

Convolutional Neural Network (Basic) Cheat Sheet

(source: DeepLearning.AI - Andrew Ng)
V2021.01.02
(Dr Yan Xu)

Core Motivations

- Significantly reduce # of parameters to extract features in hidden layers
- Sparse localized connections
- Parameter sharing

Basic Concepts

→ Feature Extraction Layers:

- Convolution Layer
 - Stride (default: 1)
 - Padding (default: 0)
 - Kernel size (or Filter size)
 - Kernel count (or Filter count)
- Non-linear activation function (e.g. ReLU)
- Pooling Layer
 - max pooling

$$\begin{matrix} n \times n \text{ image} & f \times f \text{ filter} \\ \text{padding } p & \text{stride } s \end{matrix}$$

$$\left\lfloor \frac{n+2p-f}{s} + 1 \right\rfloor \times \left\lfloor \frac{n+2p-f}{s} + 1 \right\rfloor$$

→ Classification Layers:

- Flatten operation
- Fully Connected Layer
- Softmax Layer

1: The # of parameters in feature extraction layers is small

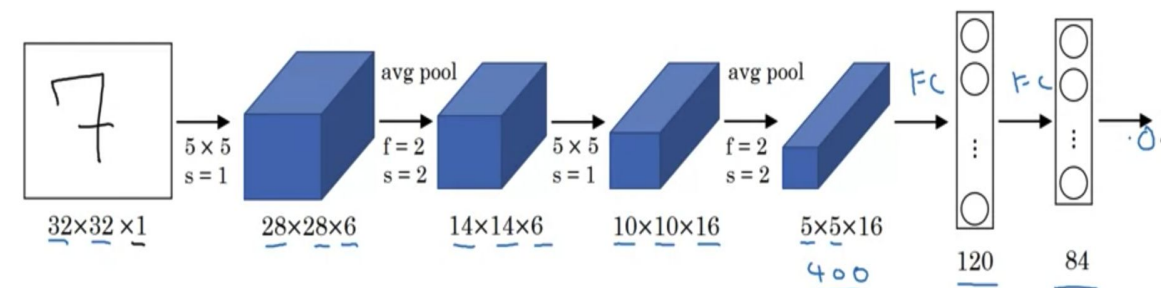
Parameter Count Example

	Activation shape	Activation Size	# parameters
Input:	(32,32,3)	3,072	0
CONV1 (f=5, s=1)	(28,28,8)	6,272	208
POOL1	(14,14,8)	1,568	0
CONV2 (f=5, s=1)	(10,10,16)	1,600	416
POOL2	(5,5,16)	400	0
FC3	(120,1)	120	48,001
FC4	(84,1)	84	10,081
Softmax	(10,1)	10	841

CNN Popular Architectures

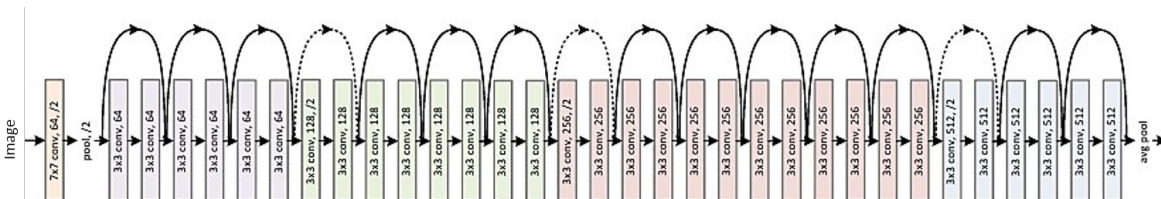
→ Classic Networks

- LeNet - 5 (*a small network*)
- AlexNet
- VGG - 16



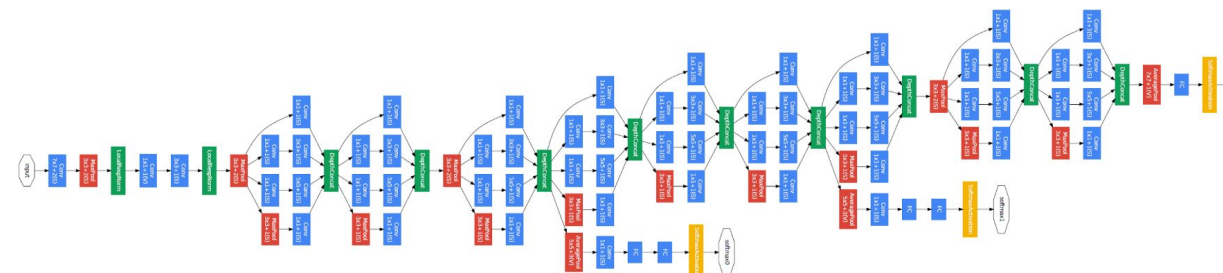
→ ResNet (deep networks)

