

# Neural Networks and Convolutional Neural Network

## [Spring 2020 CS-8395-02 Deep Learning in Medical Image Computing]

Instructor: Yuankai Huo, Ph.D.  
Department of Electrical Engineering and Computer Science  
Vanderbilt University

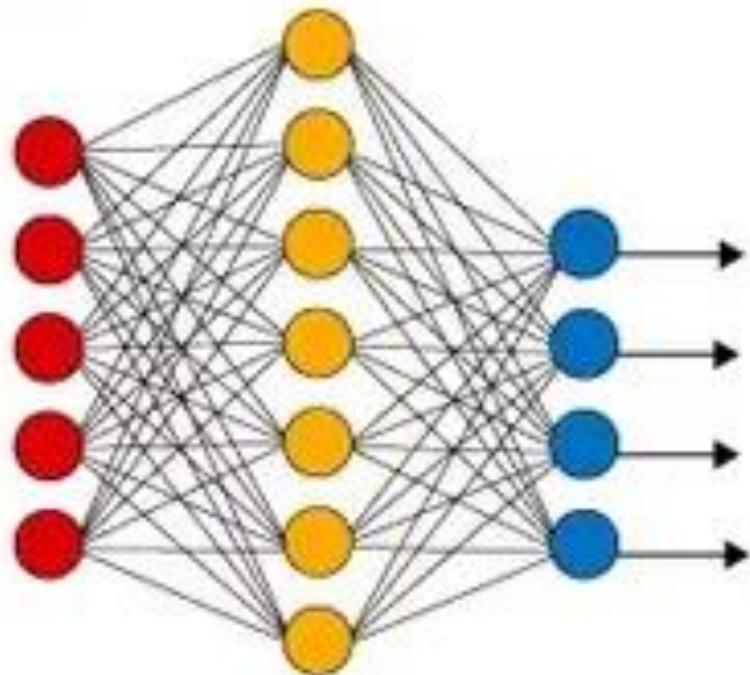
# Topics



- Review
- Neural Network
