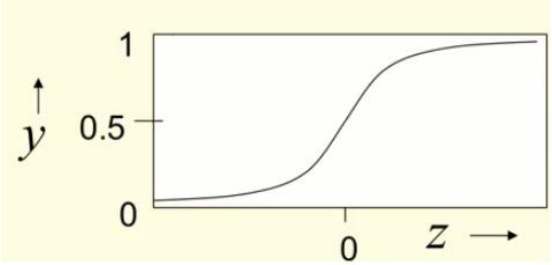
For example, if we had 6 5x5 filters, we'll get 6 separate activation maps:



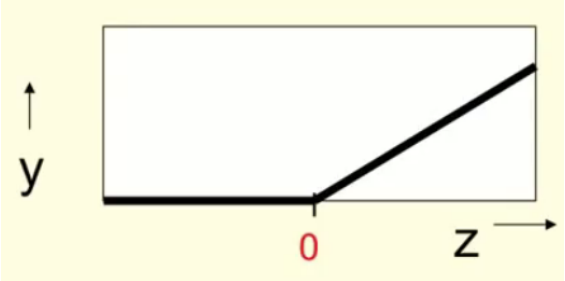
We stack these up to get a "new image" of size 28x28x6!

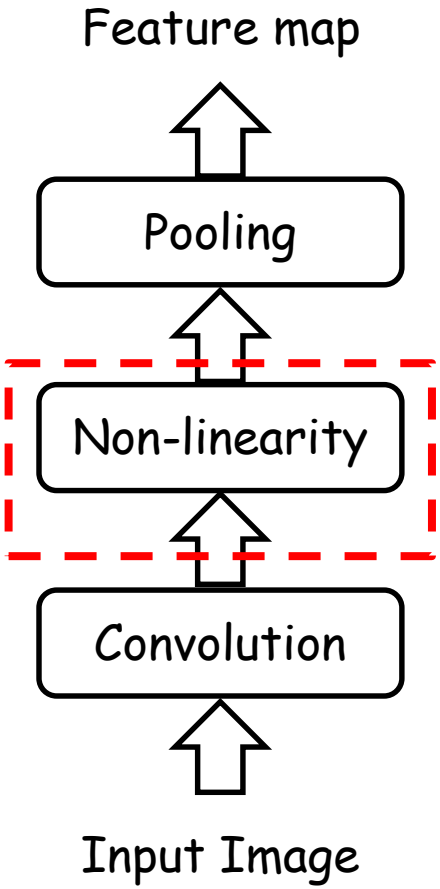
Non-linearity

Sigmoid

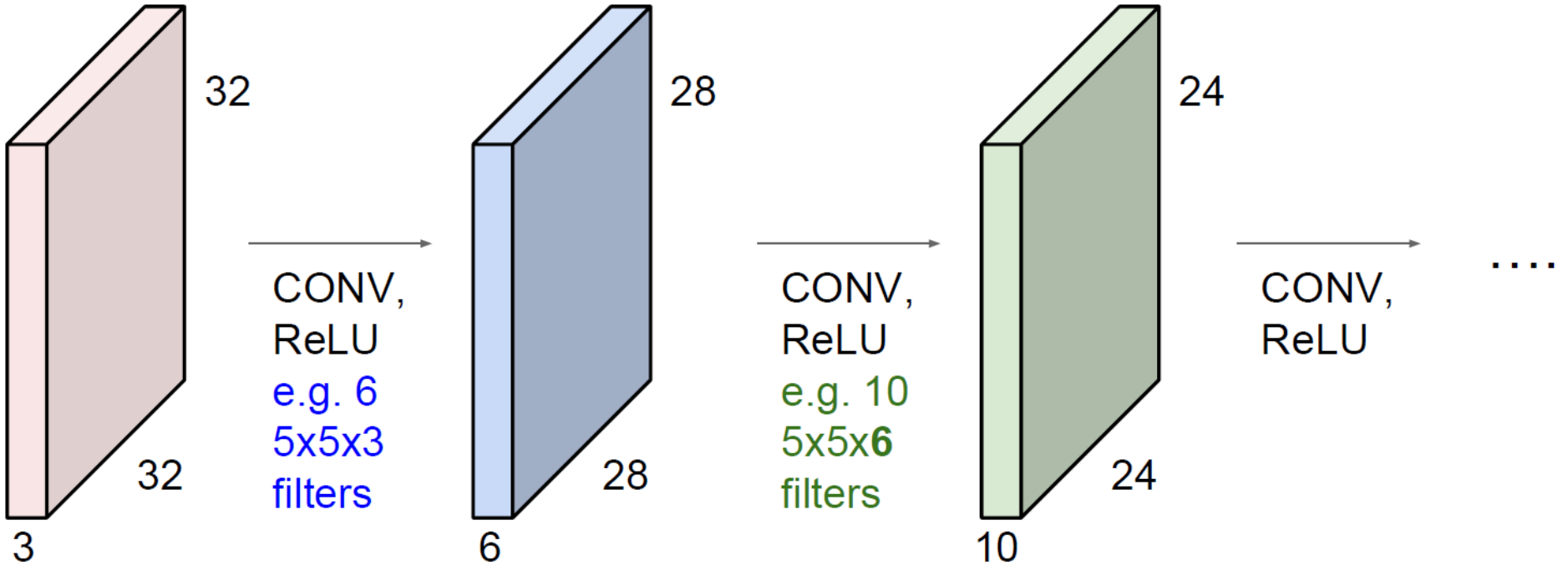


Rectified linear unit



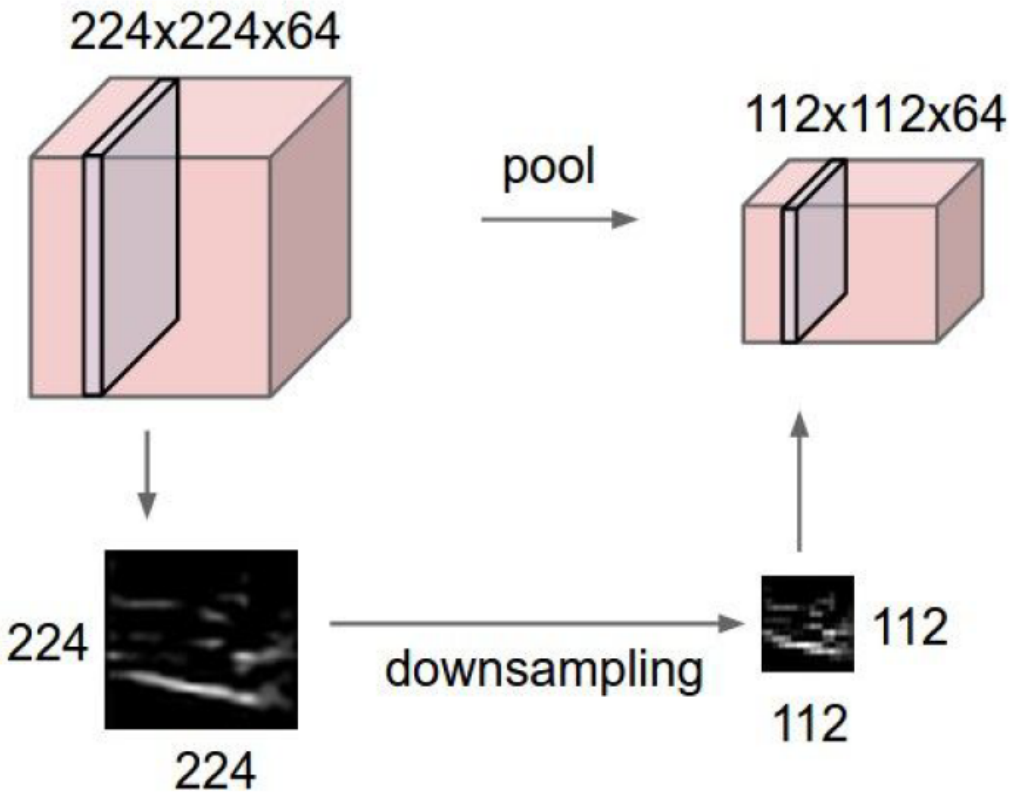


Convolution, Filter

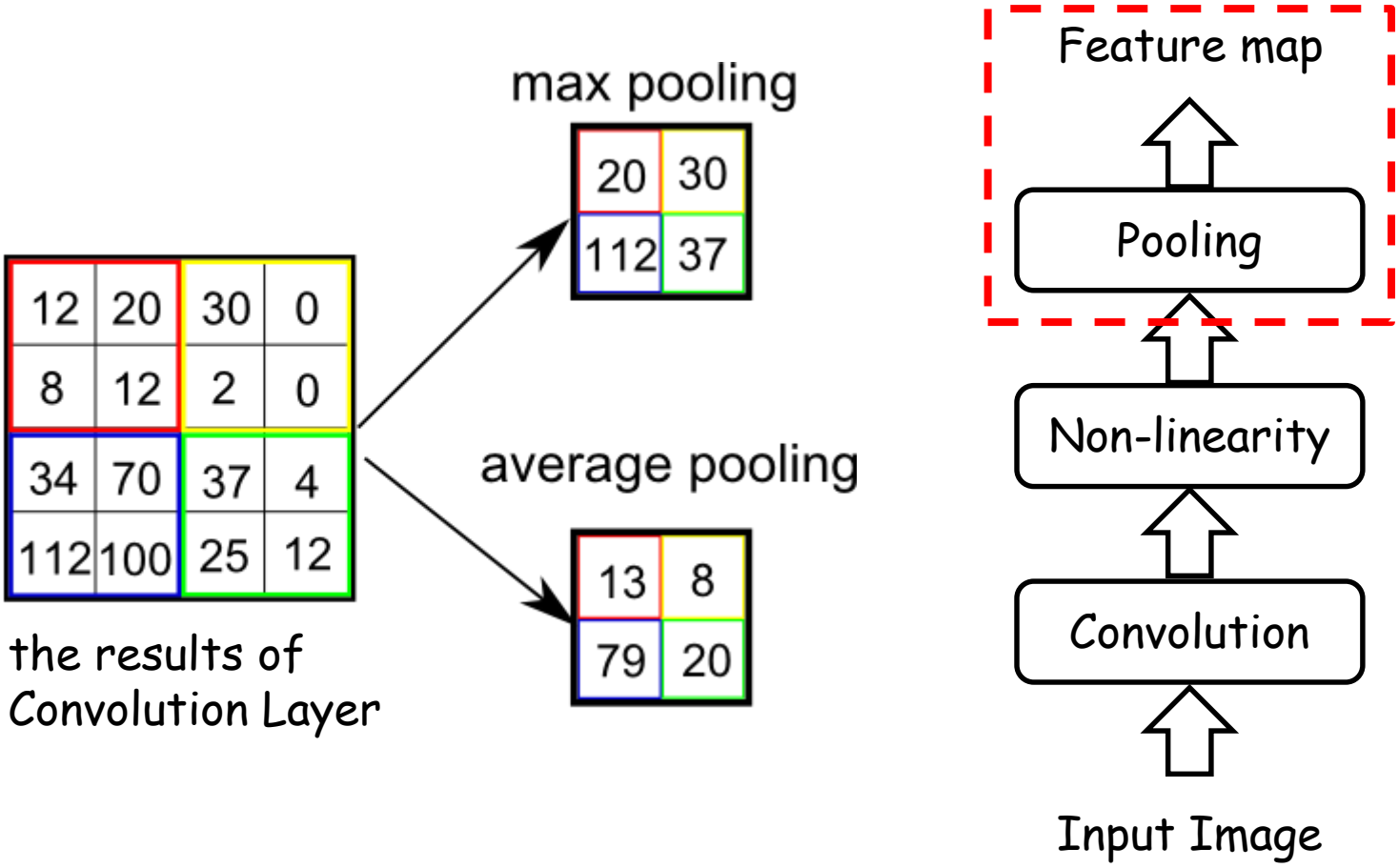


Pooling

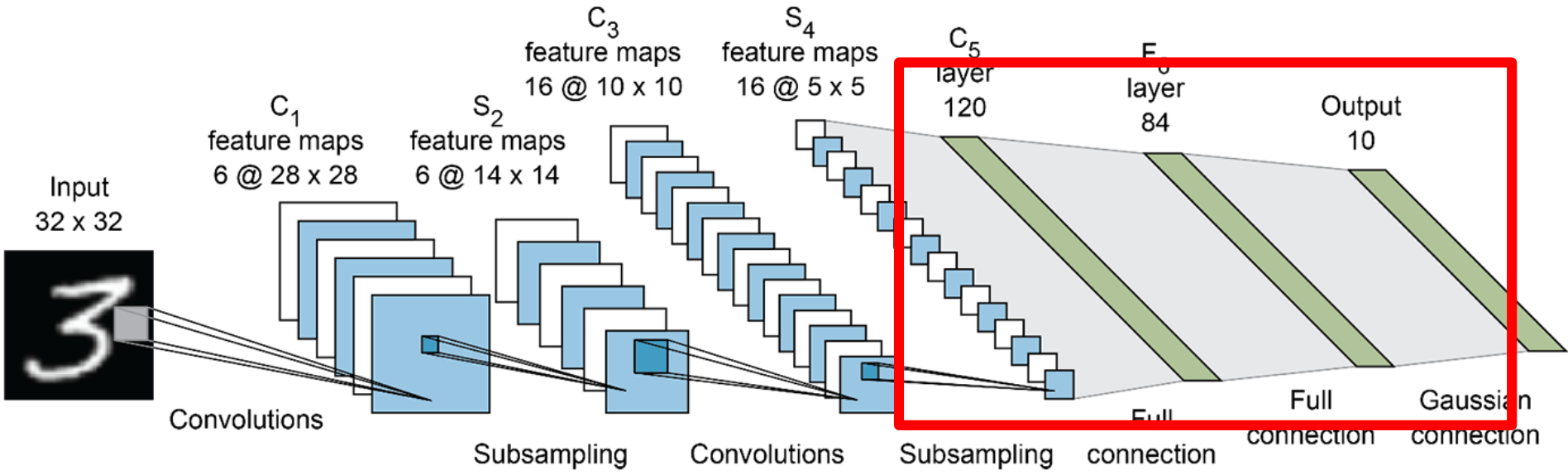
- makes the representations smaller and more manageable
- operates over each activation map independently:



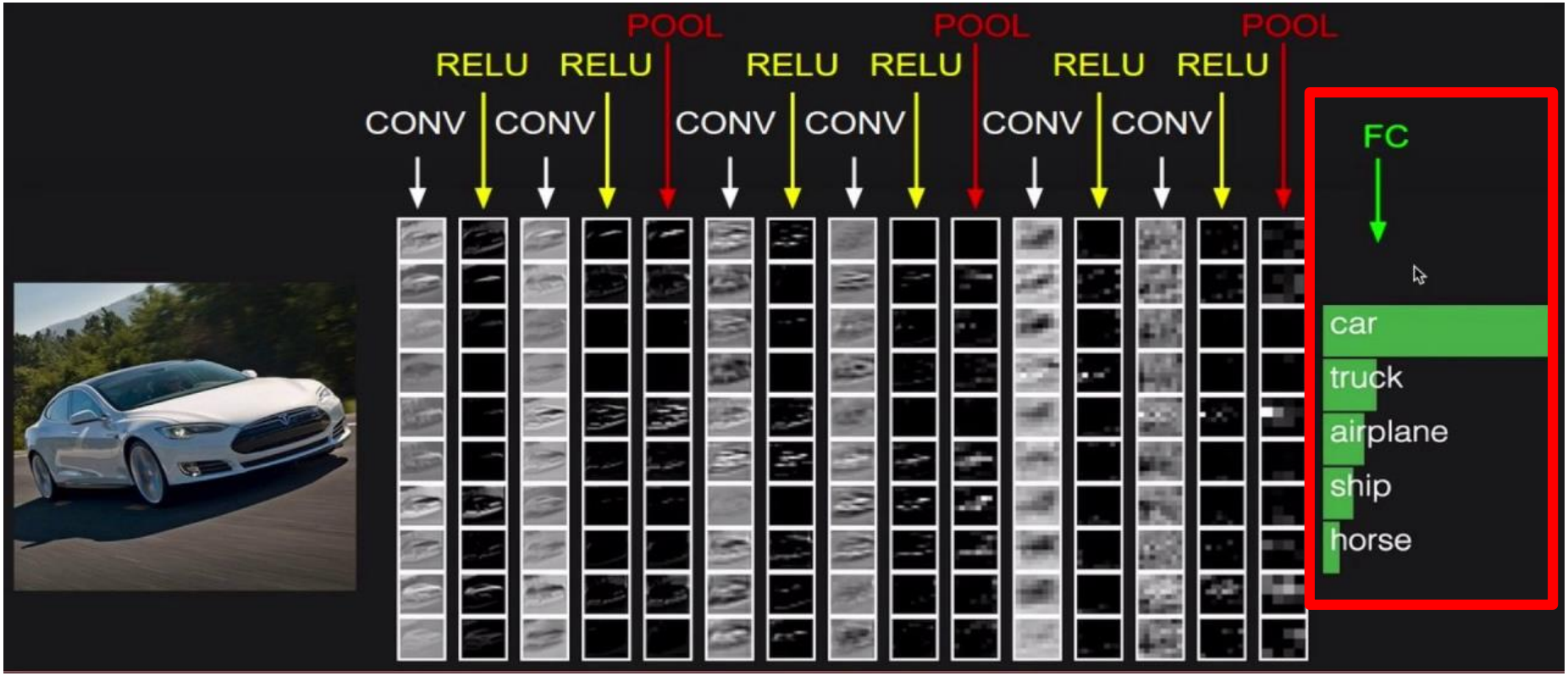
Pooling


max pooling
20 30
112 37
12 20 30 0
8 12 2 0
34 70 37 4
112 100 25 12
average pooling
13 8
79 20
the results of
Convolution Layer
Feature map
Pooling
Non-linearity
Convolution
Input Image

Fully connected layer

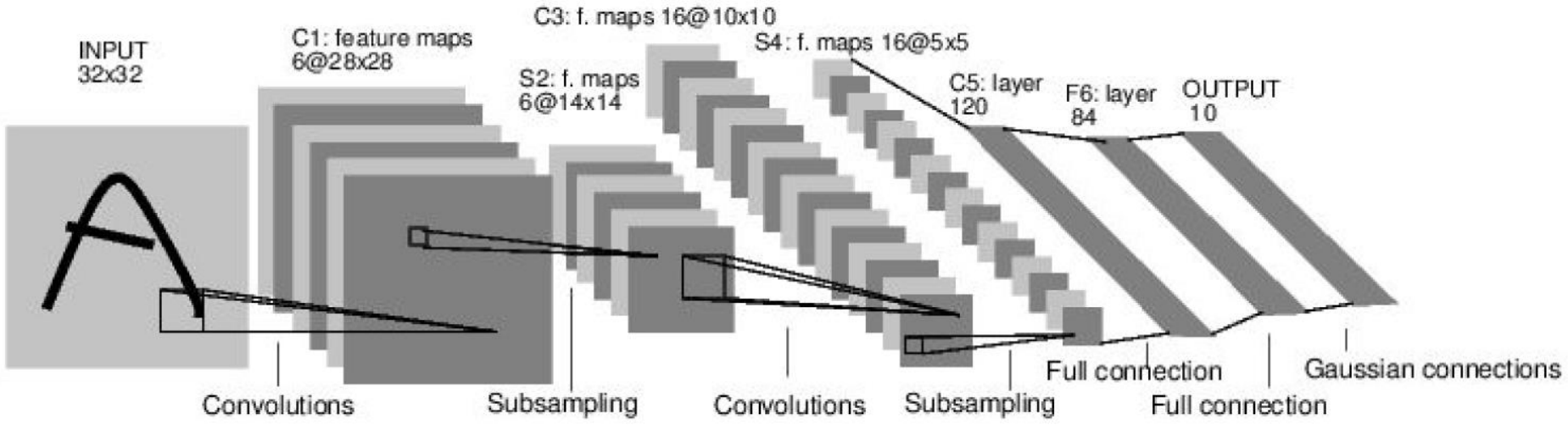


Fully connected layer



LeNet-5

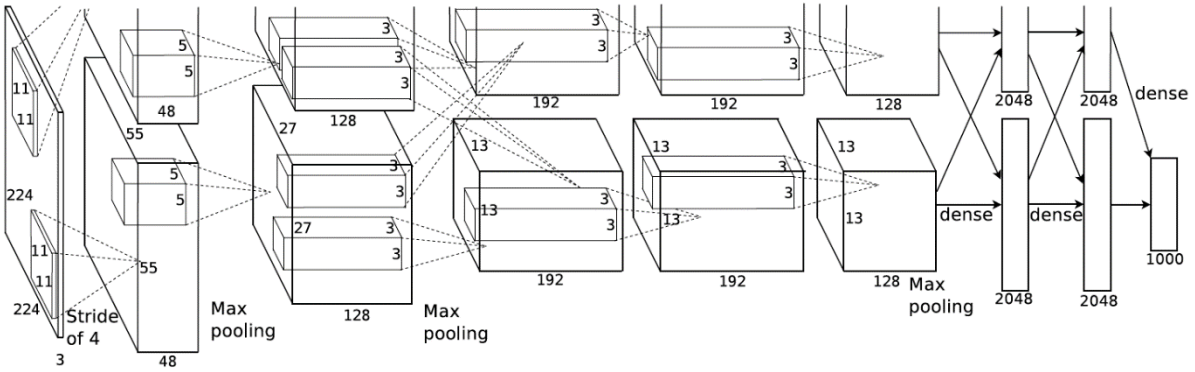
- LeNet [LeCun et al., 1998]



Conv filters were 5x5, applied at stride 1
Subsampling (Pooling) layers were 2x2 applied at stride 2
i.e. architecture is [CONV-POOL-CONV-POOL-CONV-FC]

CNN for object recognition

- AlexNet [Krizhevsky et al., 2012]