- ability to learn the identity transformation



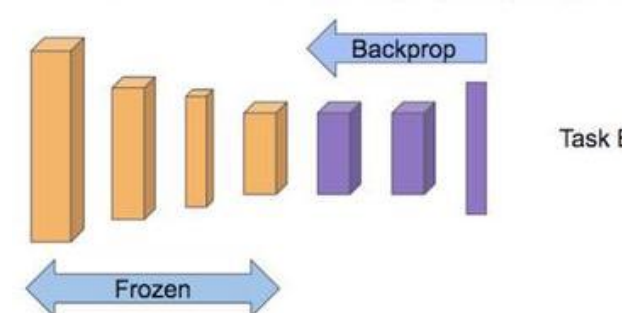
→ Inception Network

- To concat various sizing filters & pooling together
- Trick: use 1*1 convolution to reduce filters
- Variant: GoogLeNet



Transfer Learning

- Frozen layers
- Initial parameters



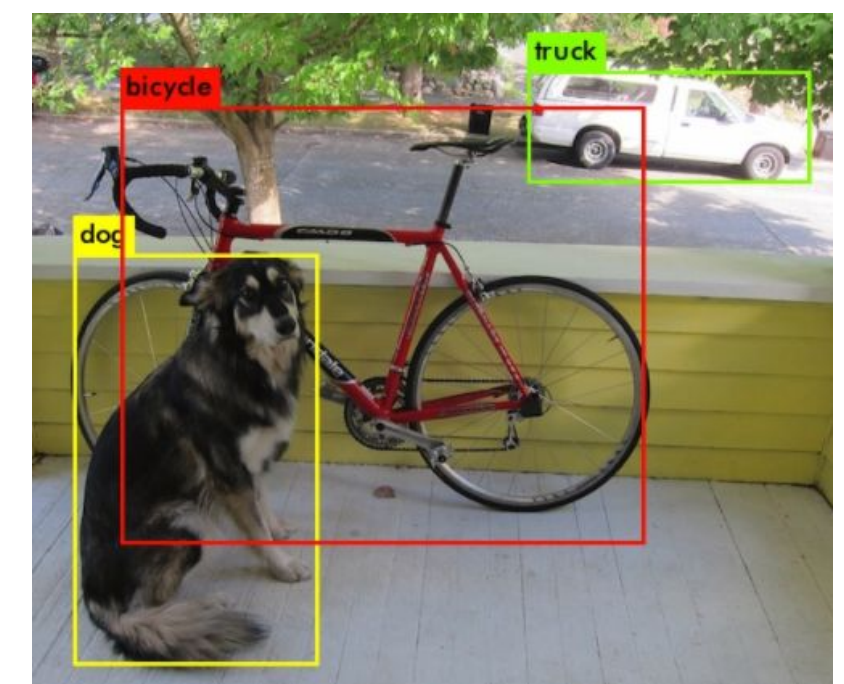
Object Detection (e.g. YOLO)

- Idea 1: Object & Location Learning Together
 - location encoding should be consistent
- Idea 2: Sliding Window Detection
- Idea 3: Sliding Window as CNN
 - speed up calculation
- Idea 4: Object/Cell Ownership
 - divide an image to N * N cells (e.g. n = 19)
 - each object belongs to one cell
 - training / prediction targets for each cell

(note: each cell is learning itself based on nearby cells)
- Idea 5: Anchor Box
 - Each cell could have 1+ objects
- Idea 6: Non-Max Suppression
 - Remove predictions which have a higher IOU with a better prediction with the same class
- Accuracy Measure: IOU

$$IoU = \frac{\text{Area of Overlap}}{\text{Area of Union}}$$

→ Finally, Object Detection: YOLO



(source: <https://pjreddie.com/darknet/yolo/>)

Other Applications

- Face Recognition
- Image Style Transfer
- Generative Adversarial Networks