- Convolutional Neural Network

# Deep Learn = Deep Neural Network

**Simple Neural Network**

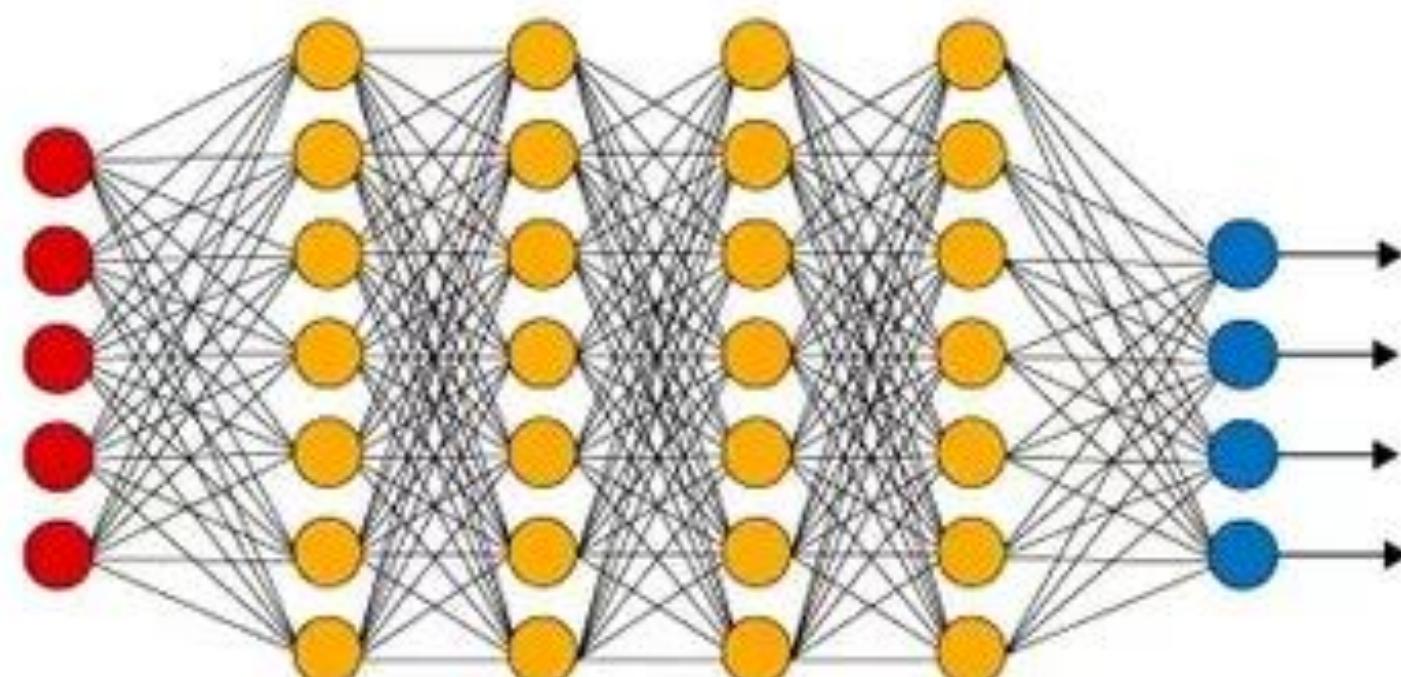


● Input Layer

○ Hidden Layer

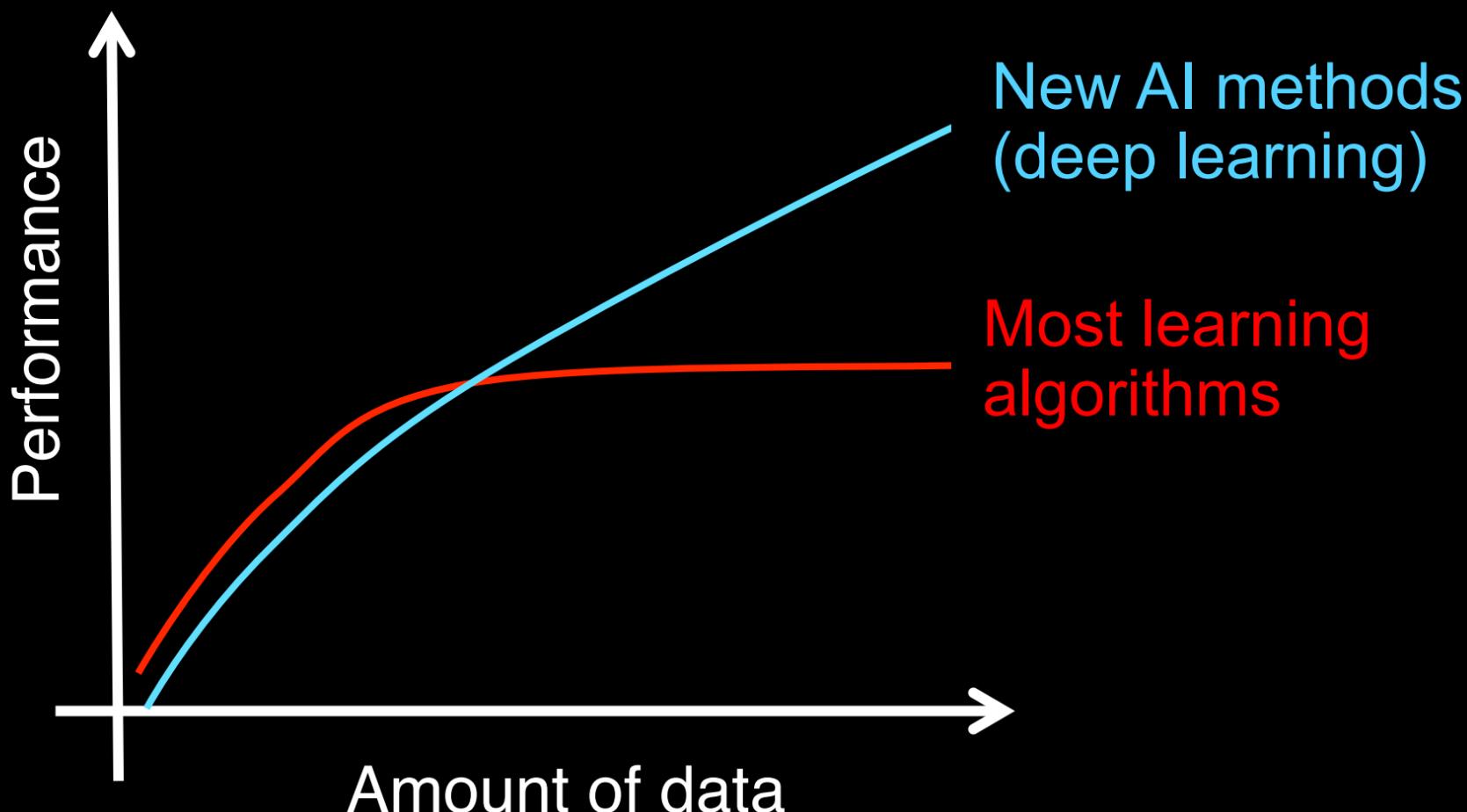
● Output Layer

**Deep Learning Neural Network**

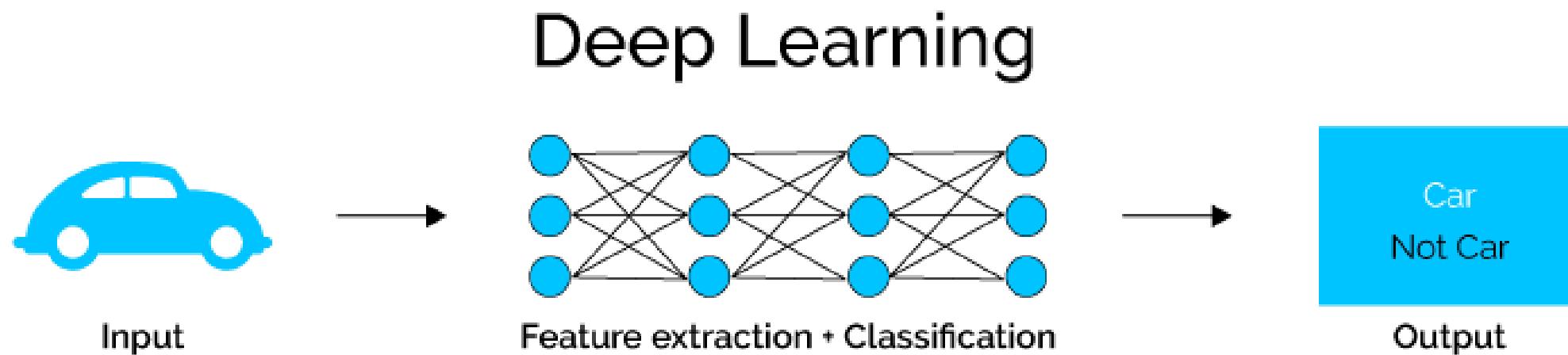
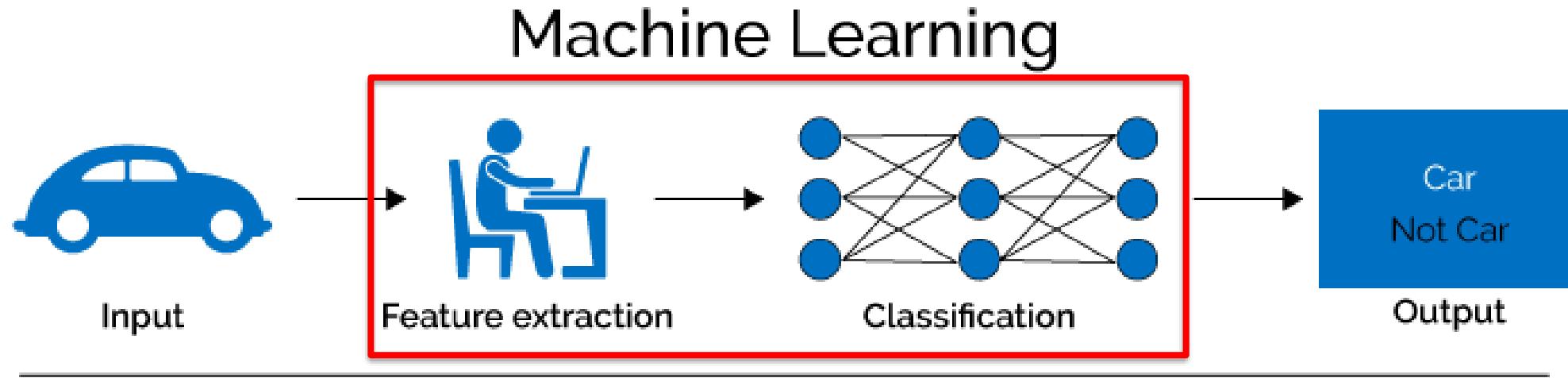


