#### **Psychoinformatics & Neuroinformatics**



Week 13
Image Processing &
Computer Vision



by Tsung-Ren (Tren) Huang 黄從仁

#### **Applications of Computer Vision**

Automated inference on criminality using face images:



(a) Three samples in criminal ID photo set  $S_c$ .



(b) Three samples in non-criminal ID photo set  $S_n$  Figure 1. Sample ID photos in our data set.

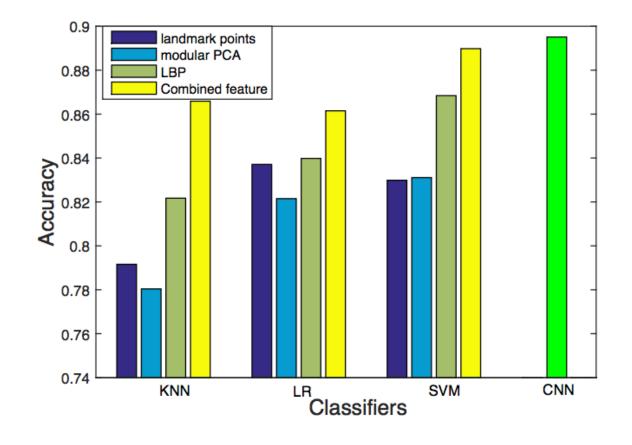


Figure 2. Accuracy of all four classifiers in all thirteen cases.

### **Topics for today**

**Image Processing** 

The processing unit is a pixel

Traditional Computer Vision

The processing unit is an object

**DL-based Computer Vision** 

The foundation is the convolutional NN



### **Topics for today**

**Image Processing** 

The processing unit is a pixel

Traditional Computer Vision

The processing unit is an object

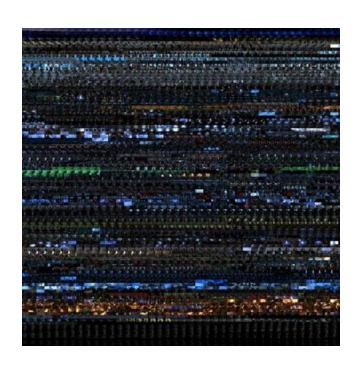
DL-based Computer Vision

The foundation is the convolutional NN



# **Examples of Data/Image Visualization**

The frames of two movies:

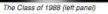




Graduation photos of different generations:













#### **Two Python Packages**

Matplotlib for visualization & PIL/Pillow for processing



#### Matplotlib: Visualization with Python

Matplotlib is a comprehensive library for creating static, animated, and interactive visualizations in Python. Matplotlib makes easy things easy and hard things possible.

- · Create publication quality plots.
- Make interactive figures that can zoom, pan, update.
- Customize visual style and layout.
- Export to many file formats.
- Embed in JupyterLab and Graphical User Interfaces.
- Use a rich array of third-party packages built on Matplotlib.



The Python Imaging Library adds image processing capabilities to your Python interpreter.

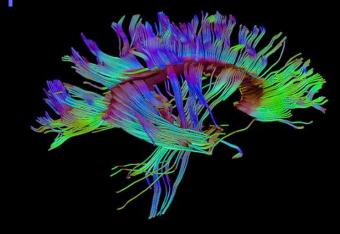
This library provides extensive file format support, an efficient internal representation, and fairly powerful image processing capabilities.

The core image library is designed for fast access to data stored in a few basic pixel formats. It should provide a solid foundation for a general image processing tool.

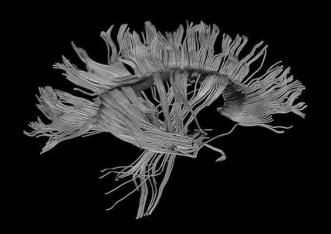
#### **Example 1: Color to gray**

You can do it by yourself or by pillow:

```
import matplotlib.pyplot as plt, numpy as np
img1=plt.imread('DTI.jpg')
plt.imshow(img1); plt.figure()
img2=np.mean(img,2)
plt.imshow(img2,cmap=plt.cm.gray)
```



```
from PIL import Image
img1=Image.open('DTI.jpg')
img2=img1.convert('L')
img2.show(); img2.save('DTI_gray.jpg')
```



