

Advanced Topics in Computer Science Lecture 3

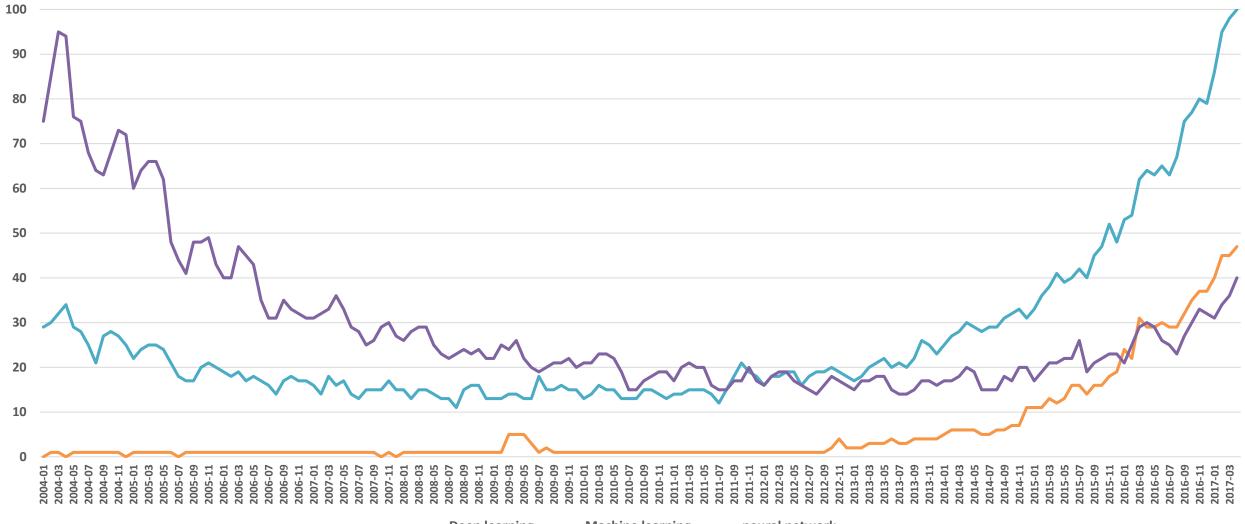


Convolutional Neural Networks

Dr Mohamed Loey Lecturer, Faculty of Computers and Information Benha University Eaypt