<https://www.quora.com/What-is-the-difference-between-Neural-Networks-and-Deep-Learning>

# Advantages

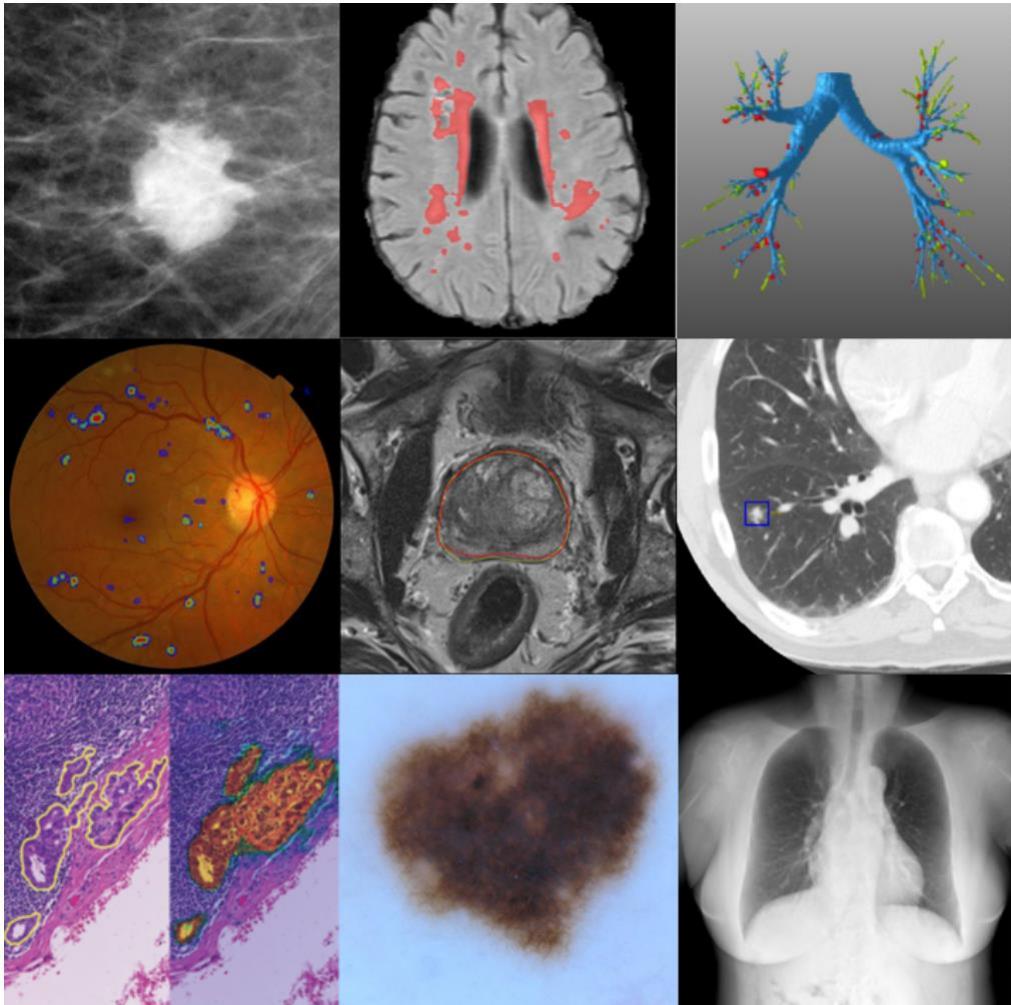


# Deep Learning vs. Machine Learning



<https://www.quora.com/How-do-I-start-learning-machine-learning-and-deep-learning-using-C++>

# Applications



Some medical imaging applications in which deep learning has achieved state-of-the-art results. From top-left to bottom-right:

- mammographic mass classification (Kooi et al., 2016)
- segmentation of lesions in the brain (Ghafoorian et al. (2016b))
- leak detection in airway tree segmentation (Charbonnier et al., 2017)
- diabetic retinopathy classification (Kaggle challenge 2015)
- image from van Grinsven et al. (2016)
- prostate segmentation (top rank in PROMISE12 challenge)
- nodule classification (top ranking in LUNA16 challenge)
- breast cancer metastases detection in lymph nodes (CAMELYON16)
- human expert performance in skin lesion classification (Esteva et al., 2017)
- state-of-the-art bone suppression in x-rays, image from Yang et al. (2016c).