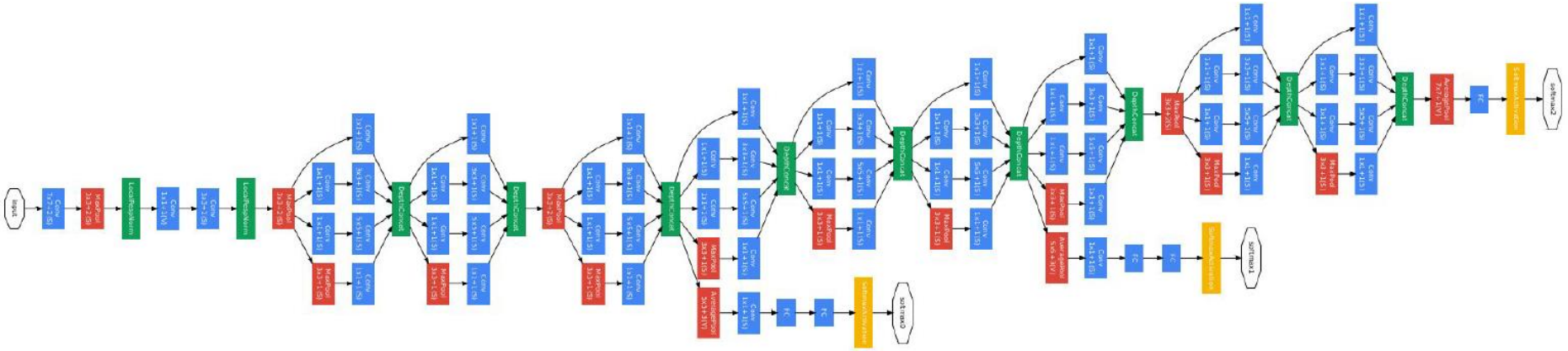


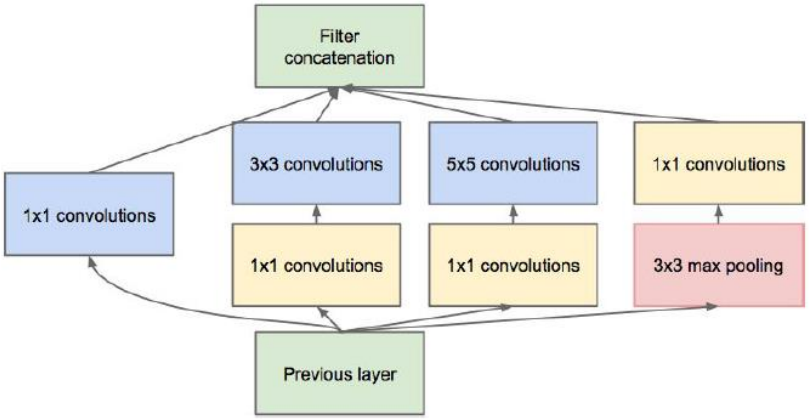
Full (simplified) AlexNet architecture:
[227x227x3] INPUT
[55x55x96] CONV1: 96 11x11 filters at stride 4, pad 0
[27x27x96] MAX POOL1: 3x3 filters at stride 2
[27x27x96] NORM1: Normalization layer
[27x27x256] CONV2: 256 5x5 filters at stride 1, pad 2
[13x13x256] MAX POOL2: 3x3 filters at stride 2
[13x13x256] NORM2: Normalization layer
[13x13x384] CONV3: 384 3x3 filters at stride 1, pad 1
[13x13x384] CONV4: 384 3x3 filters at stride 1, pad 1
[13x13x256] CONV5: 256 3x3 filters at stride 1, pad 1
[6x6x256] MAX POOL3: 3x3 filters at stride 2
[4096] FC6: 4096 neurons
[4096] FC7: 4096 neurons
[1000] FC8: 1000 neurons (class scores)

Details/Retrospectives:
- first use of ReLU
- used Norm layers (not common anymore)
- heavy data augmentation
- dropout 0.5
- batch size 128
- SGD Momentum 0.9
- Learning rate 1e-2, reduced by 10 manually when val accuracy plateaus
- L2 weight decay 5e-4
- 7 CNN ensemble: 18.2% -> 15.4%

CNN for object recognition

- GoogleNet [Szegedy et al., 2014]





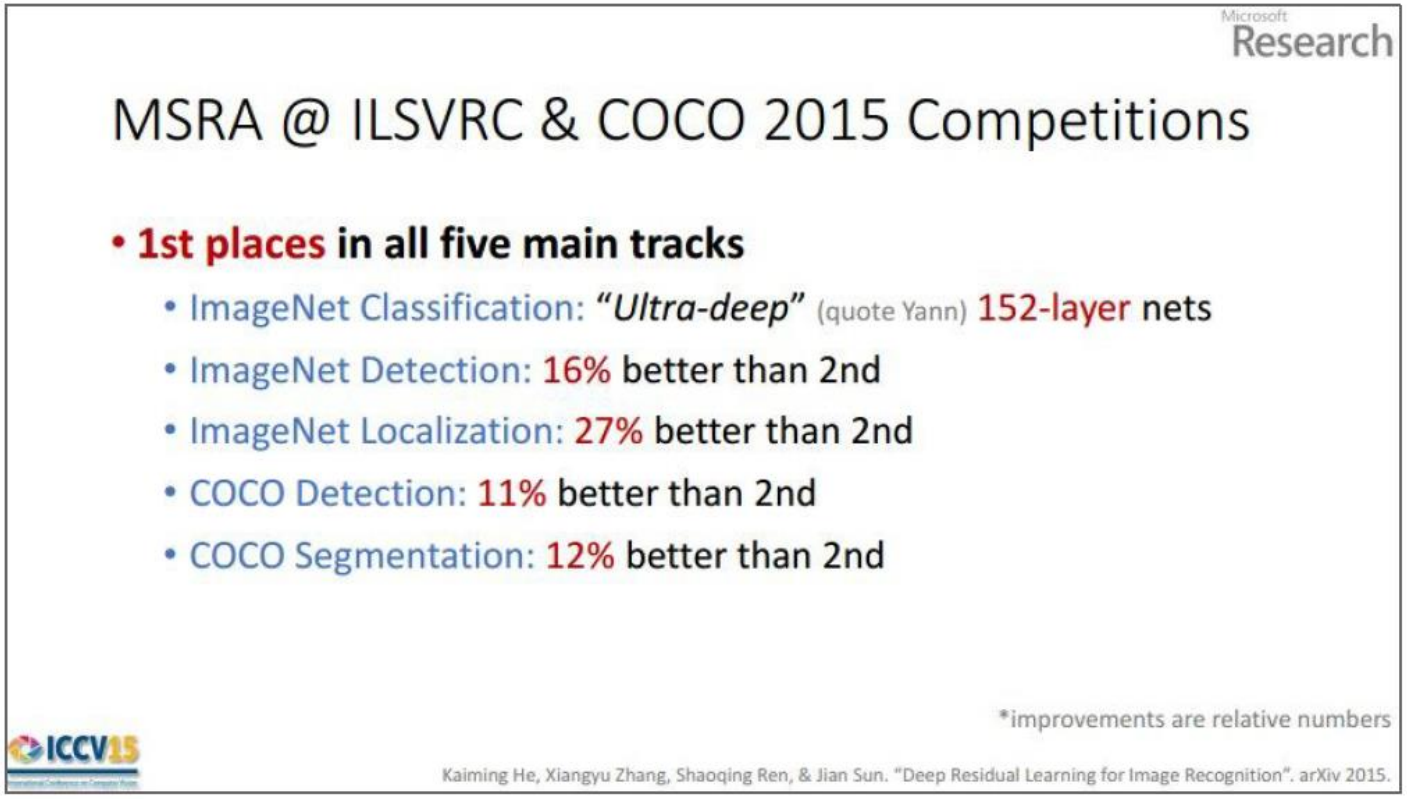
Inception module

ILSVRC 2014 winner (6.7% top 5 error)

CNN for object recognition

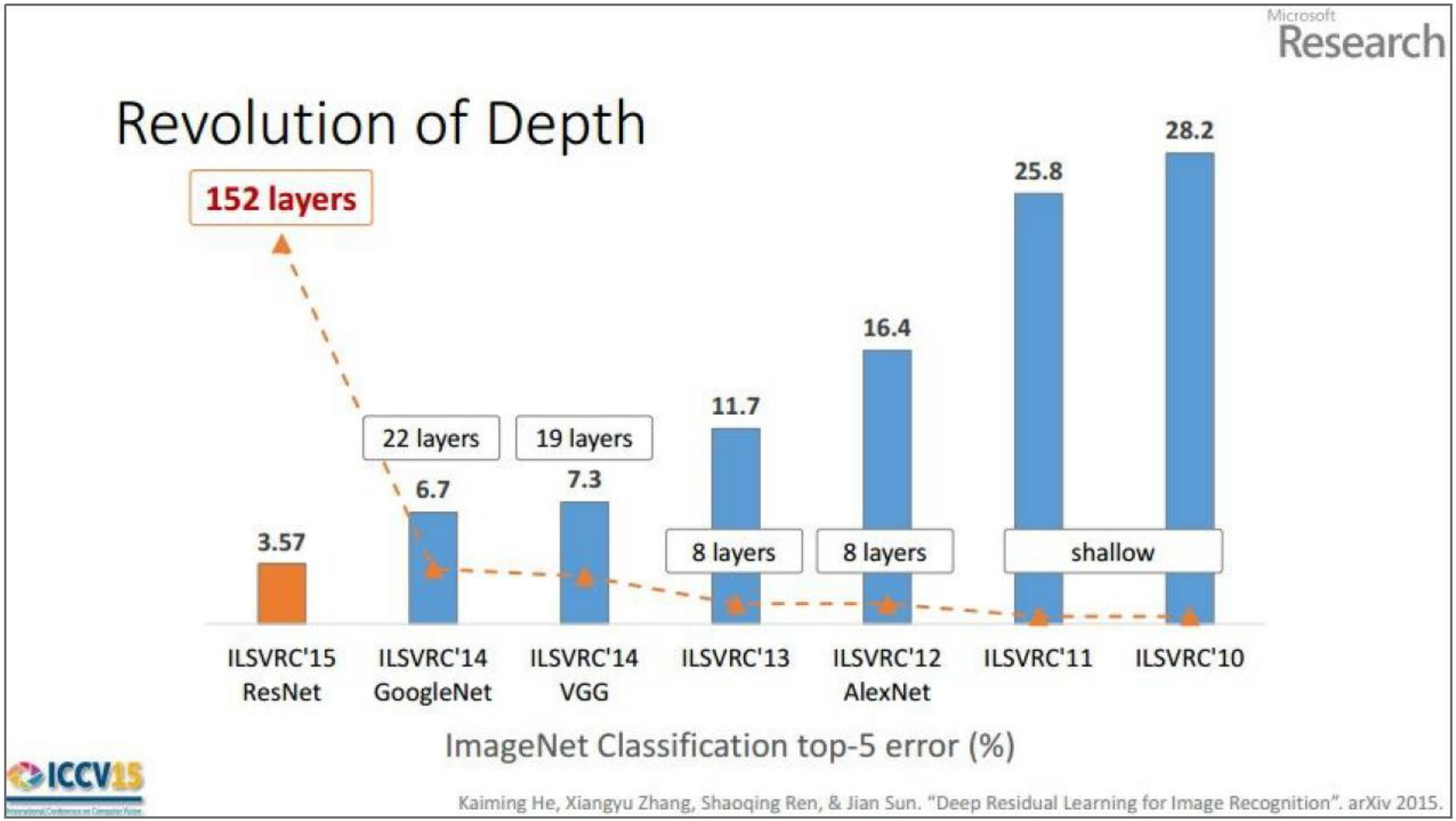
- ResNet [He et al., 2015]

ILSVRC 2015 winner (3.6% top 5 error)