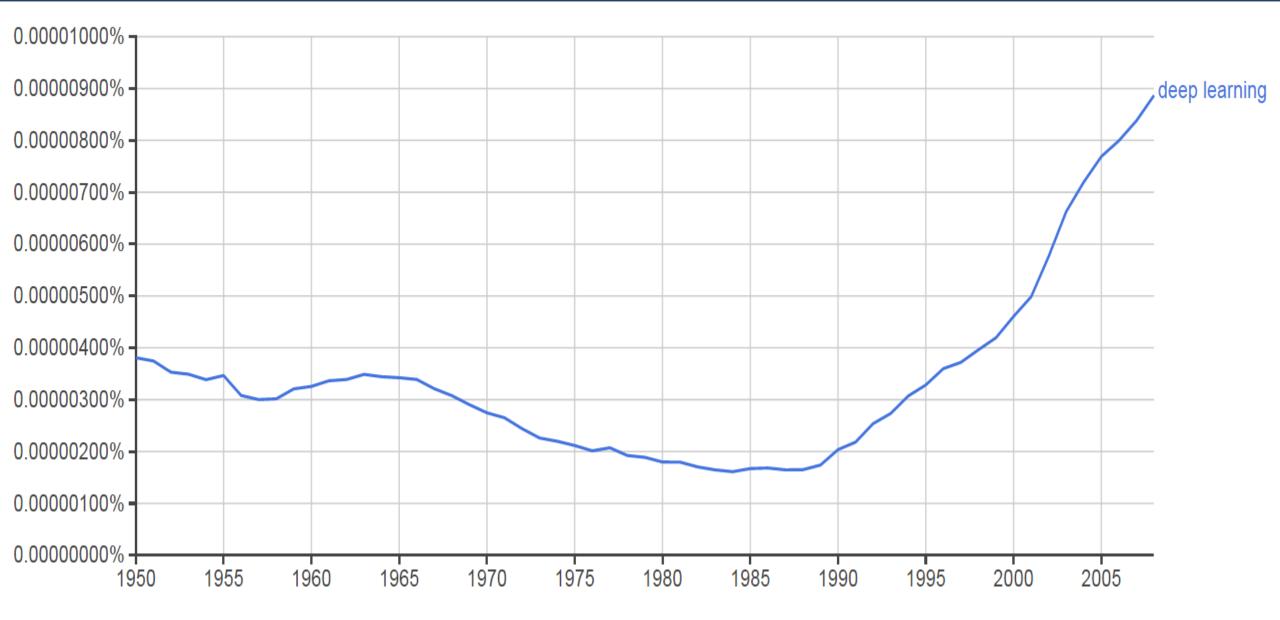
Google Trends

Google Trends



Deep learning — Machine learning — neural network

Google NGRAM



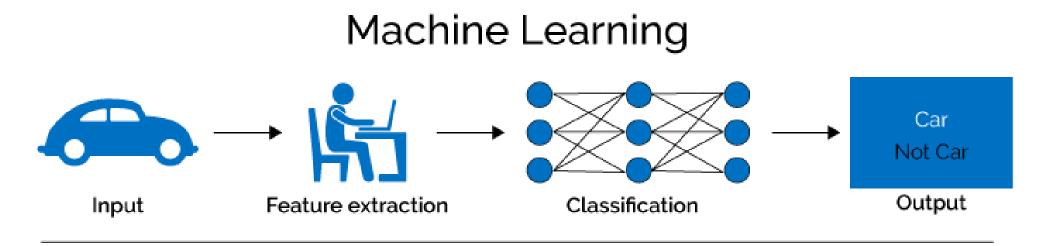
Google Queries

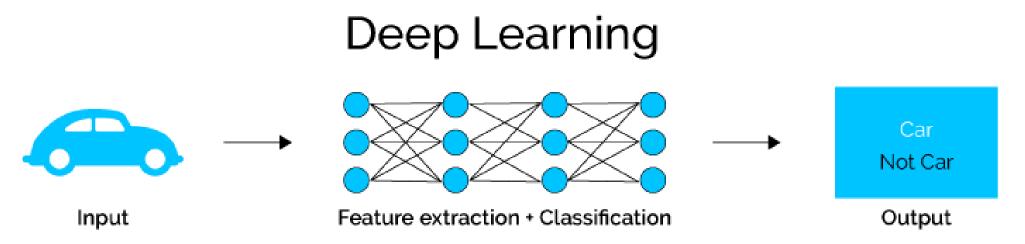
Queries	Тор	Rising
deep machine learning	100	
deep learning network	85	
google deep learning	75	
deep learning tutorial	75	
deep neural networks	60	
neural networks	60	
deep neural network	60	

What is Deep Learning?

- Part of the machine learning field of learning representations of data.
- Indicate the initial provides the second second
- □ If you provide the system tons of information, it begins to understand it and respond in useful ways.

Deep vs Machine Learning







Why we needs Deep Learning?

- **U** SuperIntelligent Devices
- □ Best Solution for
 - **D**image recognition
 - □ speech recognition
 - Inatural language processing

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Big Data

A brief History

1958 Perceptron	1974 Backpropagation	Convolution Neural Neur		ogle Brain Project on 16k Cores
	awkward silence (A	Al Winter)		
1969 Perceptron cr		1995 SVM reigns	2006 Restricted Boltzmann Machine	2012 AlexNet wins ImageNet IM GENET

Geoffrey Hinton: University of Toronto & Google

Yann LeCun: New York University & Facebook

Andrew Ng: Stanford & Baidu

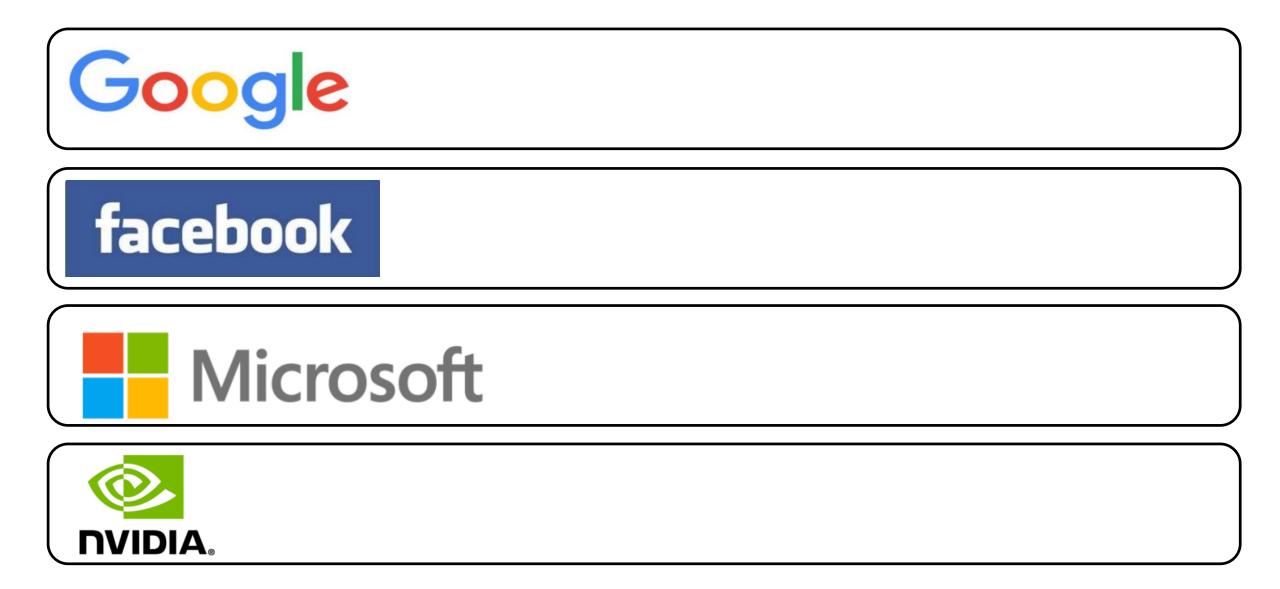






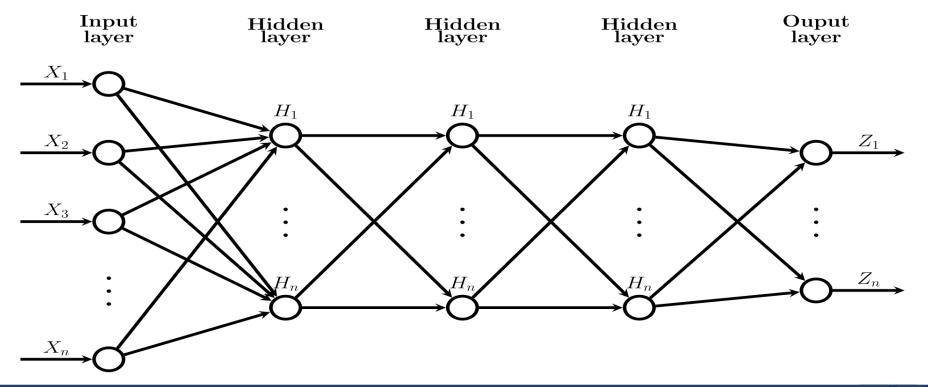


Superstar Companies





Deep learning (DL) is a hierarchical structure network which through simulates the human brain's structure to extract the internal and external input data's features



Deep Learning Requirements

- Large data set with good quality
- Measurable and describable goals
- Enough computing power
- Deural Network (Brain of Human)