<https://arxiv.org/abs/1702.05747>

# Organization of Course



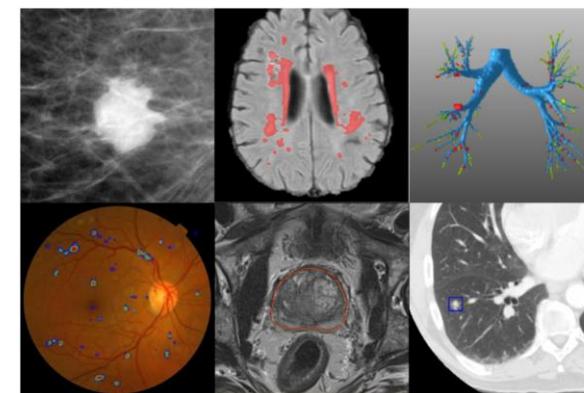
## Overview

Overview of Deep Learning in Medical Image Computing  
Neural Networks and CNN

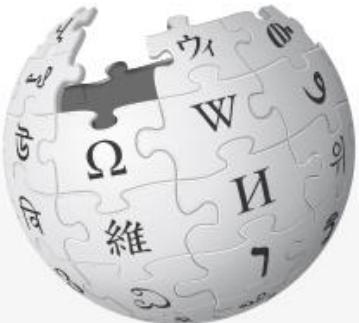
## Key Techs

Classification (Medical Image Diagnosis)  
Detection (Landmark Localization and Detection)  
Segmentation (Medical Image Segmentation)  
GAN (Medical Image Synthesis)