CNN for object recognition

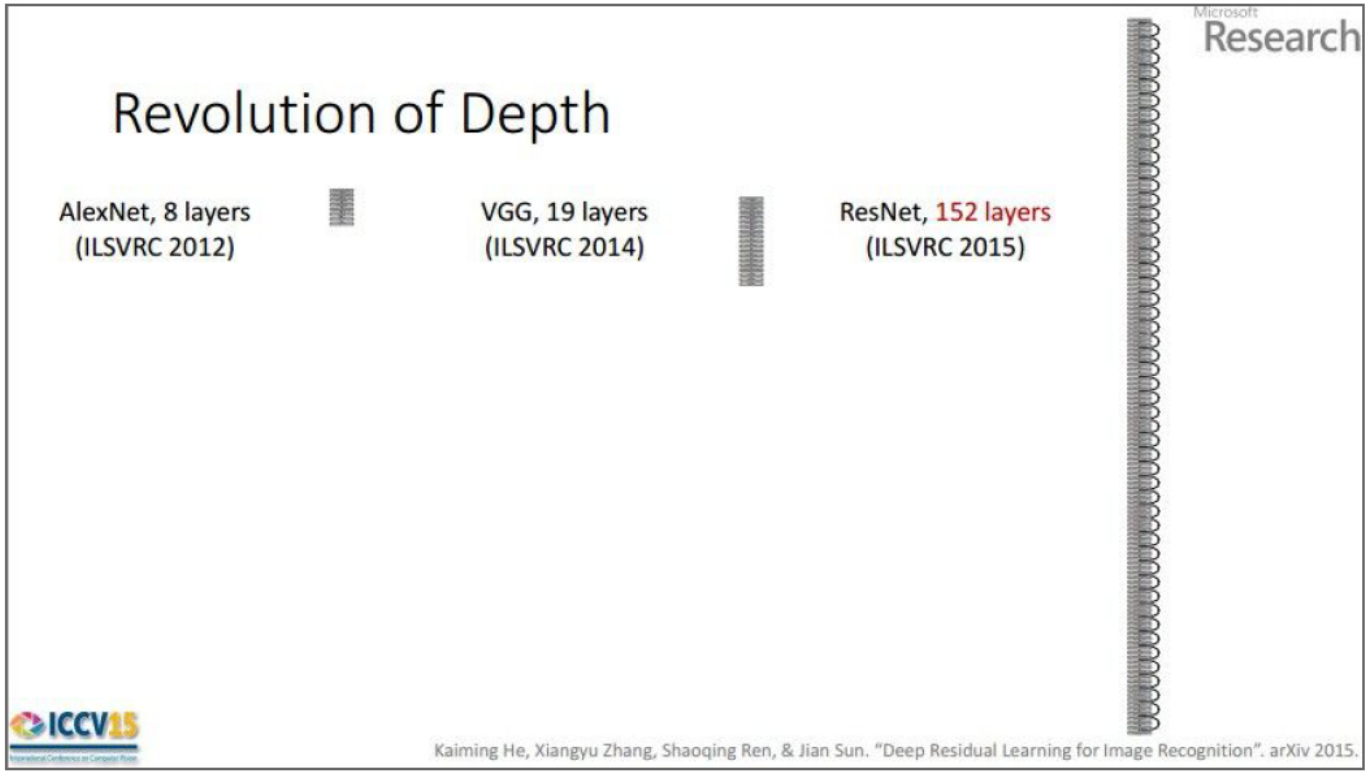
- ResNet [He et al., 2015]



CNN for object recognition

- ResNet [He et al., 2015]

ILSVRC 2015 winner (3.6% top 5 error)

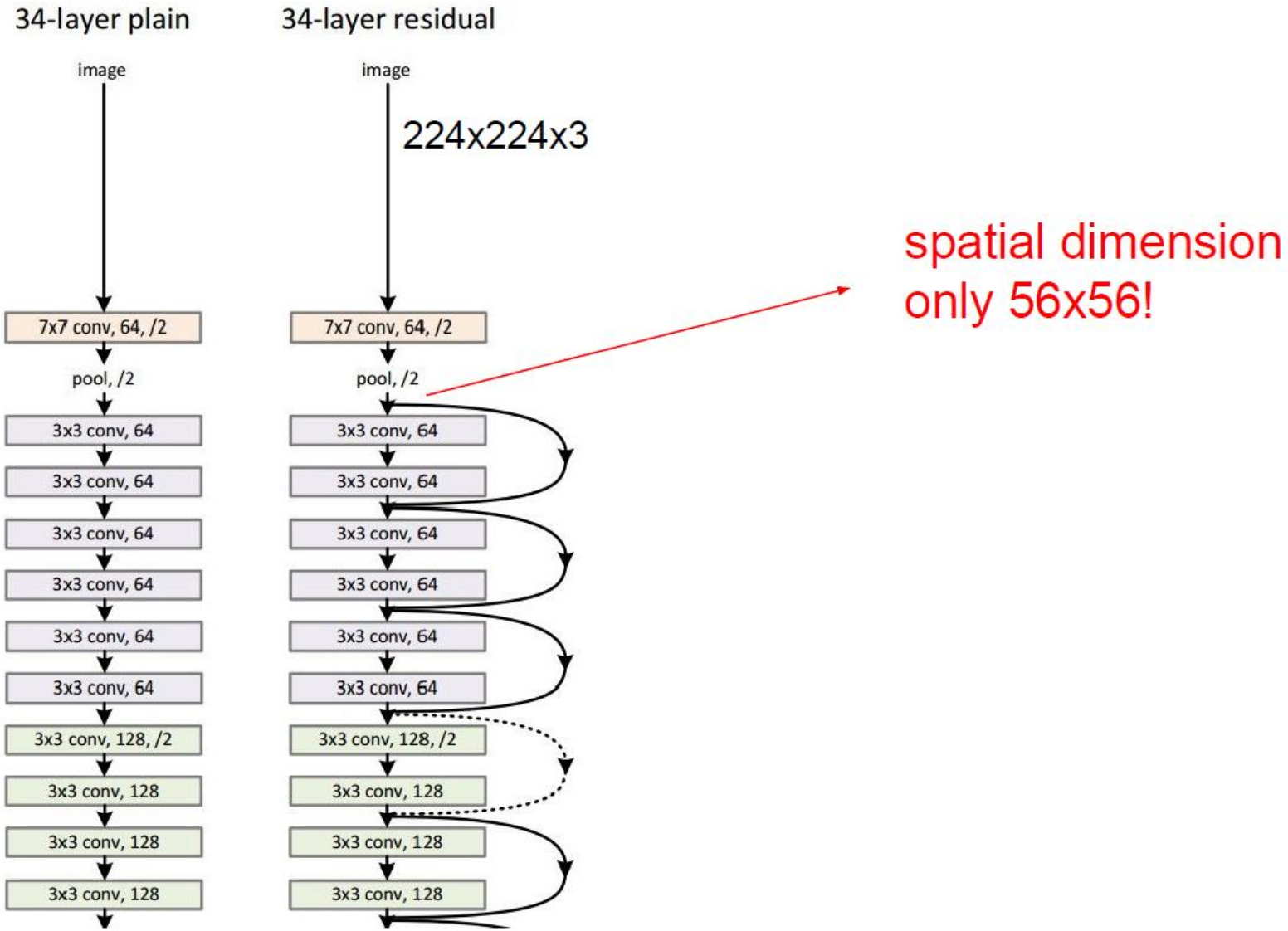


2-3 weeks of training on 8 GPU machine

at runtime: faster than a VGGNet! (even though it has 8x more layers)

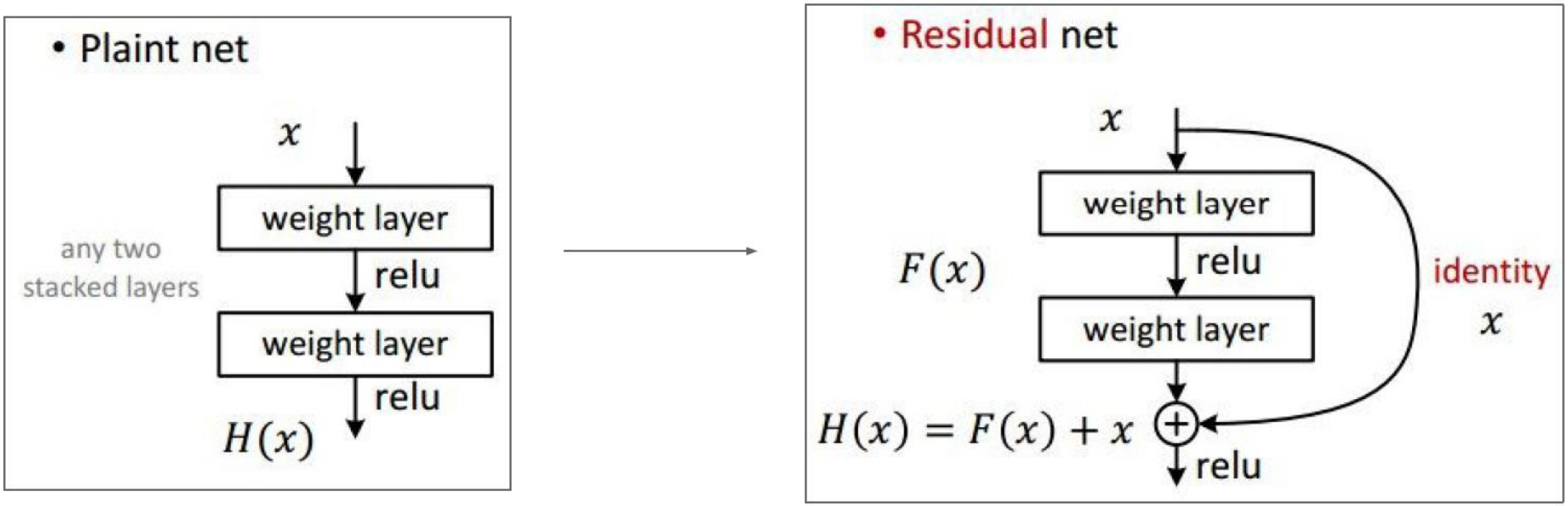
CNN for object recognition

- ResNet [He et al., 2015]



CNN for object recognition

- ResNet [He et al., 2015]



CNN for text classification

- CNN for sentence classification [Kim et al., 2014]

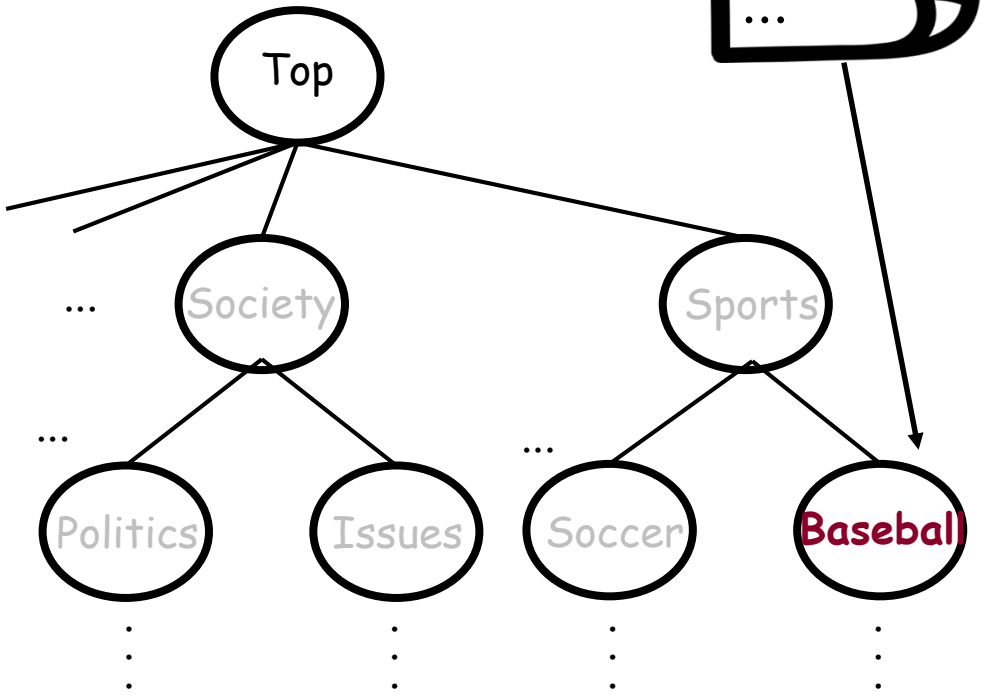
Five stars!