Deep Learning Architectures

Deep Neural Networks

Deep Belief Networks

Convolutional Neural Networks

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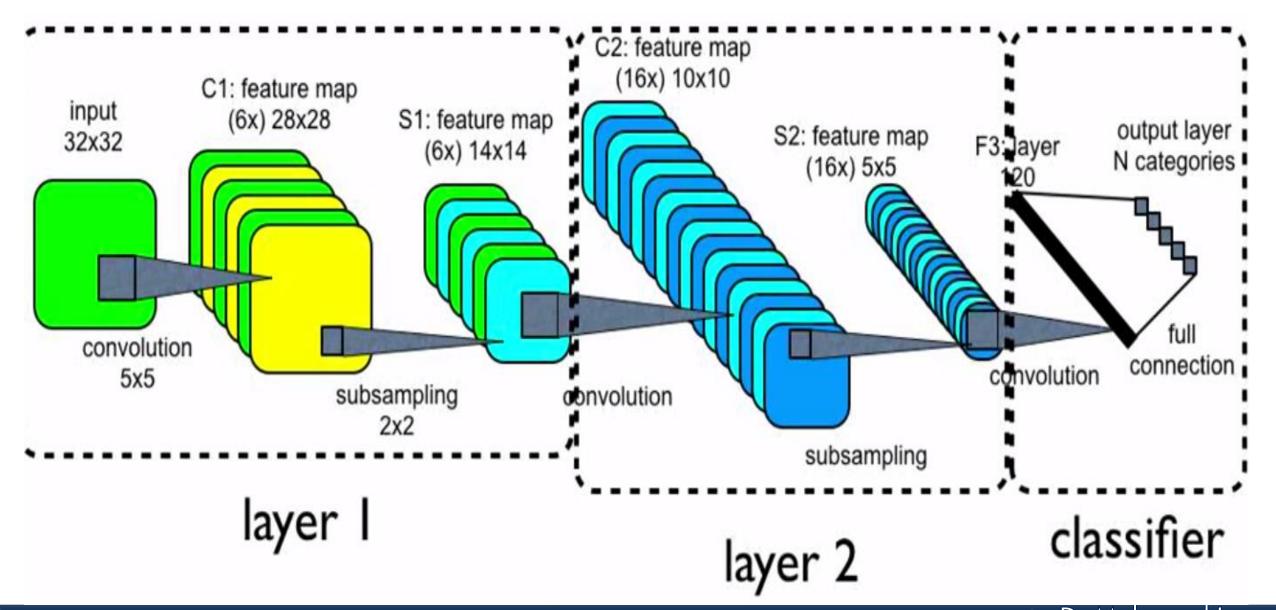
Deep Boltzmann Machines

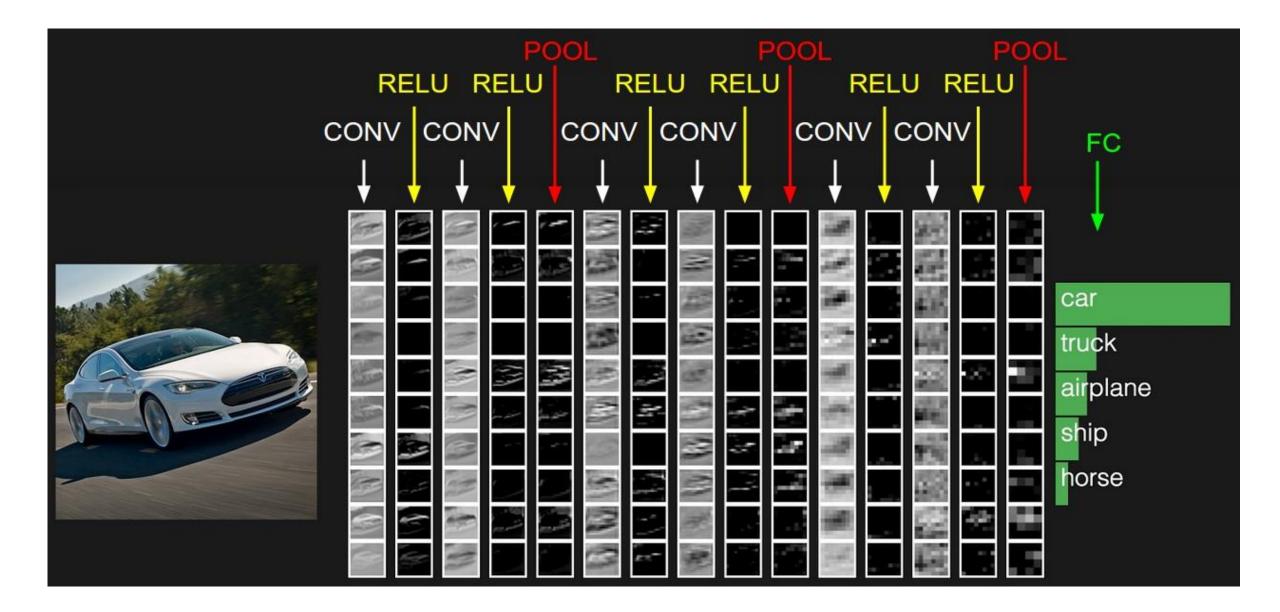
Deep Stacking Networks

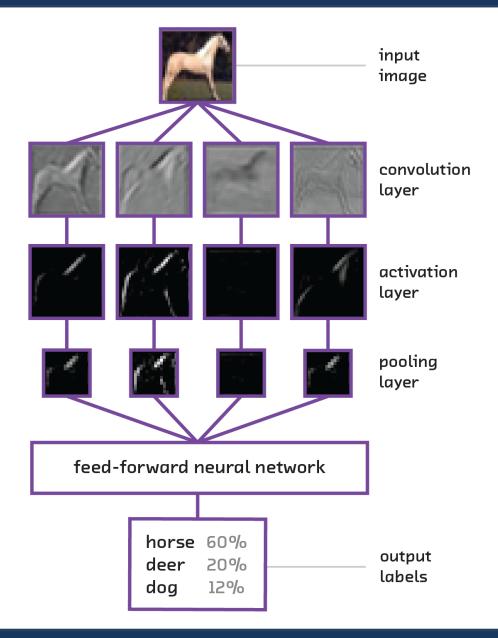
Convolution Neural Networks (CNN) is supervised learning and a family of multi-layer neural networks particularly designed for use on two dimensional data, such as images and videos.