#### **Example 2: Image Contrast**

You can run +-\*/ on images:

```
import numpy as np,
import matplotlib.pyplot as plt
img=[]
img.append(np.float64(plt.imread('MRI1.jpg')))
img.append(np.float64(plt.imread('MRI2.jpg')))
img.append(img[1]-img[0]) #contrast
for i in range(3):
plt.subplot(1,3,i+1); plt.axis('off')
plt.imshow(img[i],cmap=plt.cm.gray)
```

### **Example 3: Image Blending & Filtering**

You can run +-\*/ on images:

```
from PIL import Image
from PIL.ImageFilter import CONTOUR
img1=Image.open('house.jpg')
img2=Image.open('face.jpg')
k=np.arange(1,10,2)/10.0
for i in range(5):
  hybrid=Image.blend(img1,img2,k[i])
  hybrid=hybrid.filter(CONTOUR)
  hybrid.show()
```









### **Topics for today**

Image Processing
The processing unit is a pixel

Traditional Computer Vision

The processing unit is an object

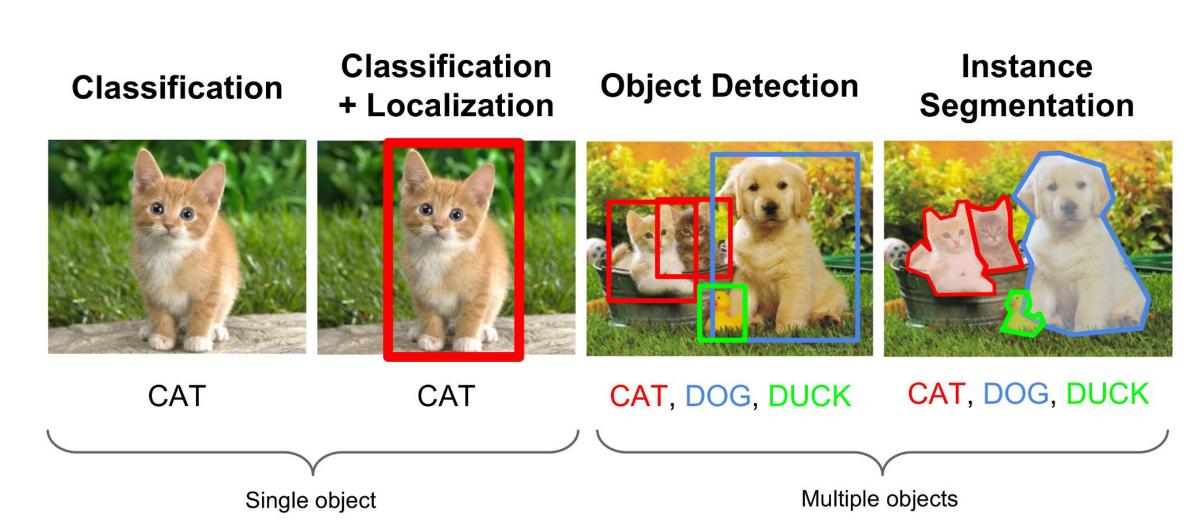
DL-based Computer Vision

The foundation is the convolutional NN



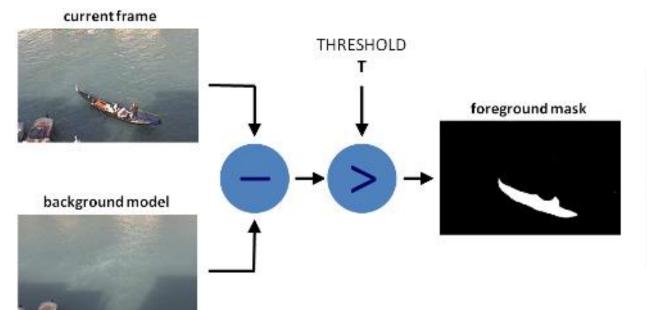
#### Computer Vision (1/2)

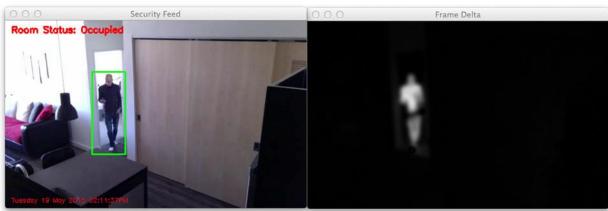
From what to where:



#### Computer Vision (2/2)

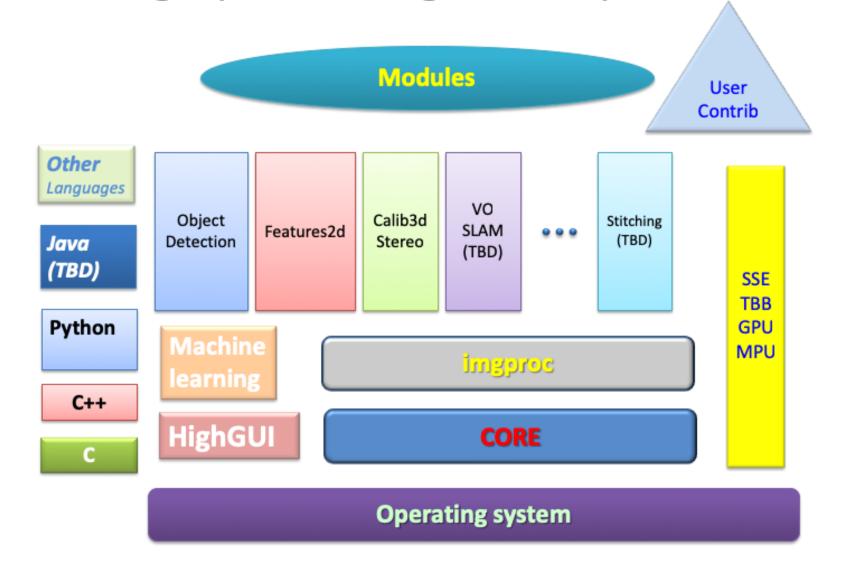
From images to videos:





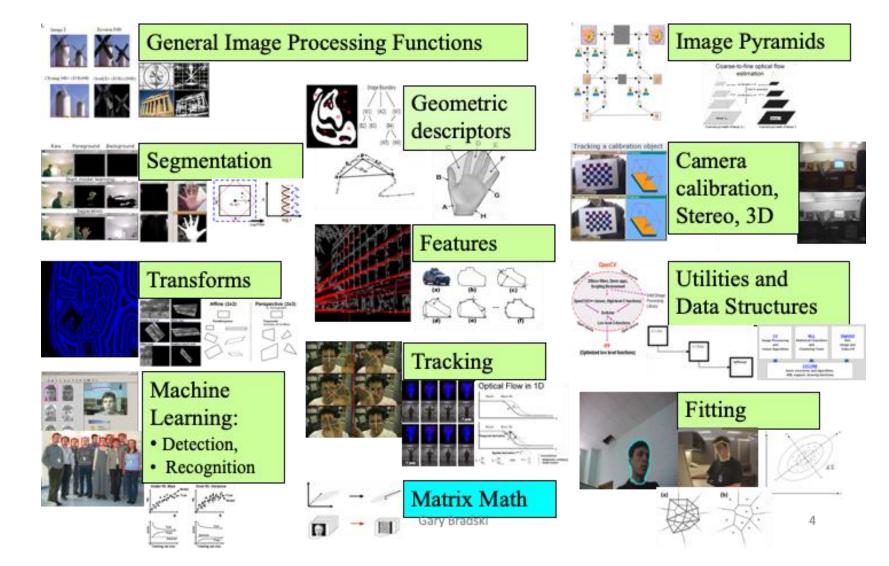
#### **OpenCV (1/2)**

A library for image processing + computer vision:



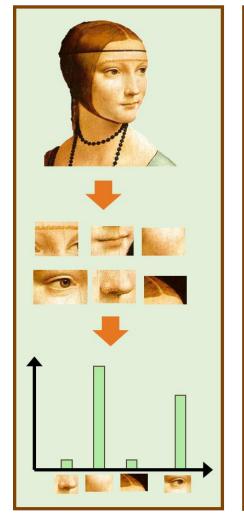
### **OpenCV (2/2)**

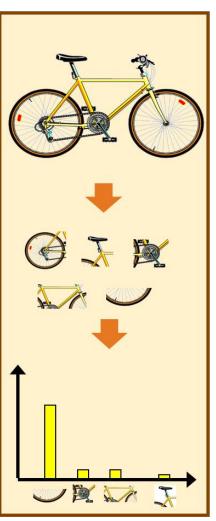
More than 2,500 algorithms:

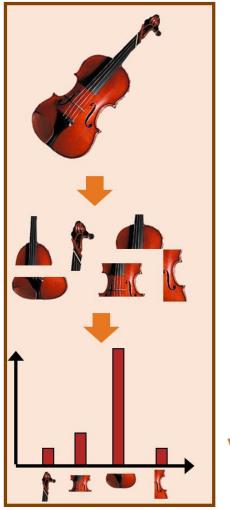


#### Features for Object Recognition

Bag of Visual Words (BOVW) is just one example:



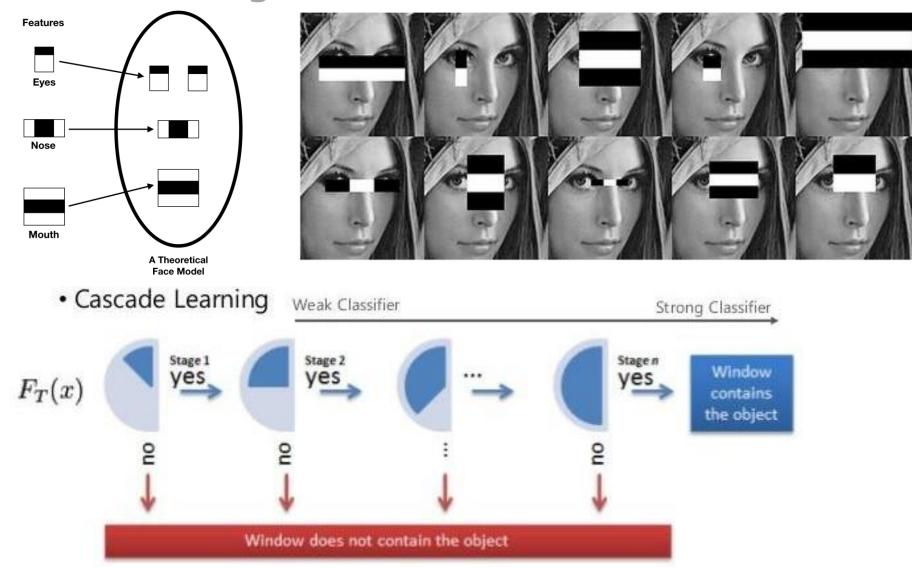






#### Face Recognition: Haar Cascades

AdaBoost on basic edge/line features:



### **Topics for today**

Image Processing

The processing unit is a pixel

Traditional Computer Vision

The processing unit is an object

**DL-based Computer Vision** 

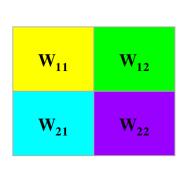
The foundation is the convolutional NN

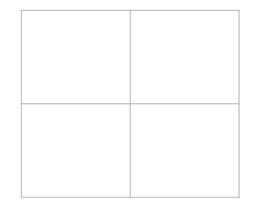


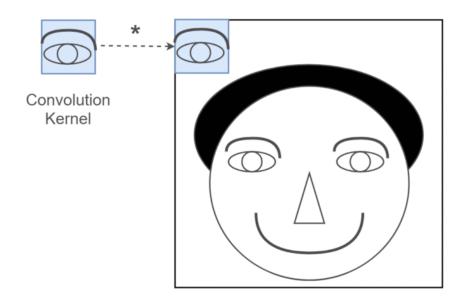
## 2D Convolution/Filtering

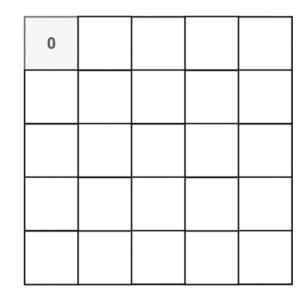
Conv2D(img, krnl)=filter2D(img,上下顛倒左右相反(krnl))

X <sub>11</sub>	X <sub>12</sub>	X <sub>13</sub>
X <sub>21</sub>	X <sub>22</sub>	X <sub>23</sub>
X <sub>31</sub>	X <sub>32</sub>	X <sub>33</sub>