Positive
Negative

The Pitcher
…

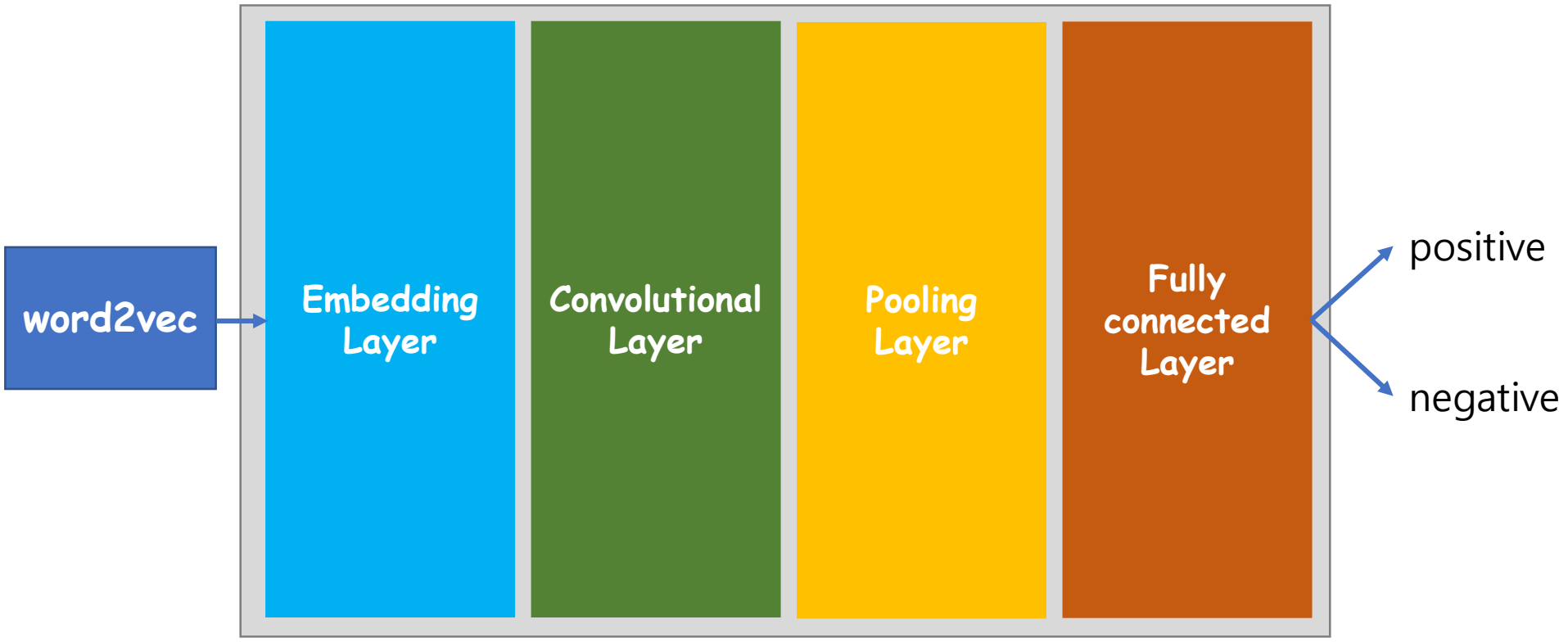
Economy
Sports
Weather
…

The Pitcher
…



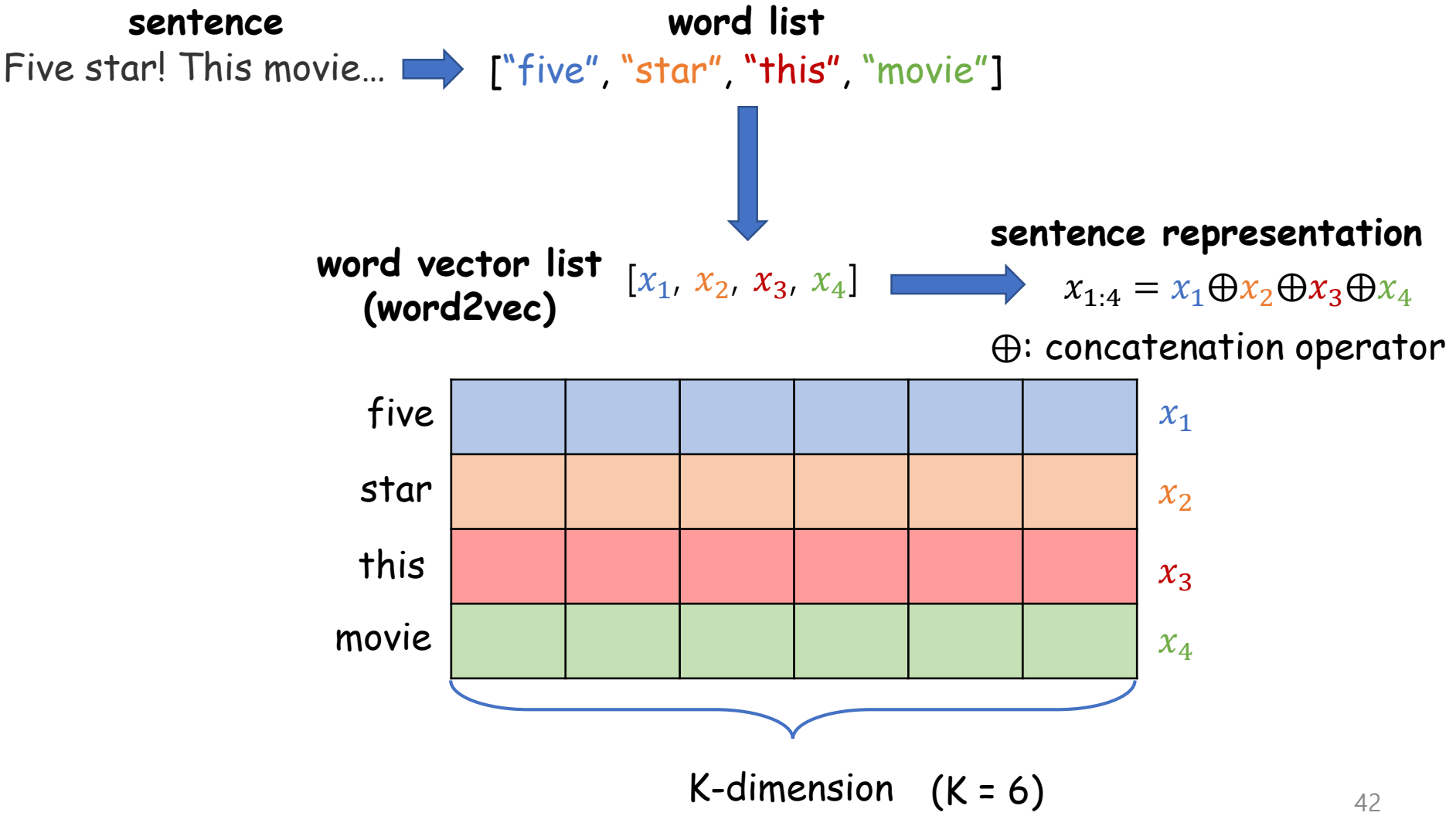
CNN for text classification

- Model overview



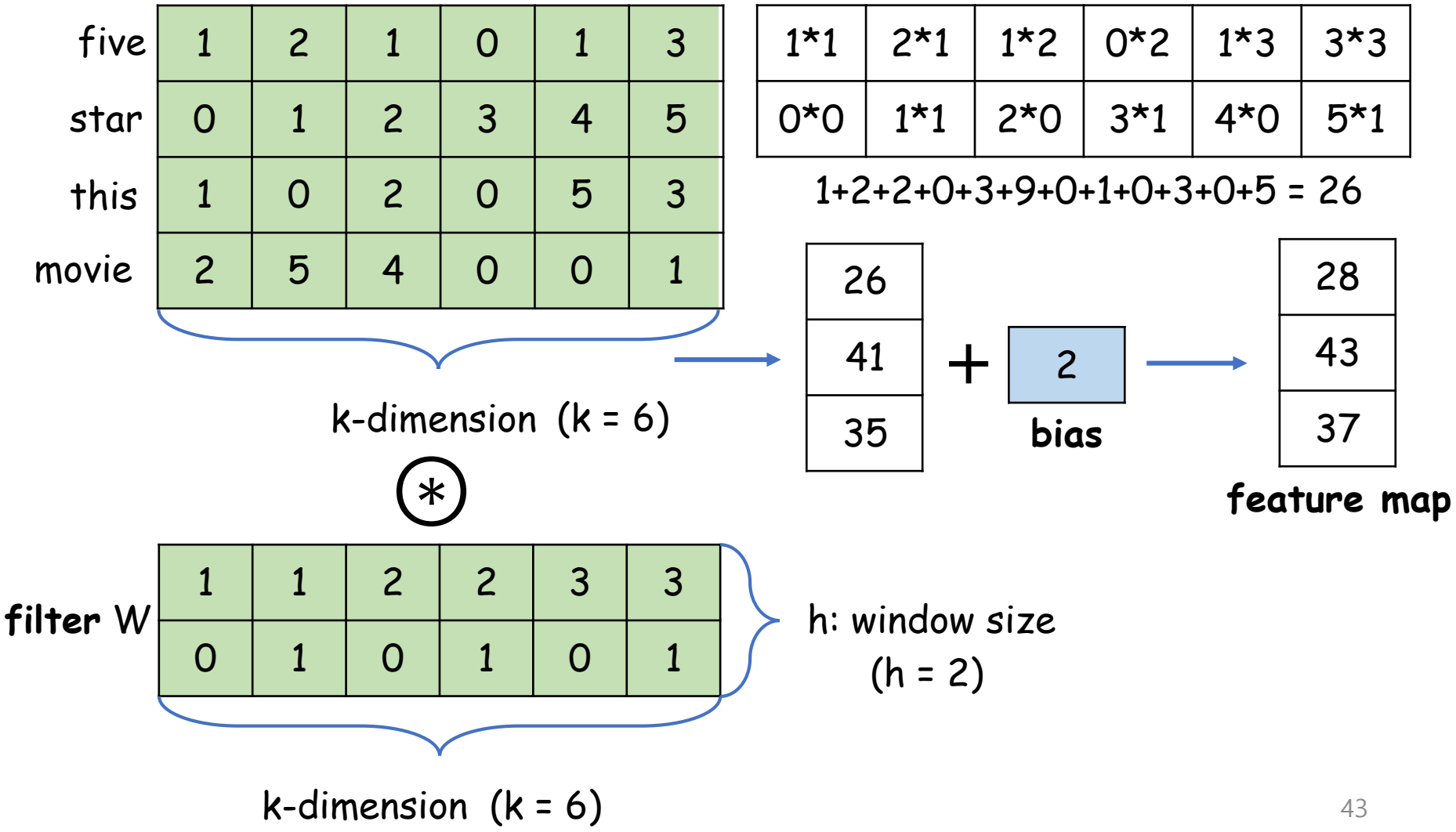
CNN for text classification

- Word2vec & Embedding layer



CNN for text classification

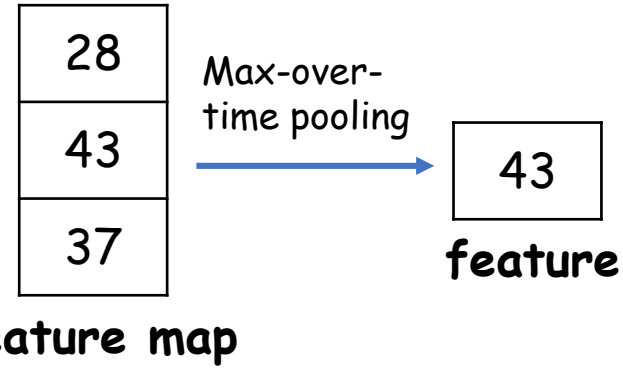
- Convolutional layer



CNN for text classification

- Pooling layer

five	1	2	1	0	1	3
star	0	1	2	3	4	5
this	1	0	2	0	5	3
movie	2	5	4	0	0	1

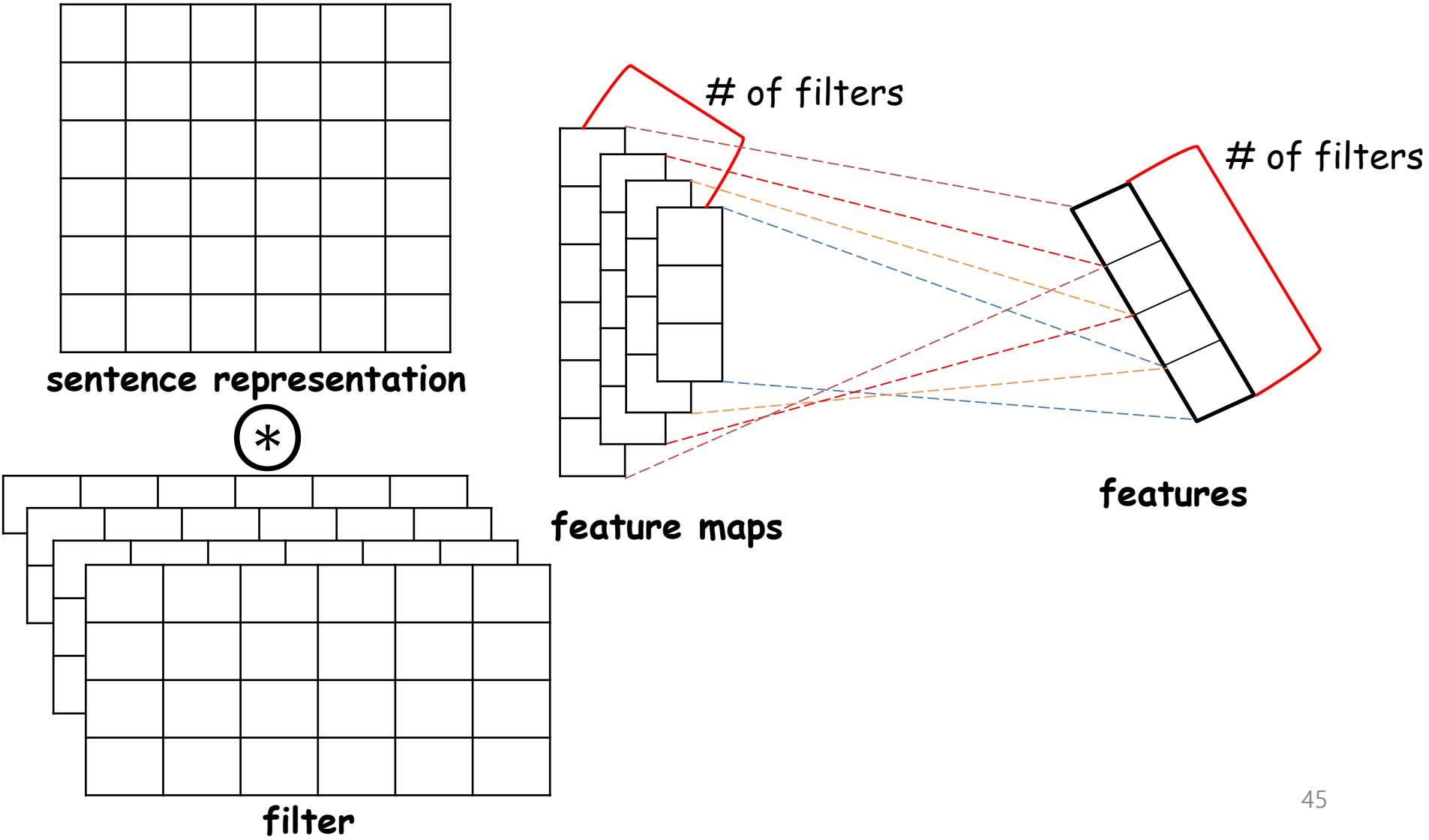


filter W	1	1	2	2	3	3
	0	1	0	1	0	1

One filter -> **One** feature
Multiple filter -> **Multiple** feature

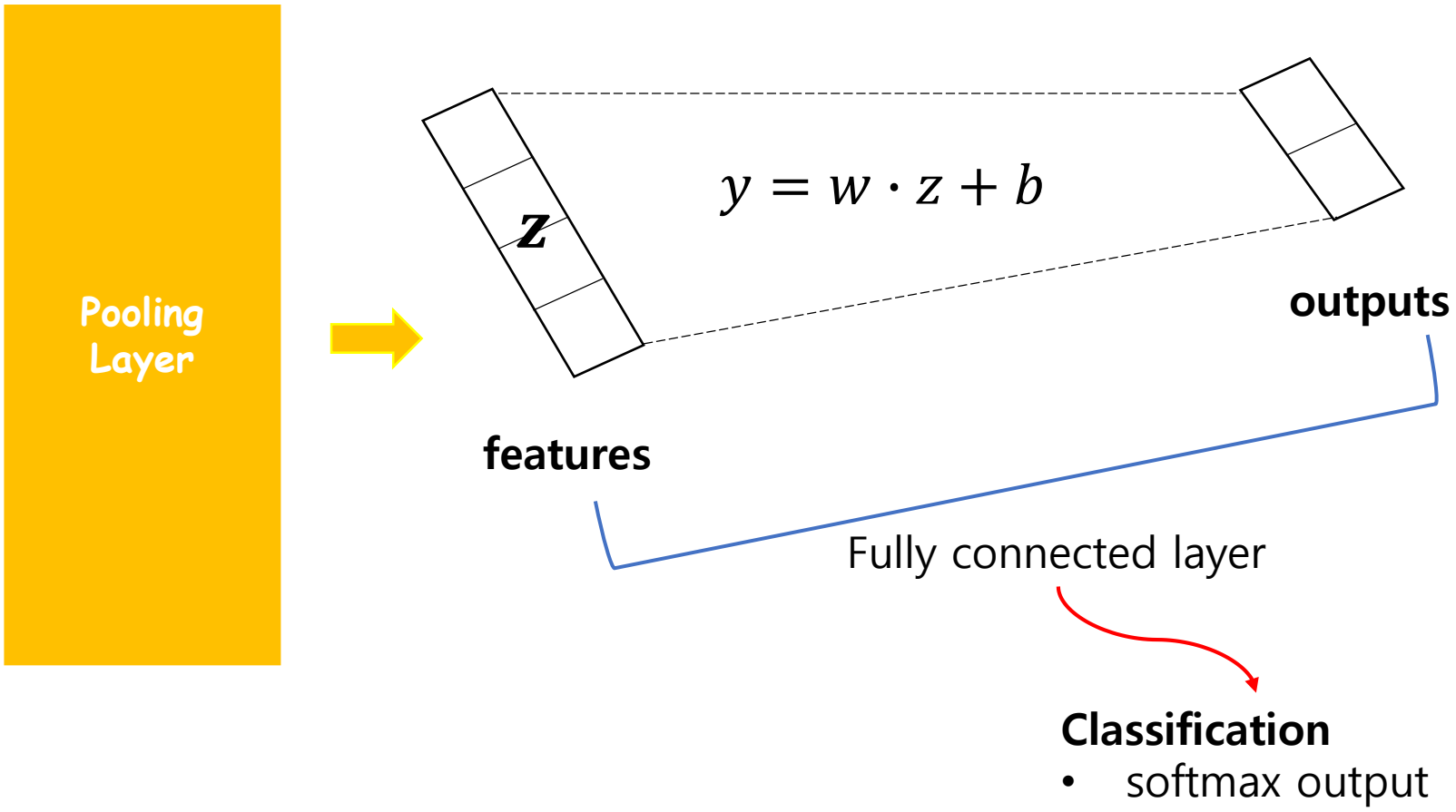
CNN for text classification

- Pooling layer



CNN for text classification

- Fully connected layer



AI School 6기 5주차

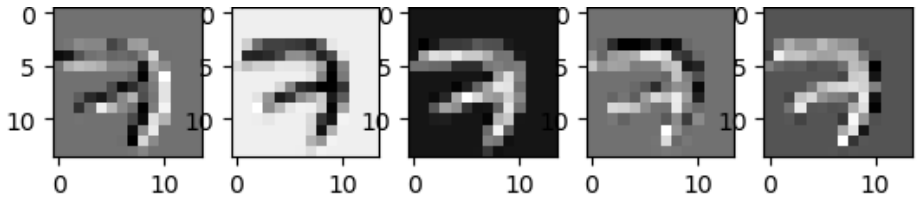
CNN 모델을 활용한 객체 분류

Image Input

```
X = tf.placeholder(tf.float32, [None, 784] , name="X")
X_img = tf.reshape(X, [-1, 28, 28, 1])   # img 28x28x1 (black/white)
Y = tf.placeholder(tf.float32, [None, 10], name="Y")
keep_prob = tf.placeholder(tf.float32, name="keep_prob")
```

Convolution

```
import matplotlib.pyplot as plt
import numpy as np
import tensorflow as tf
```



```
from tensorflow.examples.tutorials.mnist import input_data