## Topics in Medical Image Computing



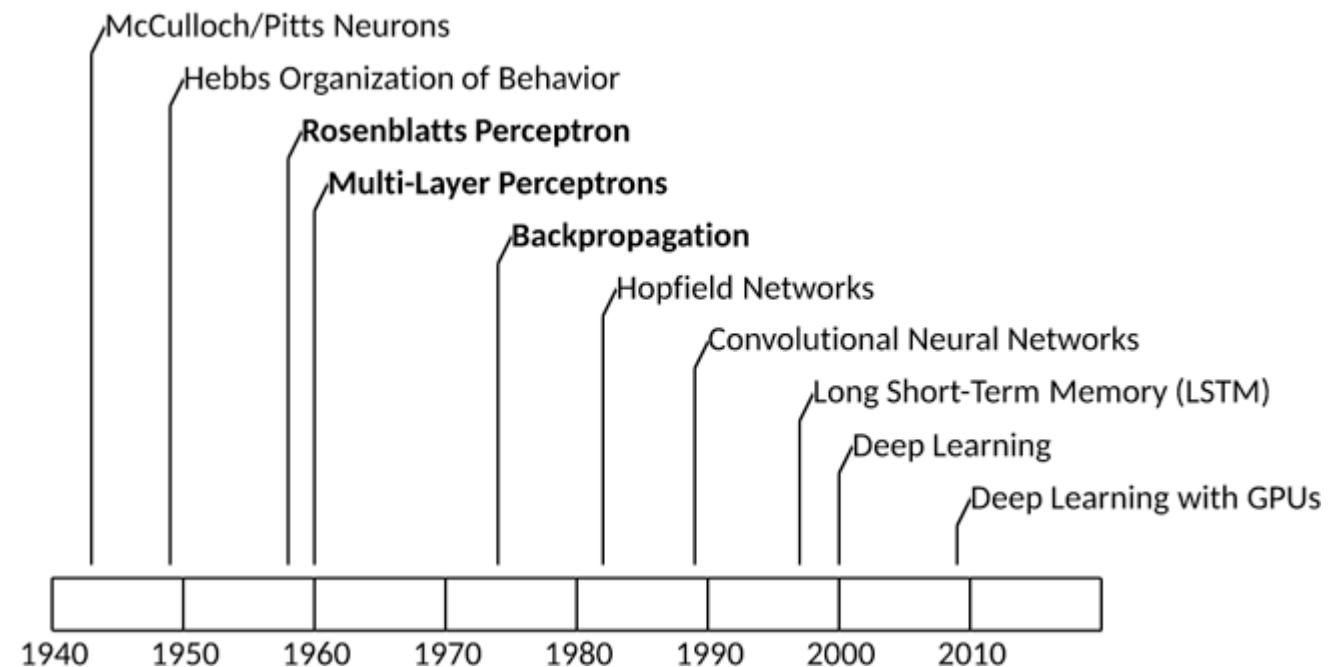
# Definition



**WIKIPEDIA**  
The Free Encyclopedia

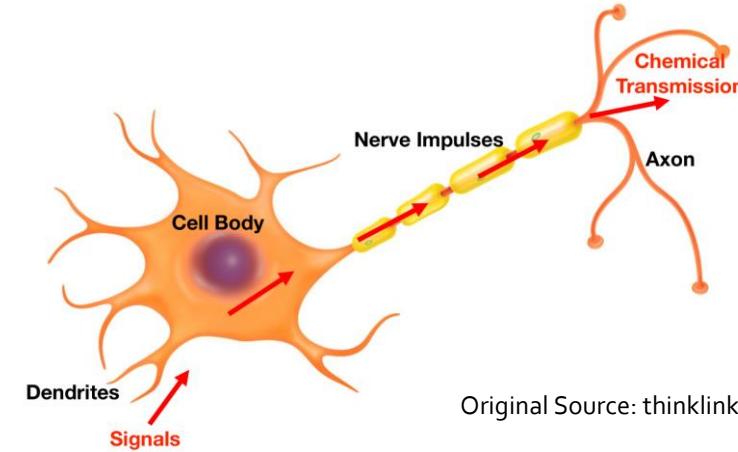
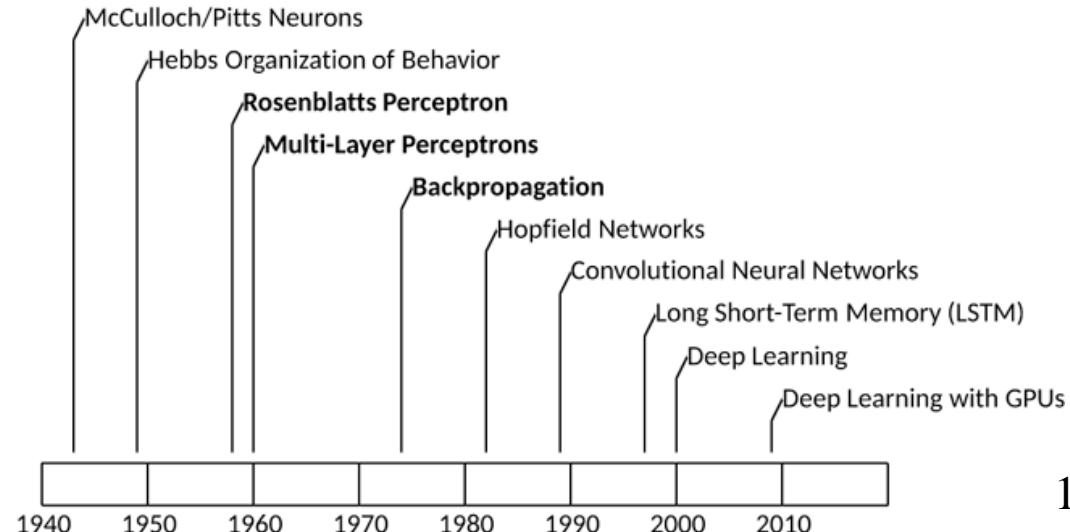
**Artificial neural networks (ANN):** computing systems vaguely inspired by the biological neural networks that constitute animal brains. Such systems "learn" to perform tasks by considering examples, generally without being programmed with any task-specific rules.

# History



<https://www.ibm.com/developerworks/library/cc-cognitive-neural-networks-deep-dive/index.html>

# McCulloch/Pitts Neurons

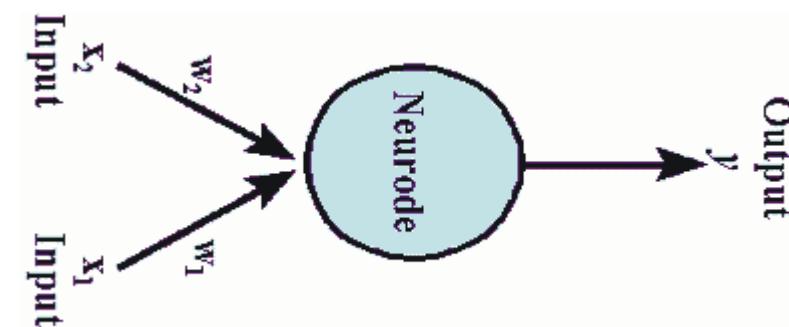


Original Source: thinklink.com

1943 paper, "A Logical Calculus of Ideas Immanent in Nervous Activity,"

In the early 1940s, McCulloch and Pitts created a computational model for neural networks that spawned research not only into the brain but also its application to artificial intelligence.

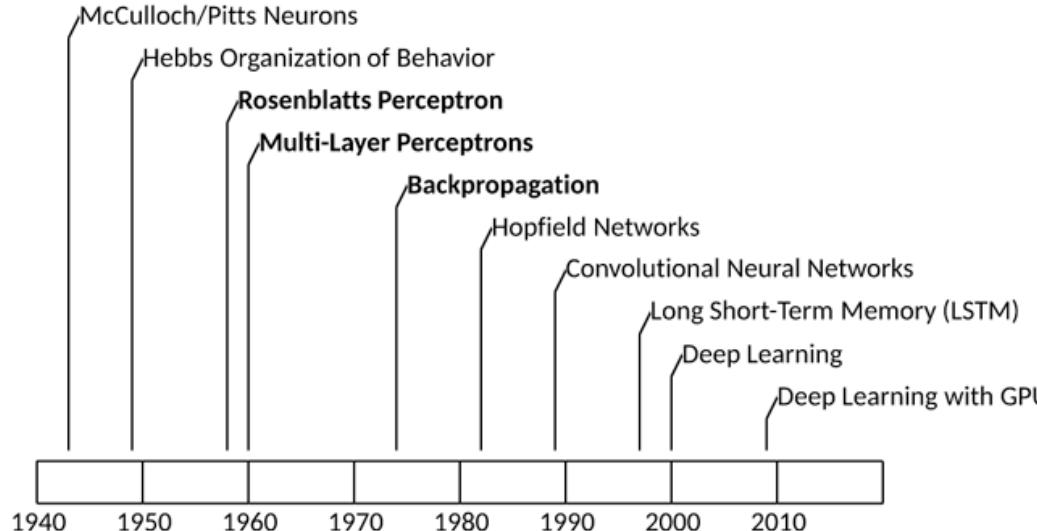
One of the difficulties with the McCulloch-Pitts neuron was its simplicity. It only allowed for binary inputs and outputs, it only used the threshold step activation function and it did not incorporate weighting the different inputs.