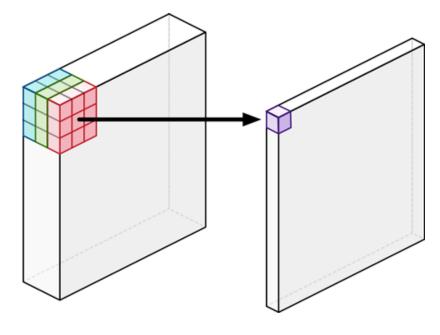








Kernel size=3\*3\*3



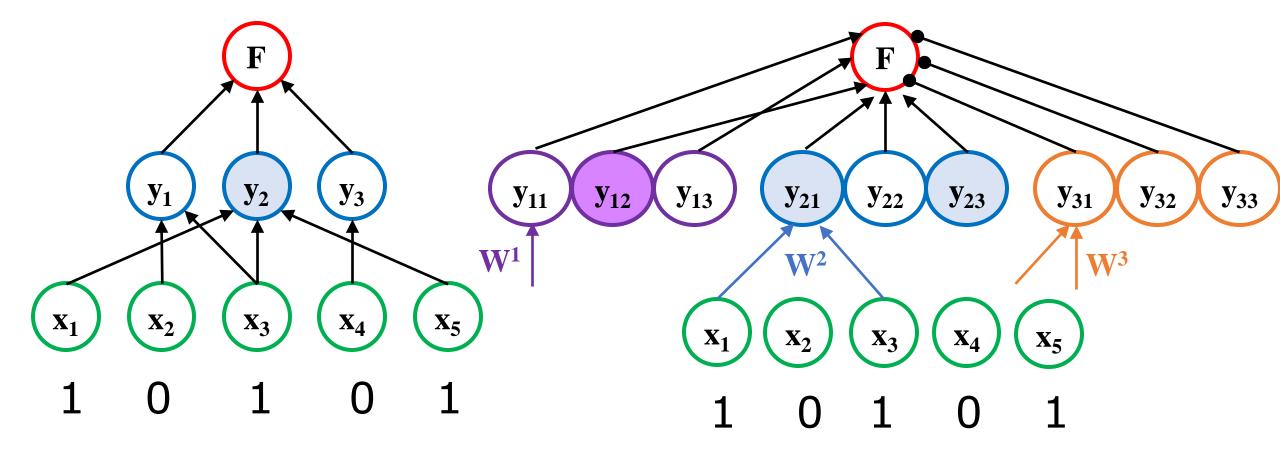
Image

Convolution Output

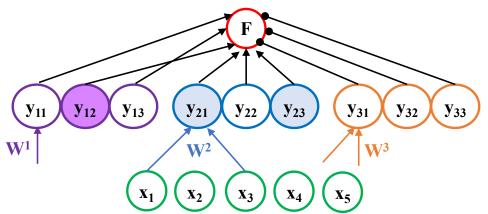
#### Non-shared vs. shared weights

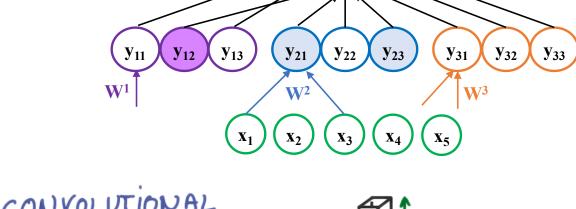
```
y₁ detects [0 1 1 0 0] globally
y<sub>2</sub> detects [1 0 1 0 1] globally
y<sub>3</sub> detects [0 0 0 1 0] globally W<sup>3</sup> detects [1 1 0] locally
```

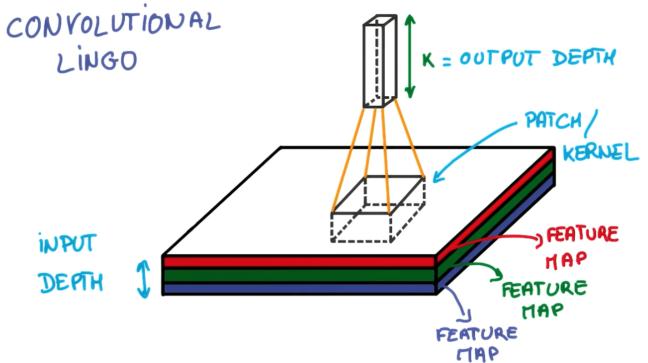
W<sup>1</sup> detects [0 1 0] locally W<sup>2</sup> detects [1 0 1] locally