mnist = input_data.read_data_sets("MNIST_data/", one_hot=True)
img = mnist.train.images[0].reshape(28,28)
sess = tf.InteractiveSession()

img = img.reshape(-1,28,28,1)
W1 = tf.Variable(tf.random_normal([3, 3, 1, 5], stddev=0.01))
conv2d = tf.nn.conv2d(img, W1, strides=[1, 2, 2, 1], padding='SAME')
print(conv2d)
sess.run(tf.global_variables_initializer())
conv2d_img = conv2d.eval()
conv2d_img = np.swapaxes(conv2d_img, 0, 3)
for i, one_img in enumerate(conv2d_img):
    plt.subplot(1,5,i+1), plt.imshow(one_img.reshape(14,14), cmap='gray')
plt.show()
```

Pooling

```
import matplotlib.pyplot as plt
import numpy as np
import tensorflow as tf
from tensorflow.examples.tutorials.mnist import input_data

mnist = input_data.read_data_sets("MNIST_data/", one_hot=True)
img = mnist.train.images[0].reshape(28,28)
sess = tf.InteractiveSession()

img = img.reshape(-1,28,28,1)
W1 = tf.Variable(tf.random_normal([3, 3, 1, 5], stddev=0.01))
conv2d = tf.nn.conv2d(img, W1, strides=[1, 2, 2, 1], padding='SAME')
pool = tf.nn.max_pool(conv2d, ksize=[1, 2, 2, 1], strides=[1, 2, 2, 1], padding='SAME')
print(pool)
sess.run(tf.global_variables_initializer())
pool_img = pool.eval()
pool_img = np.swapaxes(pool_img, 0, 3)
for i, one_img in enumerate(pool_img):
    plt.subplot(1,5,i+1), plt.imshow(one_img.reshape(7, 7), cmap='gray')
plt.show()
```