□ A CNN consists of a number of layers:

- Convolutional layers.
- Pooling Layers.
- Fully-Connected Layers.



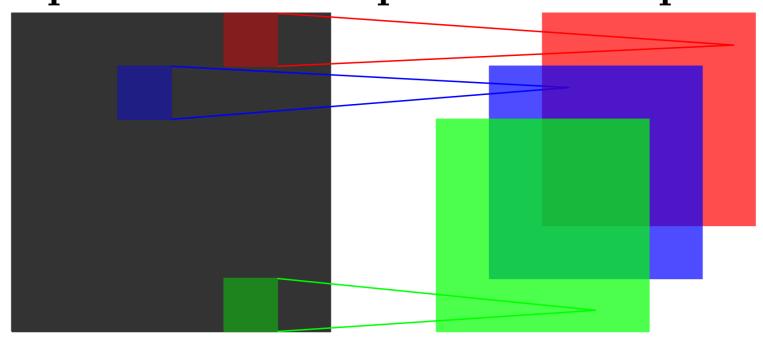




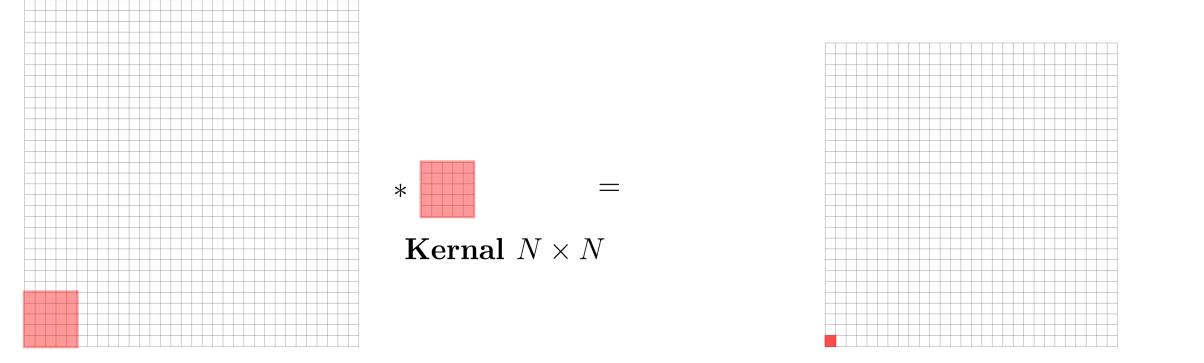


Convolutional layer acts as a feature extractor that extracts features of the inputs such as edges, corners, endpoints.

Input image Output feature or Input feature map maps



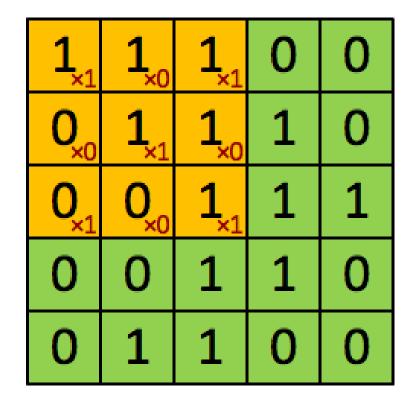


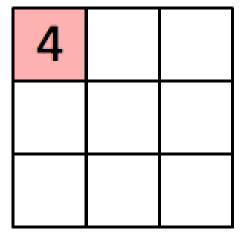


Input $M \times M$

CONV Layer $M - N + 1 \times M - N + 1$

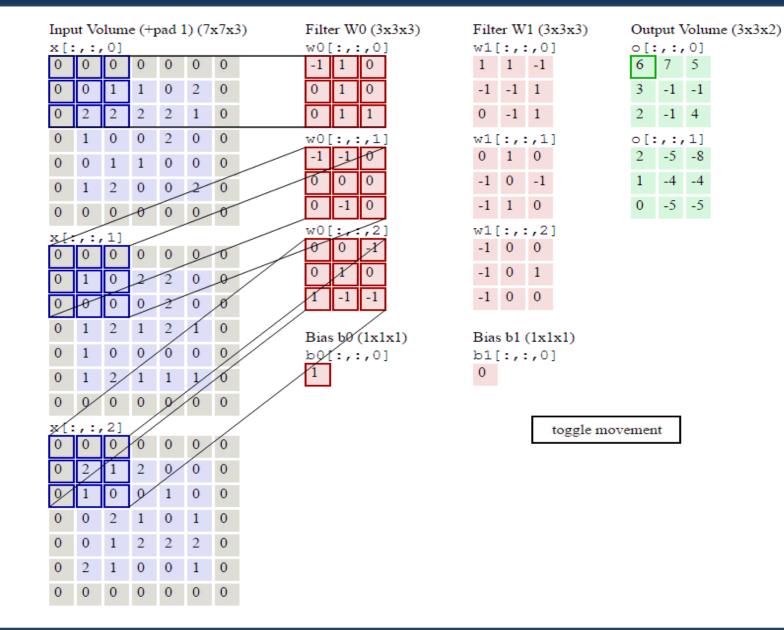






Image

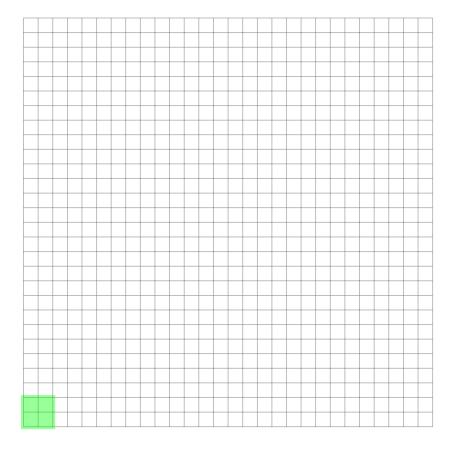
Convolved Feature



The pooling layer reduces the resolution of the image that reduce the precision of the translation (shift and distortion) effect. feature maps layer (l-1)feature maps layer l