<https://www.ibm.com/developerworks/library/cc-cognitive-neural-networks-deep-dive/index.html>

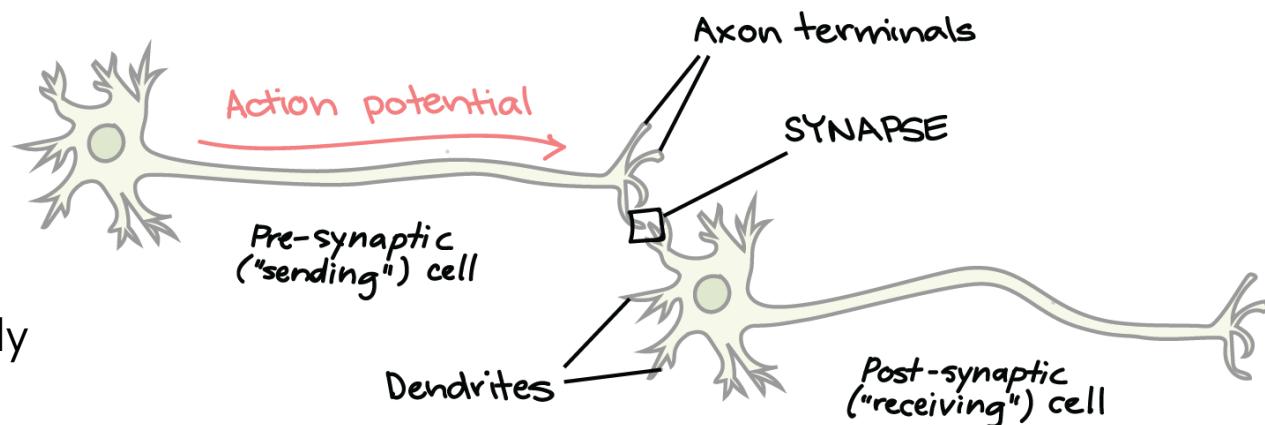
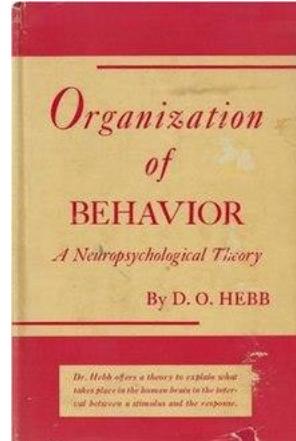
<https://web.csulb.edu/~cwallis/artificialn/History.htm>

# Hebb's Organization of Behavior

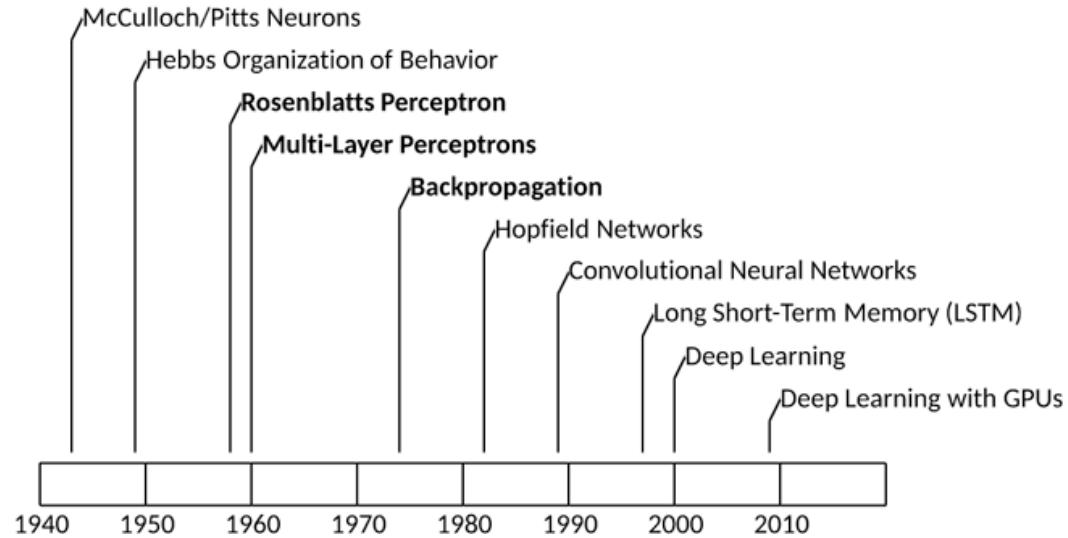


Later in this decade, Donald Hebb created *Hebbian learning*, which observed from biology that the synapse between two neurons is strengthened if the two neurons are simultaneously active.

In his book, *The Organization of Behavior*  
“When an axon of cell A is near enough to excite a cell B and repeatedly or persistently takes part in firing it, some growth process or metabolic change takes place in one or both cells such that A’s efficiency, as one of the cells firing B, is increased.”