#### **Convolutional Lingo**







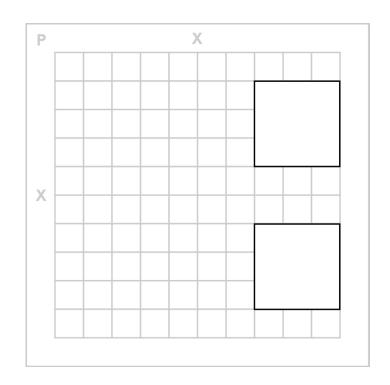


X = image size

F = filter size

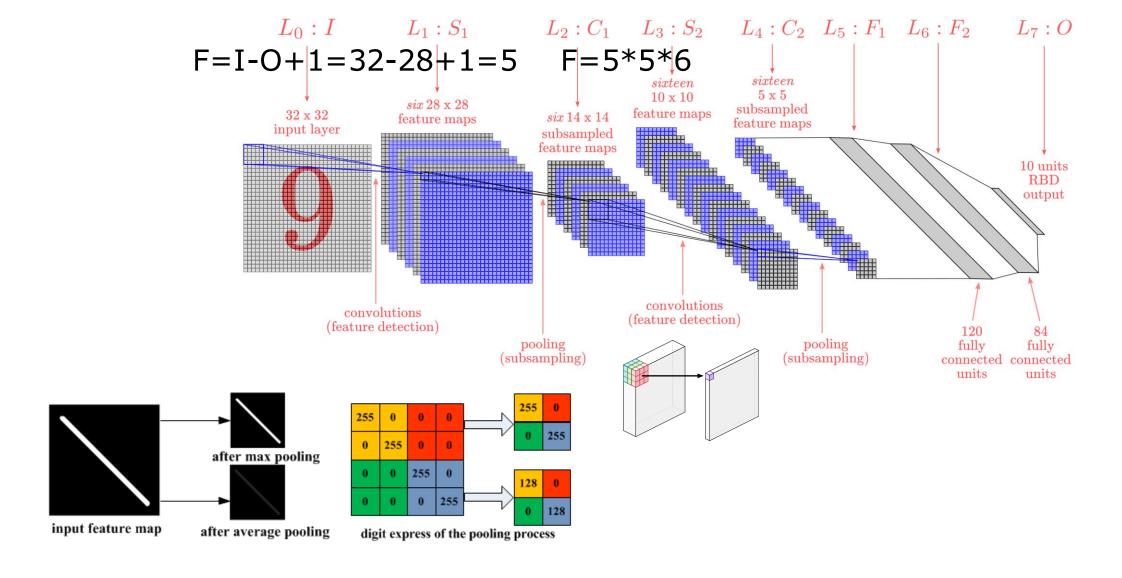
P = padding

S = stride

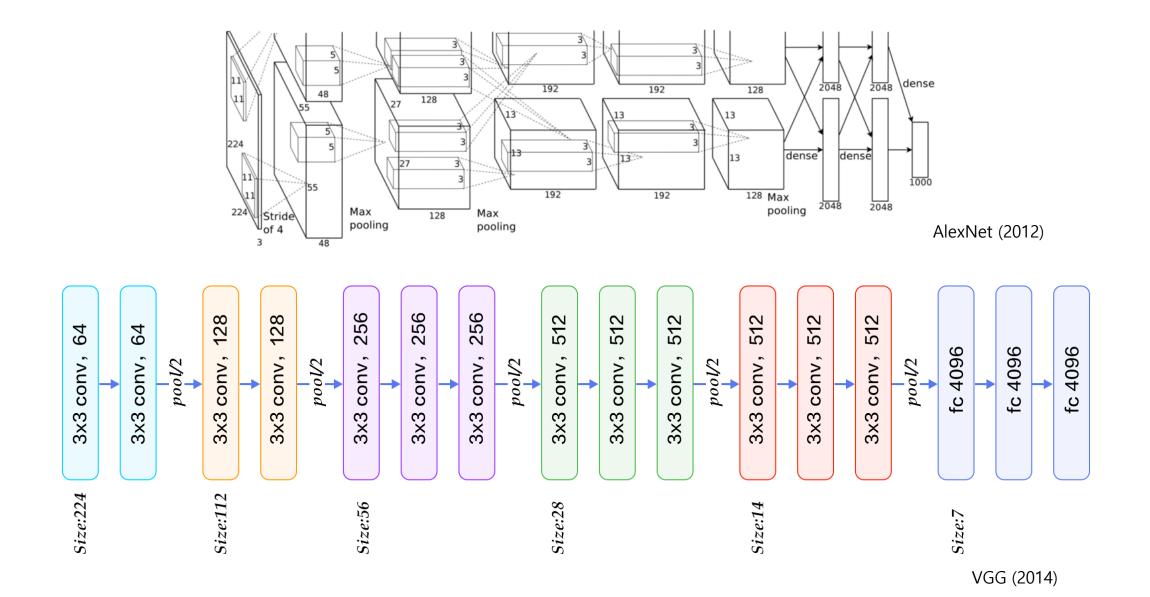


If 
$$(P,S)=(0,1)$$
,  
then  $O=1+I-F$   
 $\Rightarrow F=I-O+1$ 

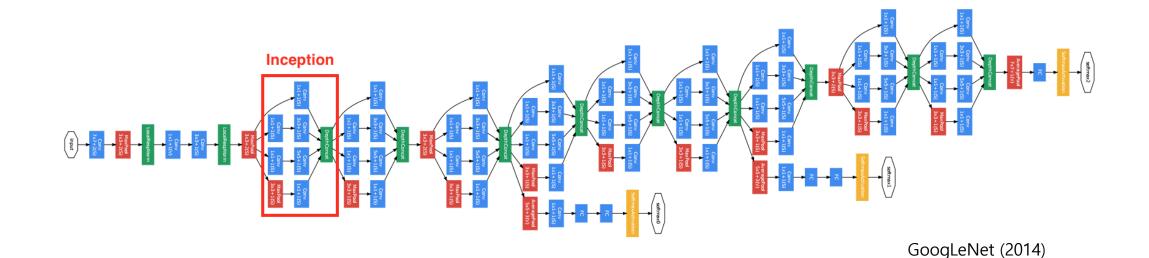
#### **Basic CNN: LeNet**

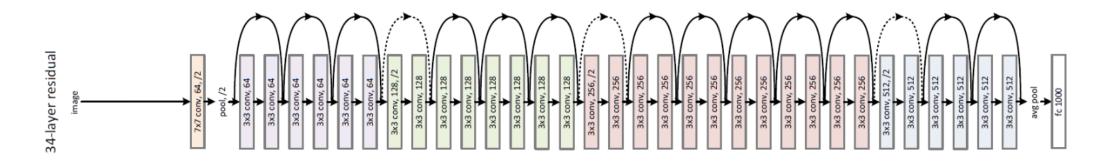


## Evolution of CNNs (1/3): GPU+Deeper



#### **Evolution of CNNs (2/3): Wider/Ensemble**

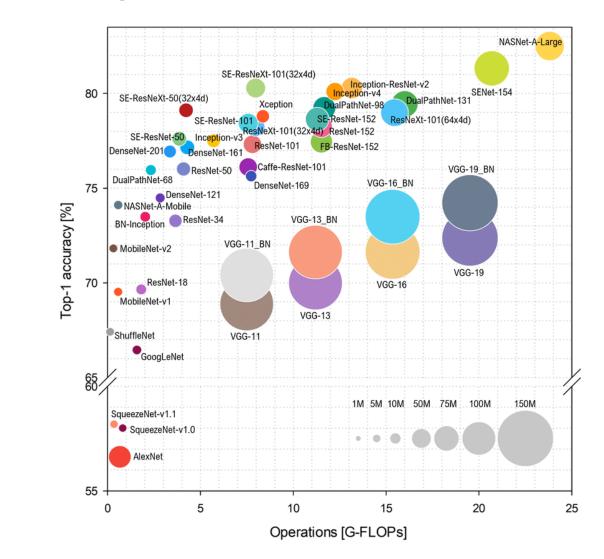