Pooling layers



Feature map

$$M - N + 1 \times M - N + 1$$

POOL Layer $\frac{M-N+1}{2} \times \frac{M-N+1}{2}$

1	3	2	9
7	4	1	5
8	5	2	3
4	2	1	4

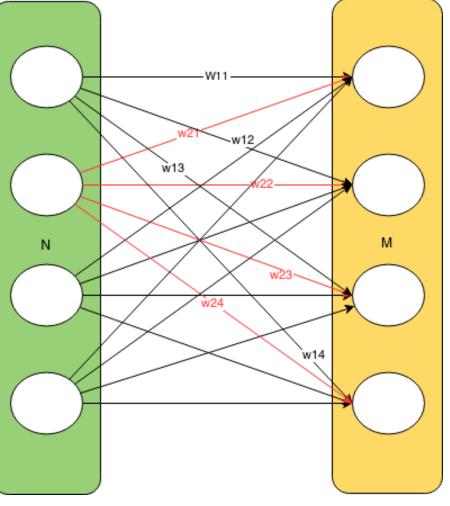
7	9
8	

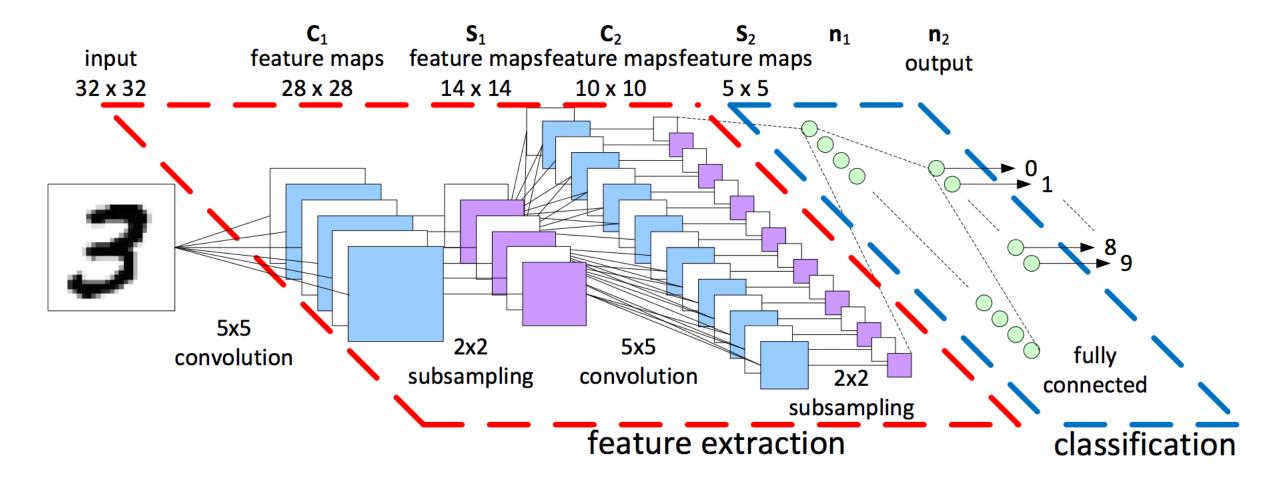


Fully-Connected Layers

u fully connected layer have full connections to all activations in the previous layer.

□ Fully connect layer act as classifier.





LeNet :The first successful applications of CNN

AlexNet: The ILSVRC 2012 winner

ZFNet: The ILSVRC 2013 winner

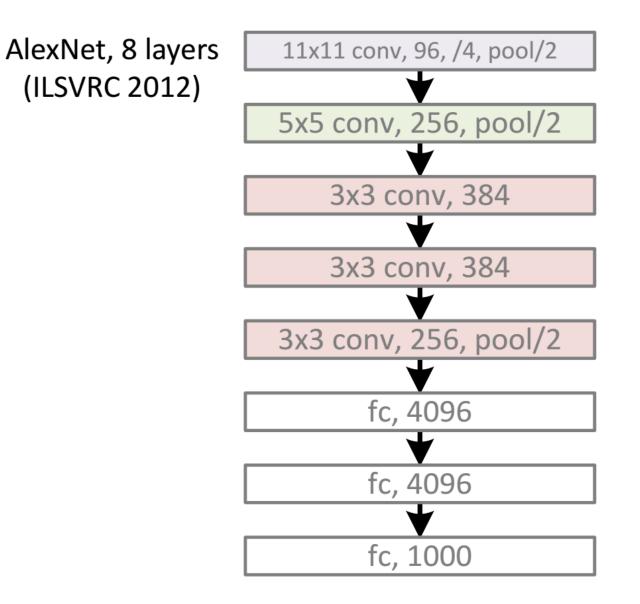
GoogLeNet: The ILSVRC 2014 winner

VGGNet: The runner-up in ILSVRC 2014

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ResNet: The winner of **ILSVRC** 2015

AlexNet



Revolution of Depth



11x11 conv, 96, /4, pool/2

5x5 conv, 256, pool/2

3x3 conv, 384

3x3 conv, 384
★
3x3 conv, 256, pool/2
∀
fc, 4096
★
fc, 4096
•
fc, 1000
*

VGG, 19 layers (ILSVRC 2014)	3x3 conv, 64 3x3 conv, 64, pool/2 3x3 conv, 128 ↓
	3x3 conv, 128, pool/2
	3x3 conv, 256
	3x3 conv, 256
	3x3 conv, 256
	V
	3x3 conv, 256, pool/2
	3x3 conv, 512
	▼
	3x3 conv, 512
	¥
	3x3 conv, 512
	3x3 conv, 512, pool/2
	5X5 CONV, 512, p001/2
	3x3 conv, 512
	