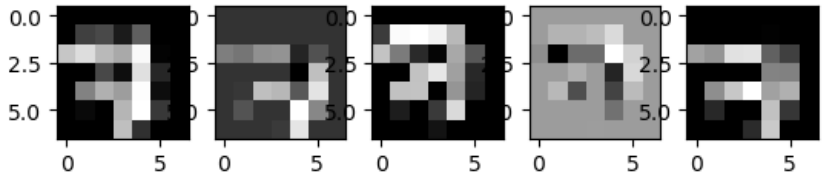


Image Input

```
X = tf.placeholder(tf.float32, [None, 784] , name="X")
X_img = tf.reshape(X, [-1, 28, 28, 1])   # img 28x28x1 (black/white)
Y = tf.placeholder(tf.float32, [None, 10], name="Y")
keep_prob = tf.placeholder(tf.float32, name="keep_prob")
```

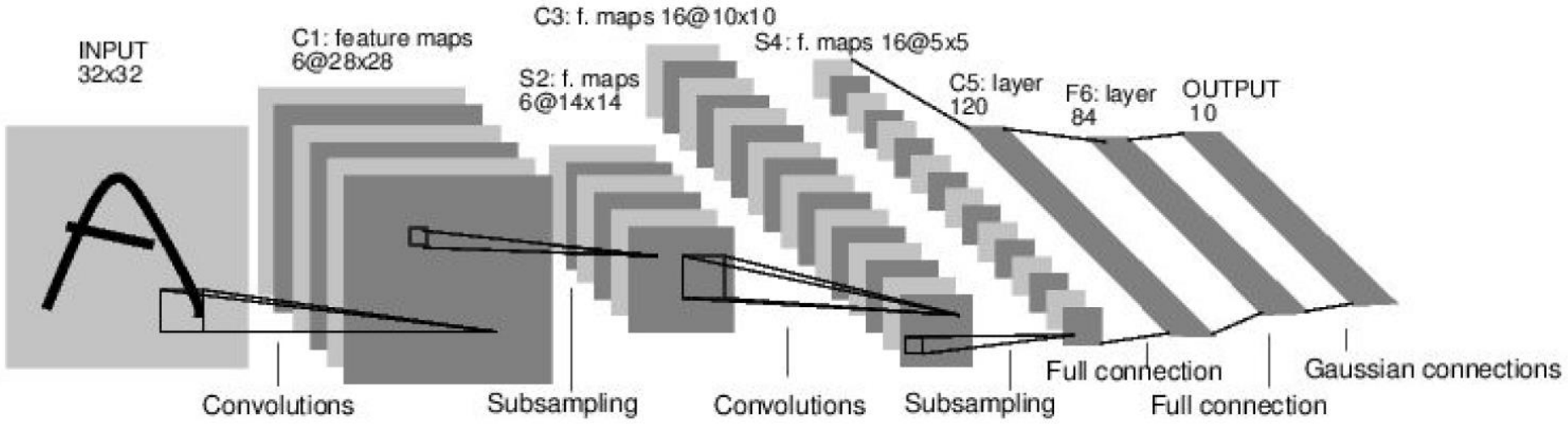
Convolutional & pooling layer 1, 2

```
# L1 ImgIn shape=(?, 28, 28, 1)
W1 = tf.Variable(tf.random_normal([3, 3, 1, 32], stddev=0.01))
#    Conv      -> (?, 28, 28, 32)
#    Pool      -> (?, 14, 14, 32)
L1 = tf.nn.conv2d(X_img, W1, strides=[1, 1, 1, 1], padding='SAME')
L1 = tf.nn.relu(L1)
L1 = tf.nn.max_pool(L1, ksize=[1, 2, 2, 1], strides=[1, 2, 2, 1], padding='SAME')
L1 = tf.nn.dropout(L1, keep_prob=keep_prob)
```

```
# L2 ImgIn shape=(?, 14, 14, 32)
W2 = tf.Variable(tf.random_normal([3, 3, 32, 64], stddev=0.01))
#    Conv      ->(?, 14, 14, 64)
#    Pool      ->(?, 7, 7, 64)
L2 = tf.nn.conv2d(L1, W2, strides=[1, 1, 1, 1], padding='SAME')
L2 = tf.nn.relu(L2)
L2 = tf.nn.max_pool(L2, ksize=[1, 2, 2, 1], strides=[1, 2, 2, 1], padding='SAME')
L2 = tf.nn.dropout(L2, keep_prob=keep_prob)
L2_flat = tf.reshape(L2, [-1, 7 * 7 * 64])
```

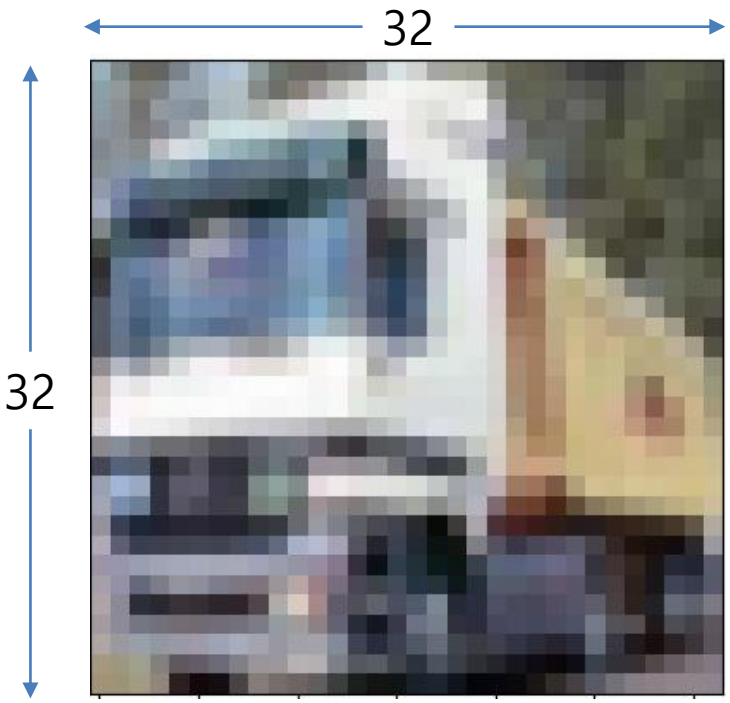
LeNet-5

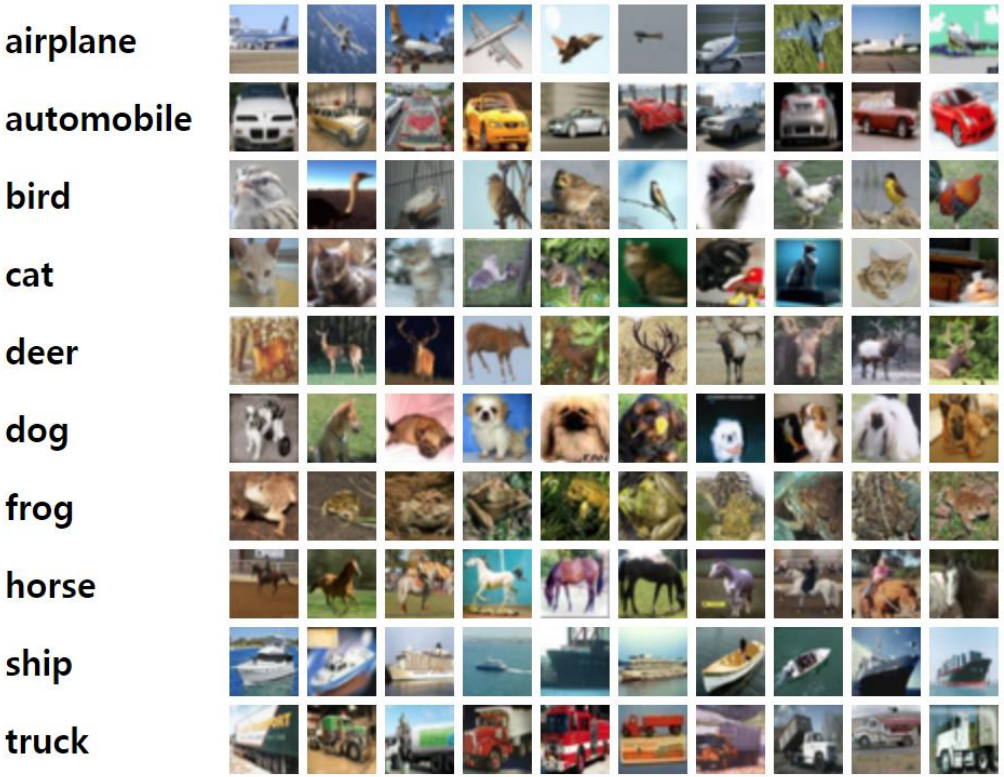
- LeNet [LeCun et al., 1998]



Conv filters were 5x5, applied at stride 1
Subsampling (Pooling) layers were 2x2 applied at stride 2
i.e. architecture is [CONV-POOL-CONV-POOL-CONV-FC]

CIFAR-10 data






```
x = tf.placeholder(tf.float32, shape=[None, 32, 32, 3])
y = tf.placeholder(tf.float32, shape=[None, 10])
```

CIFAR-10 data

```
import numpy as np
import matplotlib.pyplot as plt

from tensorflow.keras.datasets.cifar10 import load_data

(x_train, y_train), (x_test, y_test) = load_data()

print(np.shape(x_train))
print(np.shape(y_train))
print(np.shape(x_test))
print(np.shape(y_test))

# airplane, automobile, bird, cat, deer, dog, frog, horse,
ship, truck
print(y_train[1])
plt.imshow(x_train[1])
plt.show()
```

Image Input

```
X = tf.placeholder(tf.float32, [None, 32, 32, 3], name="X")
Y = tf.placeholder(tf.float32, [None, 10], name="Y")
keep_prob = tf.placeholder(tf.float32, name="keep_prob")
```

Convolutional & pooling layers

```
W1 = tf.Variable(tf.random_normal([3, 3, 3, 32], stddev=0.01))
L1 = tf.nn.conv2d(X, W1, strides=[1, 1, 1, 1], padding='SAME')
L1 = tf.nn.relu(L1)
L1 = tf.nn.max_pool(L1, ksize=[1, 2, 2, 1], strides=[1, 2, 2, 1],
padding='SAME')
L1 = tf.nn.dropout(L1, keep_prob=keep_prob)
.
.
.
W3 = tf.Variable(tf.random_normal([3, 3, 64, 128], stddev=0.01))
L3 = tf.nn.conv2d(L2, W3, strides=[1, 1, 1, 1], padding='SAME')
L3 = tf.nn.relu(L3)
L3 = tf.nn.max_pool(L3, ksize=[1, 2, 2, 1], strides=[1, 2, 2, 1],
padding='SAME')
L3 = tf.nn.dropout(L3, keep_prob=keep_prob)
```

```
L3_flat = tf.reshape(L3, [-1, 4 * 4 * 128])
```

Fully connected layers

```
W4 = tf.get_variable("W4", shape=[4 * 4 * 128, 128],
initializer=tf.initializers.he_normal())
b4 = tf.Variable(tf.random_normal([128]))
FC1 = tf.nn.relu(tf.nn.xw_plus_b(L3_flat, W4, b4))
FC1 = tf.nn.dropout(FC1, keep_prob=keep_prob)

W5 = tf.get_variable("W5", shape=[128, 64],
initializer=tf.initializers.he_normal())
b5 = tf.Variable(tf.random_normal([64]))
FC2 = tf.nn.relu(tf.nn.xw_plus_b(FC1, W5, b5))
FC2 = tf.nn.dropout(FC2, keep_prob=keep_prob)

W6 = tf.get_variable("W6", shape=[64, 10],
initializer=tf.initializers.he_normal())
b6 = tf.Variable(tf.random_normal([10]))
hypothesis = tf.nn.xw_plus_b(FC2, W6, b6, name="hypothesis")
```

preprocessing

```
max = 0
early_stopped = 0
(x_train_val, y_train_val), (x_test, y_test) = load_data()
shuffle_indices = np.random.permutation(np.arange(len(y_train_val)))
shuffled_x = np.asarray(x_train_val[shuffle_indices])
shuffled_y = y_train_val[shuffle_indices]
dev_sample_index = -1 * int(0.1 * float(len(y_train_val)))
x_train, x_val = shuffled_x[:dev_sample_index],
shuffled_x[dev_sample_index:]
y_train, y_val = shuffled_y[:dev_sample_index],
shuffled_y[dev_sample_index:]
x_test = np.asarray(x_test)

y_train_one_hot = np.eye(10)[y_train]
y_train_one_hot = np.squeeze(y_train_one_hot, axis=1)
y_test_one_hot = np.eye(10)[y_test]
y_test_one_hot = np.squeeze(y_test_one_hot, axis=1)
y_val_one_hot = np.eye(10)[y_val]
y_val_one_hot = np.squeeze(y_val_one_hot, axis=1)
```