#### **Evolution of CNNs (3/3)**

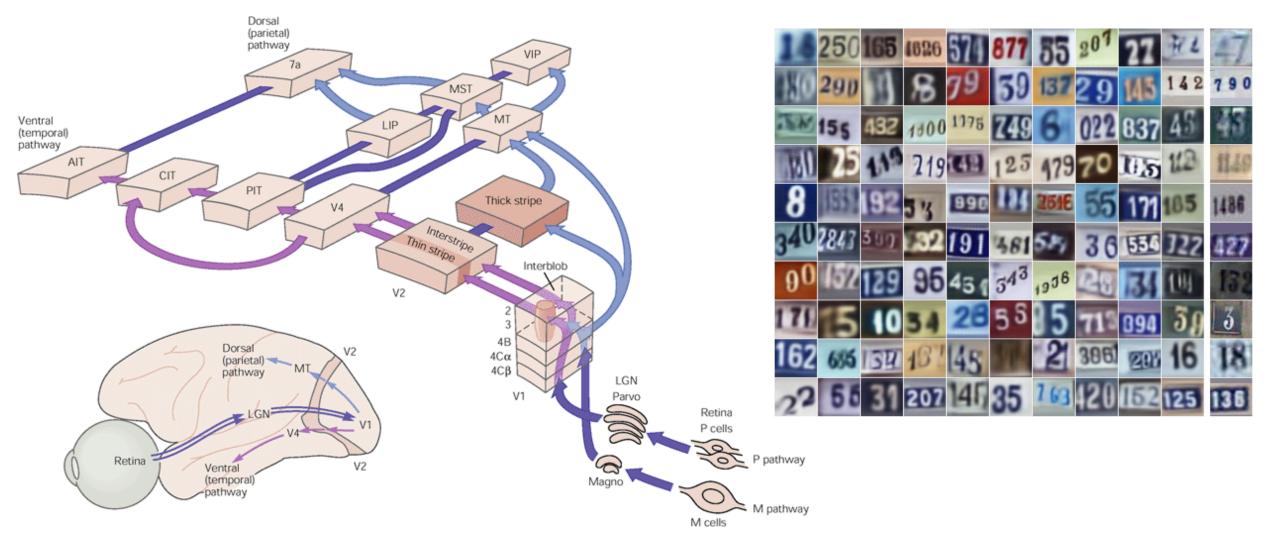
 $\neq$ 

More complex models # Better models



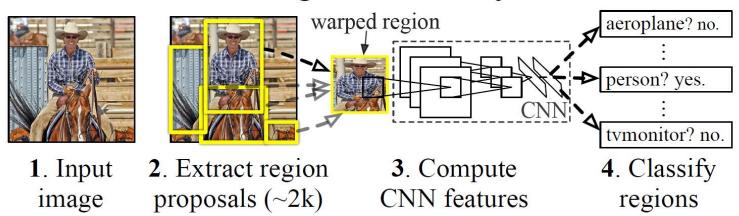
#### From "What" to "Where" Information

CV is incomplete & inefficient w/o processing "Where" info



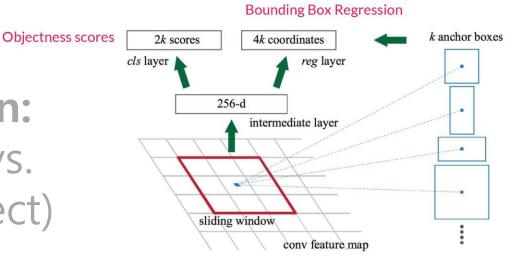
### Region-based CNN (Faster RCNN)

#### R-CNN: Regions with CNN features



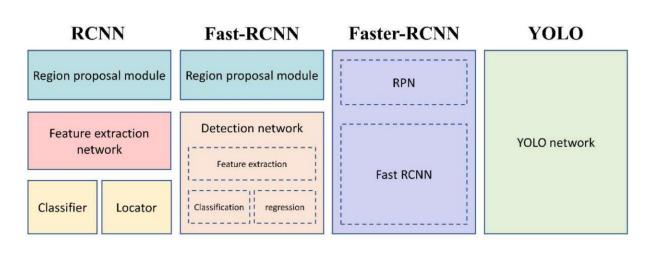
#### **Objectness classification:**

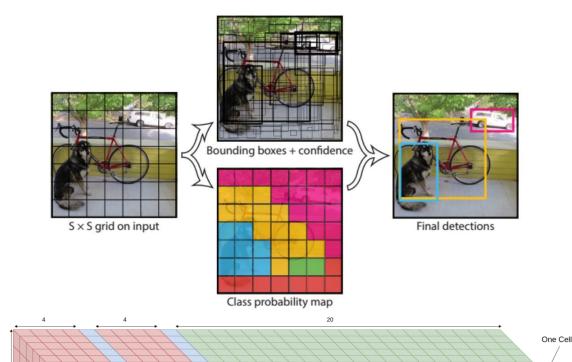
Foreground (an object) vs. Background (not an object)



### From RCNN to YOLO (U Only Look Once)

YOLO simultaneously predicts a object class & its potential bounding boxes at the same time!





### **Topics for today**

**Image Processing** 

The processing unit is a pixel

Traditional Computer Vision

The processing unit is an object

**DL-based Computer Vision** 

The foundation is the convolutional NN