	3x3 conv, 512
	▼
	3x3 conv, 512
	3x3 conv, 512, pool/2
	*
	fc, 4096
	V
	fc, 4096
	fc, 1000
	10, 1000

GoogleNet, 22 layers terres and a stress antes attack (ILSVRC 2014) COLUMN TRACTOR ATTER ATTER ATTER arran arran arran arran tating tating and terrings Section Laboratory terring berring berring berrings tantage tantage tantage ting ting ting ting terring terrings terrings terrings LITTLE LITTLE DEFENS ten autriet terriet terriet unite unite and

AlexNet, 8 layers (ILSVRC 2012)



VGG, 19 layers (ILSVRC 2014) ResNet, 152 layers (ILSVRC 2015) **MNIST** Handwritten digits – 60000 Training + 10000 Test Data

Google House Numbers from street view - 600,000 digit images

CIFAR-10 60000 32x32 colour images in 10 classes

IMAGENET 1.2 million images, >150 GB

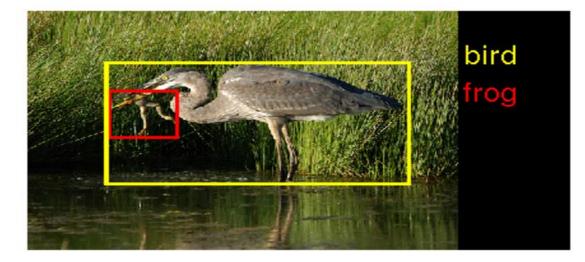
Tiny Images 80 Million tiny images

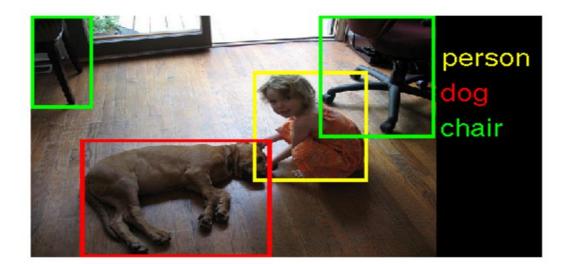
Flickr Data 100 Million Yahoo dataset

□The ImageNet Large Scale Visual Recognition Challenge (ILSVRC) evaluates algorithms for object detection and image classification at large scale.