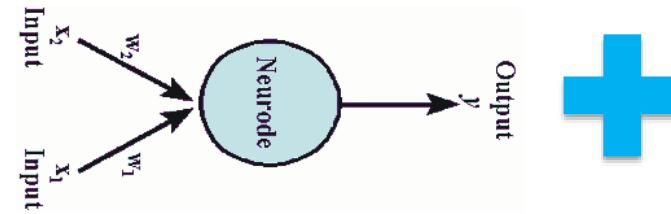


# Rosenblatt's Perceptron

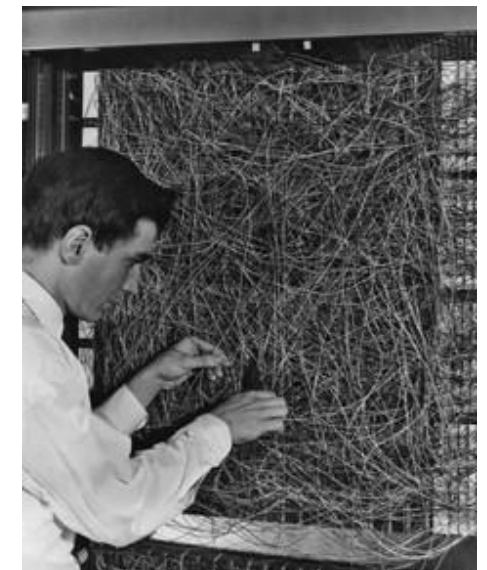
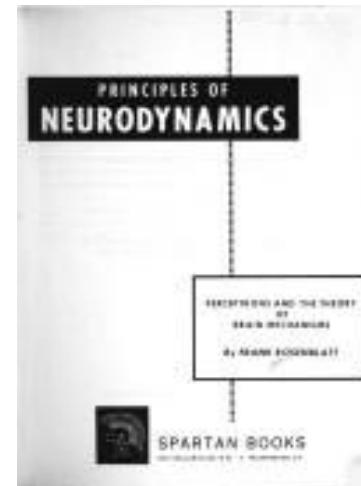


In 1958, Frank Rosenblatt created the perceptron, a simple neural model that could be used to classify data into two sets. However, this model suffered in that it could not correctly classify an exclusive-OR.

<https://www.ibm.com/developerworks/library/cc-cognitive-neural-networks-deep-dive/index.html>

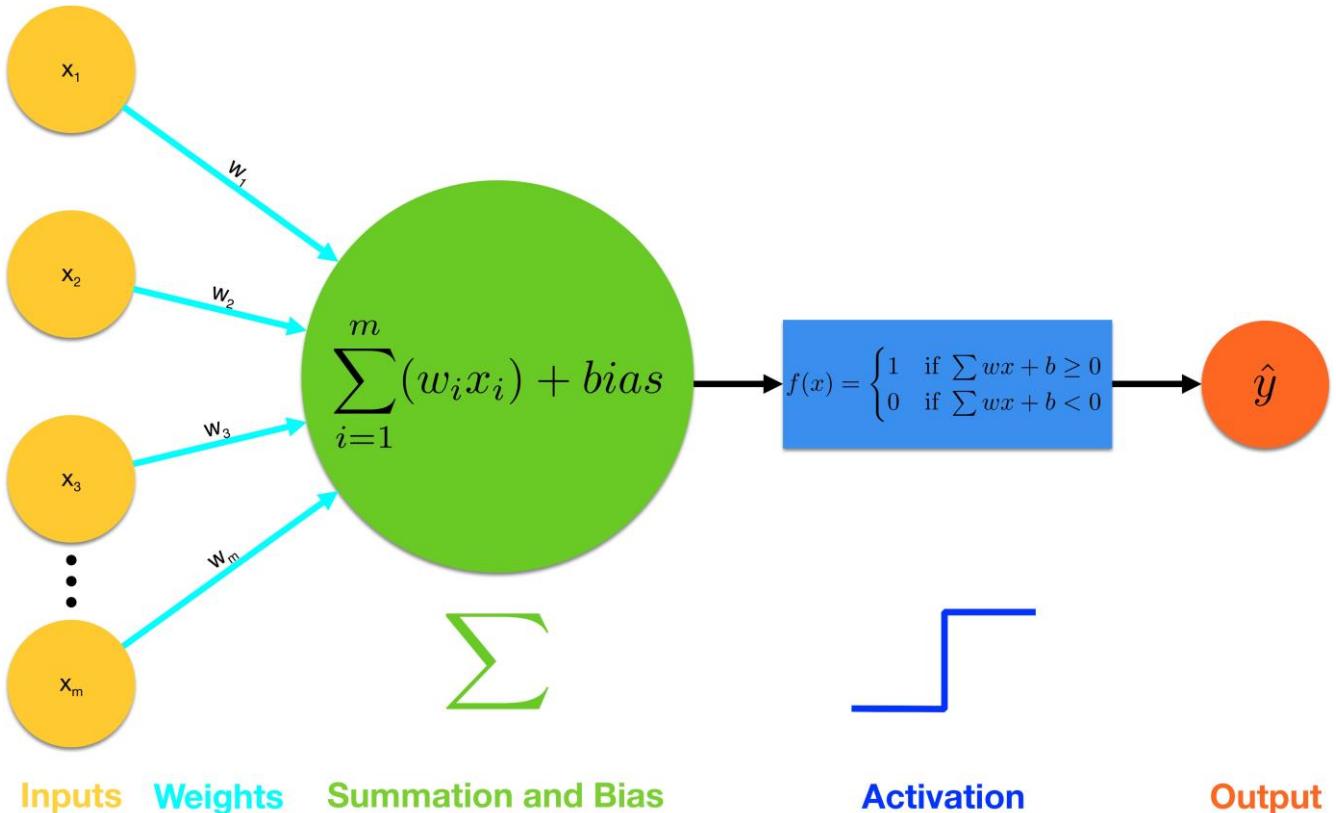


He discussed the perceptron in his 1962 book, *Principles of Neurodynamics*.