Next batch

```
def next_batch(batch_size, data):
    data = np.array(data)
    np.random.seed(10)
    shuffle_indices = np.random.permutation(np.arange(len(data)))
    shuffled_data = data[shuffle_indices]
    num_batches_per_epoch = int((len(data)-1)/batch_size) + 1
    for batch_num in range(num_batches_per_epoch):
        start_index = batch_num * batch_size
        end_index = min((batch_num + 1) * batch_size, len(data))
        yield shuffled_data[start_index:end_index]
```

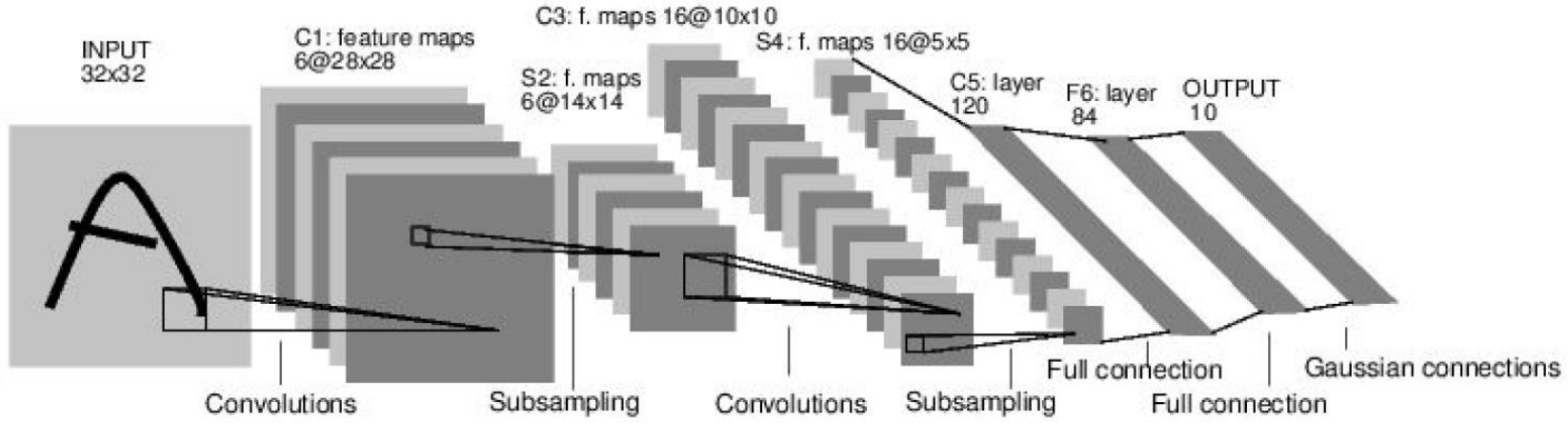
Training

```
for epoch in range(training_epochs):
    avg_cost = 0
    total_batch = int(len(y_train) / batch_size)
    batches = next_batch(batch_size, list(zip(x_train, y_train_one_hot)))
.
.
.
```

Homework

1. Implement Lenet-5!

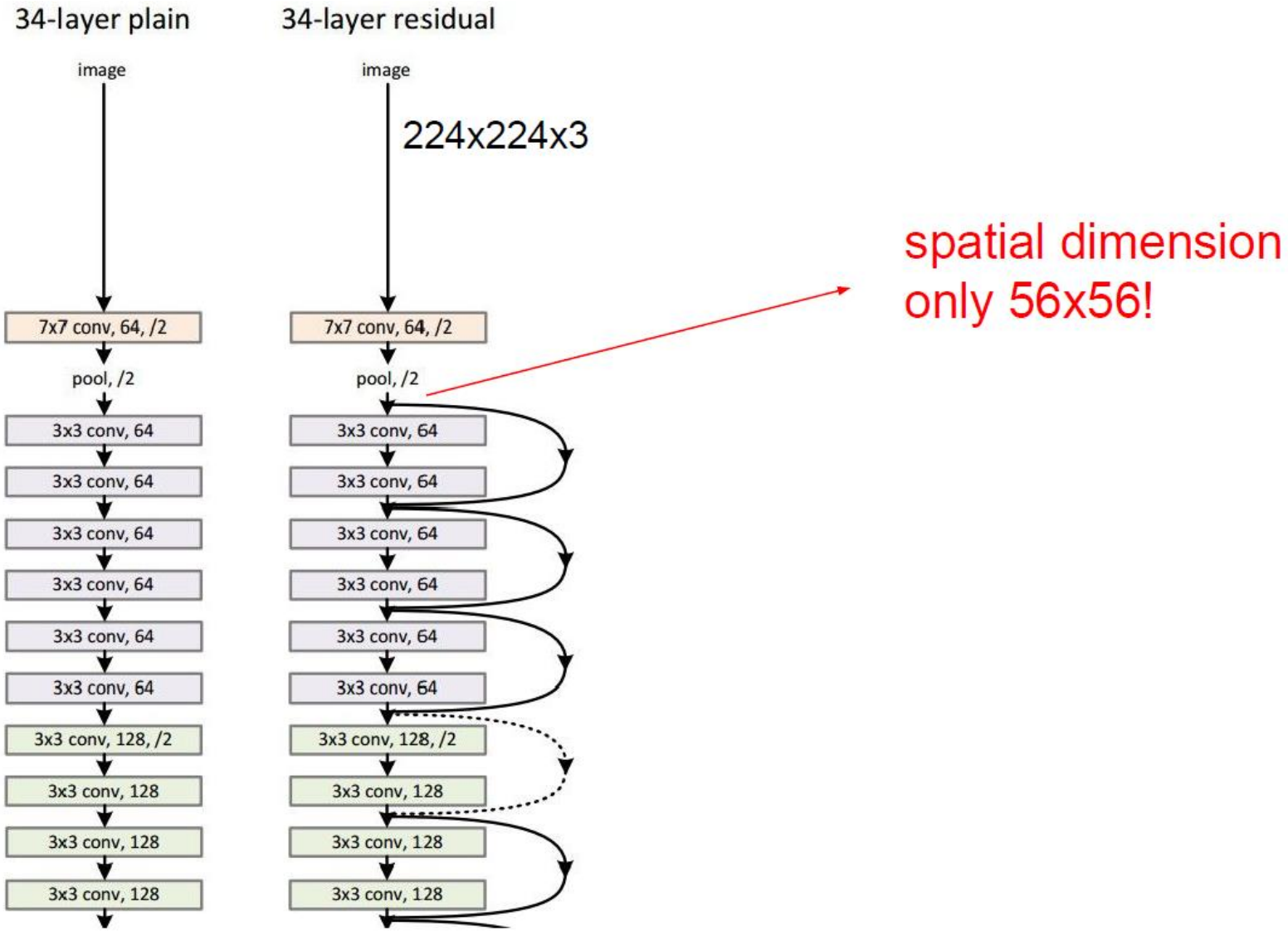
2. Change dropout ratio, initialization (He)



Conv filters were 5x5, applied at stride 1
Subsampling (Pooling) layers were 2x2 applied at stride 2
i.e. architecture is [CONV-POOL-CONV-POOL-CONV-FC]

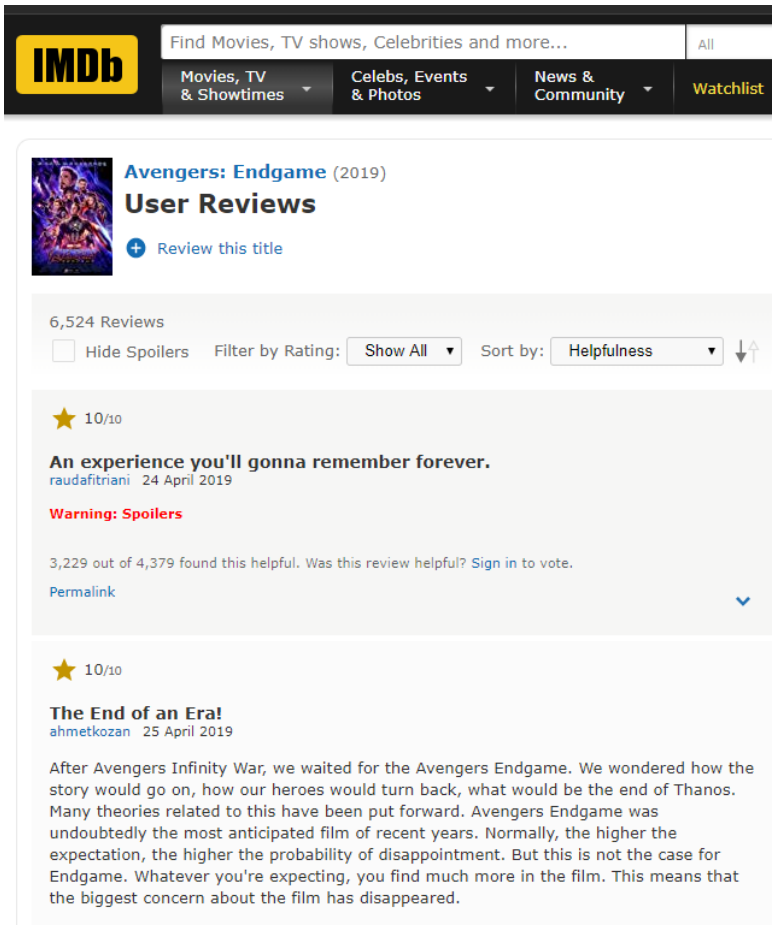
Next class (1교시)

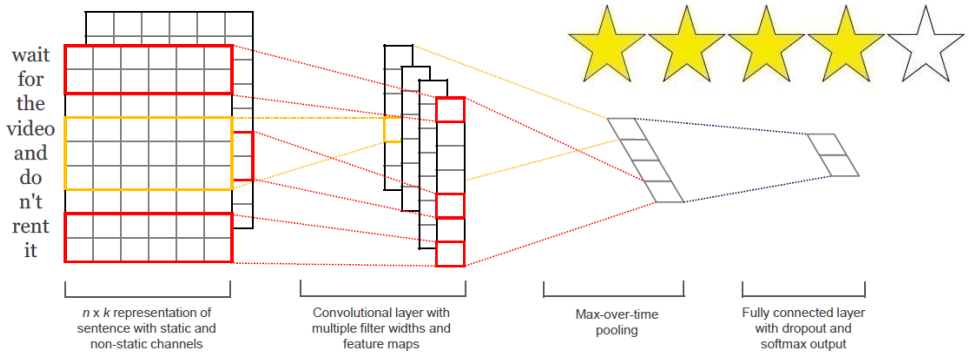
- ResNet [He et al., 2015]



Next class (2교시)

- CNN for sentence classification [Kim et al., 2014]





Q&A