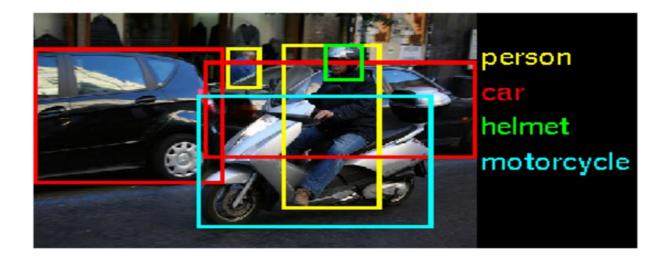


ILSVRC 2014

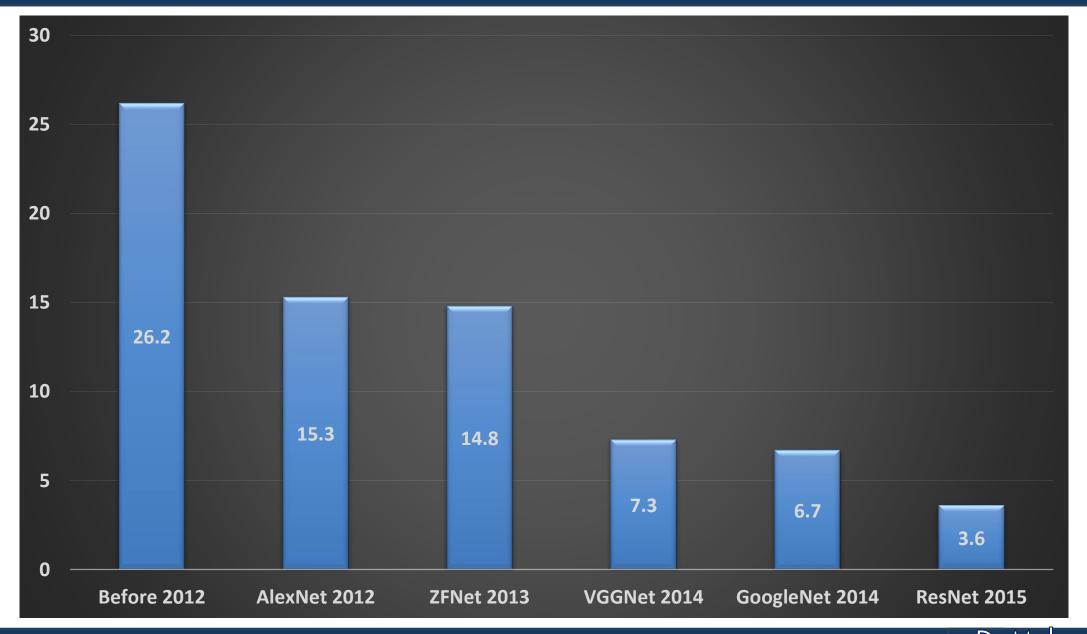








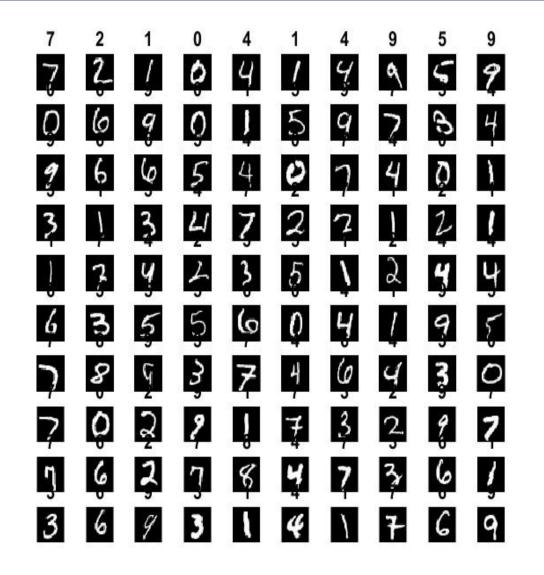
Top-5 error rates on ImageNet



MNIST Dataset

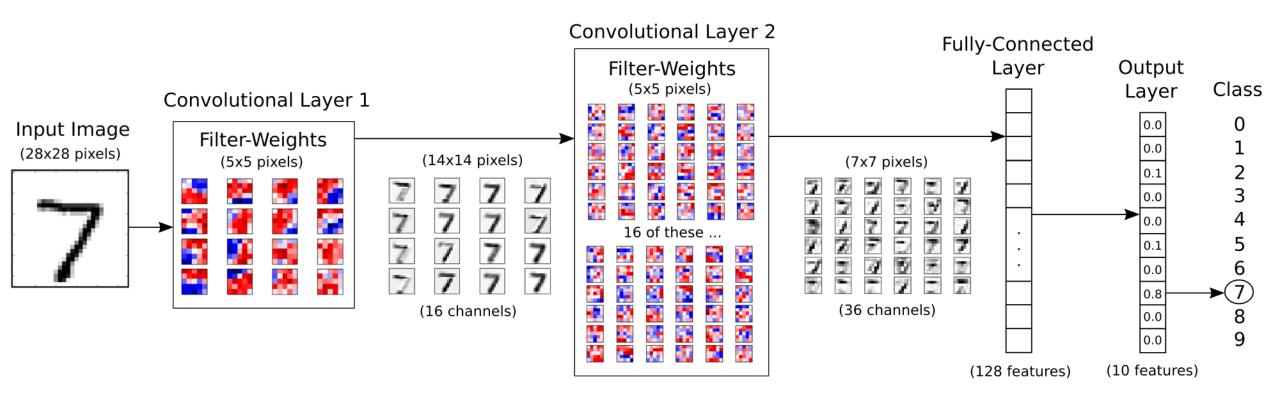
MNIST is a large database of handwritten digits.

MNIST contains 60,000 training
 images and 10,000 testing
 images



MNIST Dataset

CNN on MNIST Dataset



CIFAR-10 Dataset

CIFAR-10 dataset consists 60000 32x32 colour of images in 10 classes CIFAR-10 contains 50000 training images and 10000 test images

bird

cat

deer

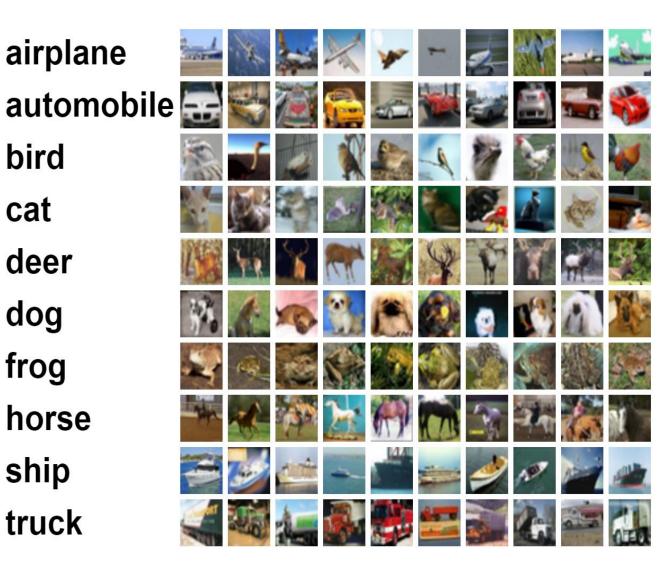
dog

frog

horse

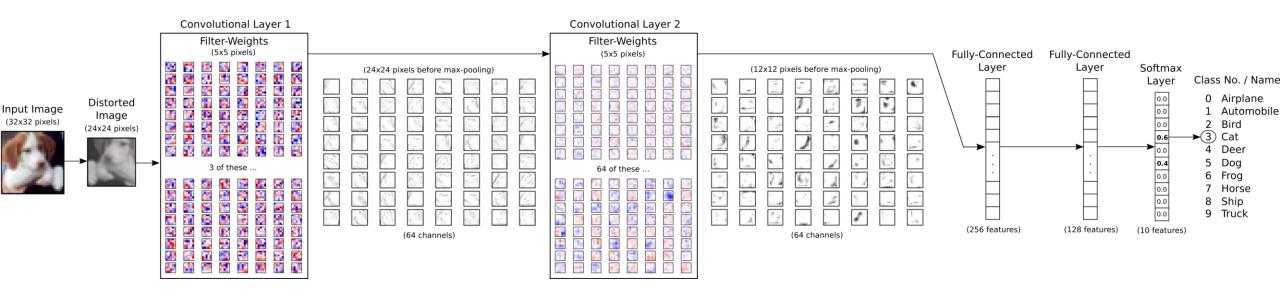
ship

truck



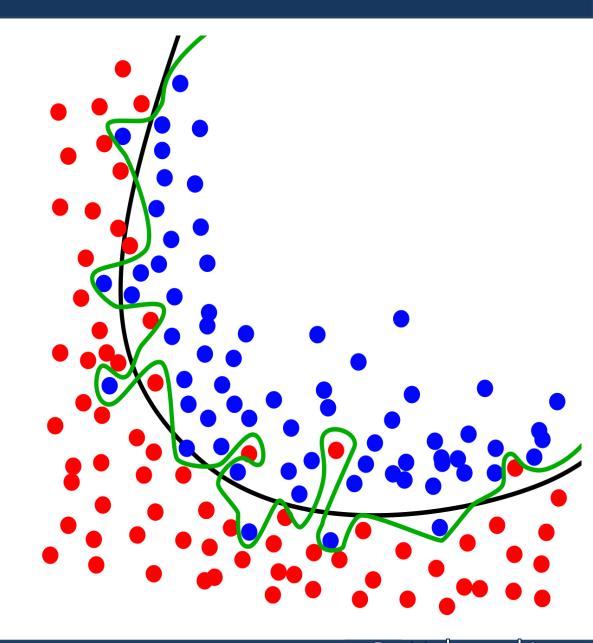
CIFAR-10 Dataset

CNN on CIFAR-10 Dataset



Overfitting Problem

- Larger network have a lots of weights this lead to high model complexity
- Network do excellent on training data but very bad on validation data

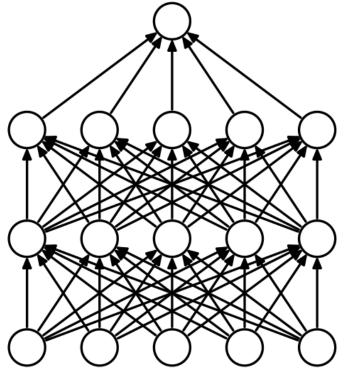


CNN Optimization used to reduce the overfitting problem in CNN by:

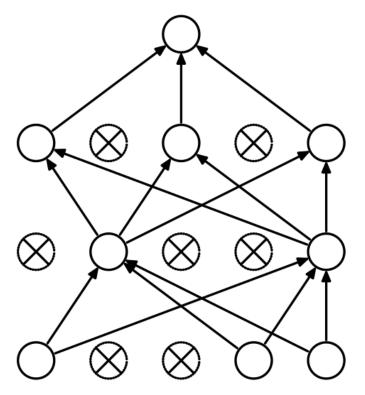
- 1) Dropout
- 2) L2 Regularization
- 3) Mini-batch
- 4) Gradient descent algorithm
- 5) Early stopping
- 6) Data augmentation



Dropout is a technique of reducing overfitting in CNN.



(a) Standard Neural Net



(b) After applying dropout.

L2 Regularization: Adding a regularization term for the weights to the loss function is a way to reduce overfitting.

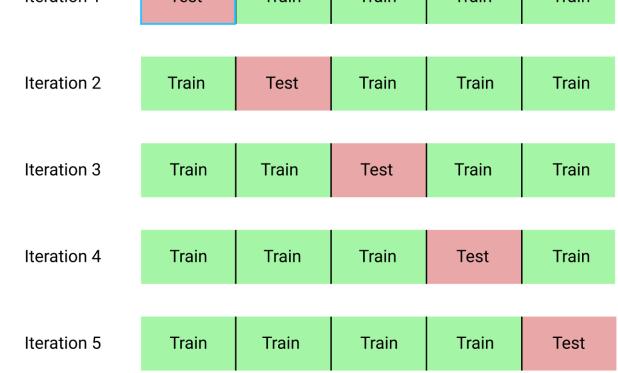
$$E_R(P_i) = E(P_i) + \lambda \Omega(w)$$

 \Box where w is the weight vector, λ is the regularization factor (coefficient), and the regularization function, $\Omega(w)$ is:

$$\Omega(w) = \frac{1}{2}w^t w$$

Mini-batch is to divide the dataset into small batches of examples, compute the gradient using a single batch, make an update, then move to the next batch.

 Iteration 1
 Test
 Train
 Train
 Train

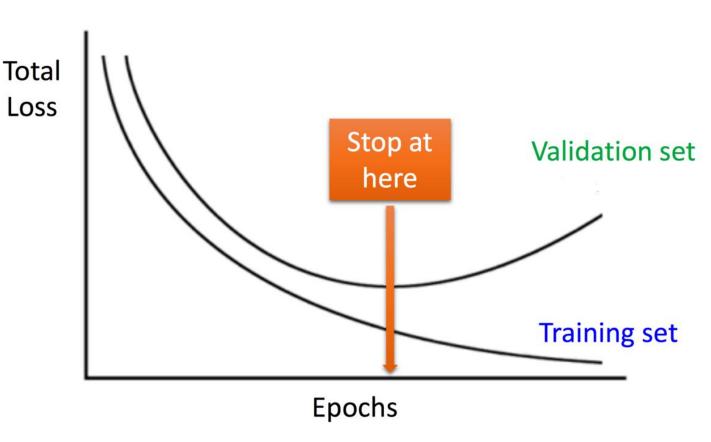


□ The gradient descent algorithm updates the coefficients (weights and biases) so as to minimize the error function by taking small steps in the direction of the negative gradient of the loss function

$$P_{i+1} = P_i - \alpha \nabla E(P_i)$$

 \Box where i stands for the iteration number, $\alpha > 0$ is the learning rate, P is the parameter vector, and E(Pi) is the loss function.

Early stopping monitoring the deep learning process of the network from overfitting. there is no more improvement, or worse, the performance on the test set degrades, then the learning process is aborted



Data augmentation means increasing the number of dataset.



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TITAN X In Research				
Deep Learning	Augmented Reality			
Machine Learning	Image Recognition			
Computer Vision	Data Science			

Deep learning is a class of machine learning algorithms.

□ Harder problems such as video understanding, image understanding, natural language processing and Big data will be successfully tackled by **deep learning algorithms**.

Each student select one of Benchmark Dataset from https://goo.gl/DNQmtj

1 1 5 4 3 7 5 3 5 3 MNIST 50 results collected

55906

Classify handwriten digits. Some additional results are available on the original dataset page.

ResultMethodVenueDetails0.21%Regularization
of Neural
Networks using
DropConnectICML 2013

Task3

CIFAR-10

who is the bes	st in CIFAR-10 ?	89.14%	Deep Convolutional Neural Networks	IJCNN 2015	Details
			as Generic Feature Extractors		
	CIFAR-10 49 results collected Units: accuracy % Classify 32x32 colour images.	89%	ImageNet Classification with Deep Convolutional Neural Networks	NIPS 2012	Details

Download the paper that describe the method used by the authors.

□ Make one page using word to summarize your selected paper.

Use MS Word

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Advanced Topics in CS2 – Task3 "

□Put your **Arabic name** on word and email body

Given Finally, press Send

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