<https://web.csulb.edu/~cwallis/artificialn/History.htm>  
Rosenblatt and Perceptron, Source: The New Yorker  
<http://ahuakang.com/page3/>

# Perceptron



1. Inputs are fed into the perceptron
2. Weights are multiplied to each input
3. Summation and then add bias
4. Activation function is applied. Note that here we use a step function, but there are other more sophisticated activation functions like **sigmoid**, **hyperbolic tangent (tanh)**, **rectifier (relu)** and more. No worries, we will cover many of them in the future!
5. Output is either triggered as 1, or not, as 0. Note we use  $\hat{y}$  to label output produced by our perceptron model

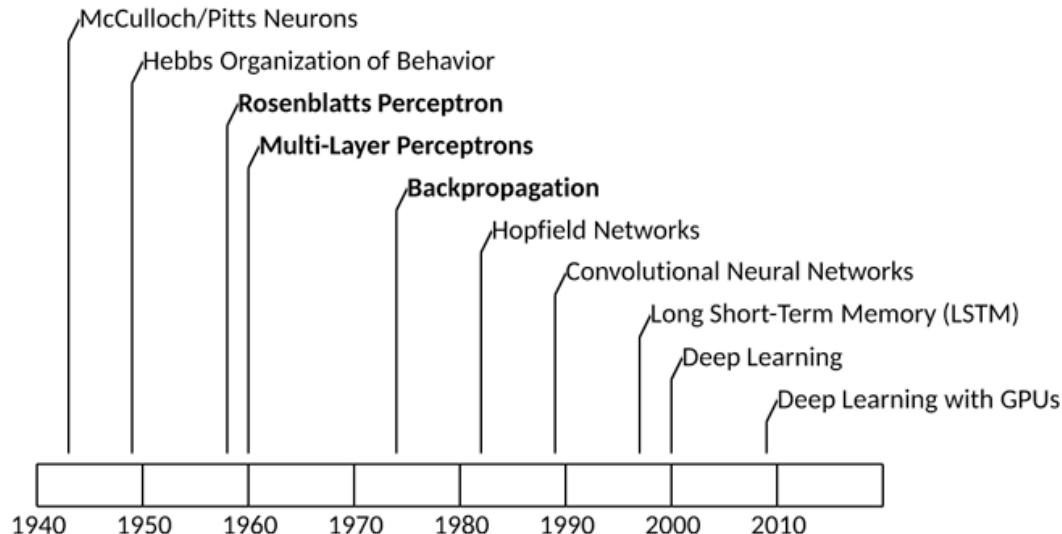
# Perceptron Example

You need to decide whether to study DL or not?

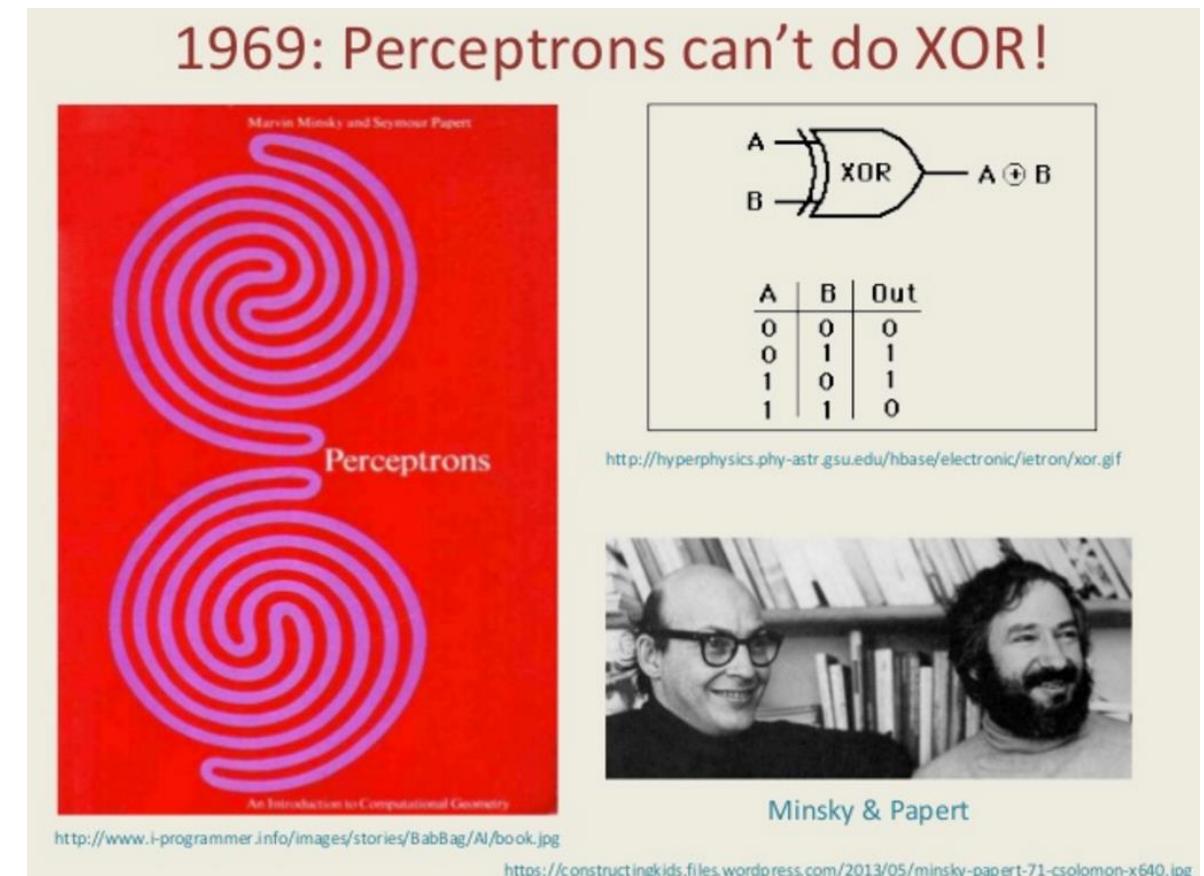
There are 3 factors that influence your decision:

- X<sub>1</sub> 1. If you will earn more money after mastering DL (Yes: 1, No: 0)
- X<sub>2</sub> 2. Is the relevant mathematics and programming easy (Yes: 1, No: 0)
- X<sub>3</sub> 3. You can work on DL immediately without the need for an expensive GPU  
(Yes: 1, No: 0)

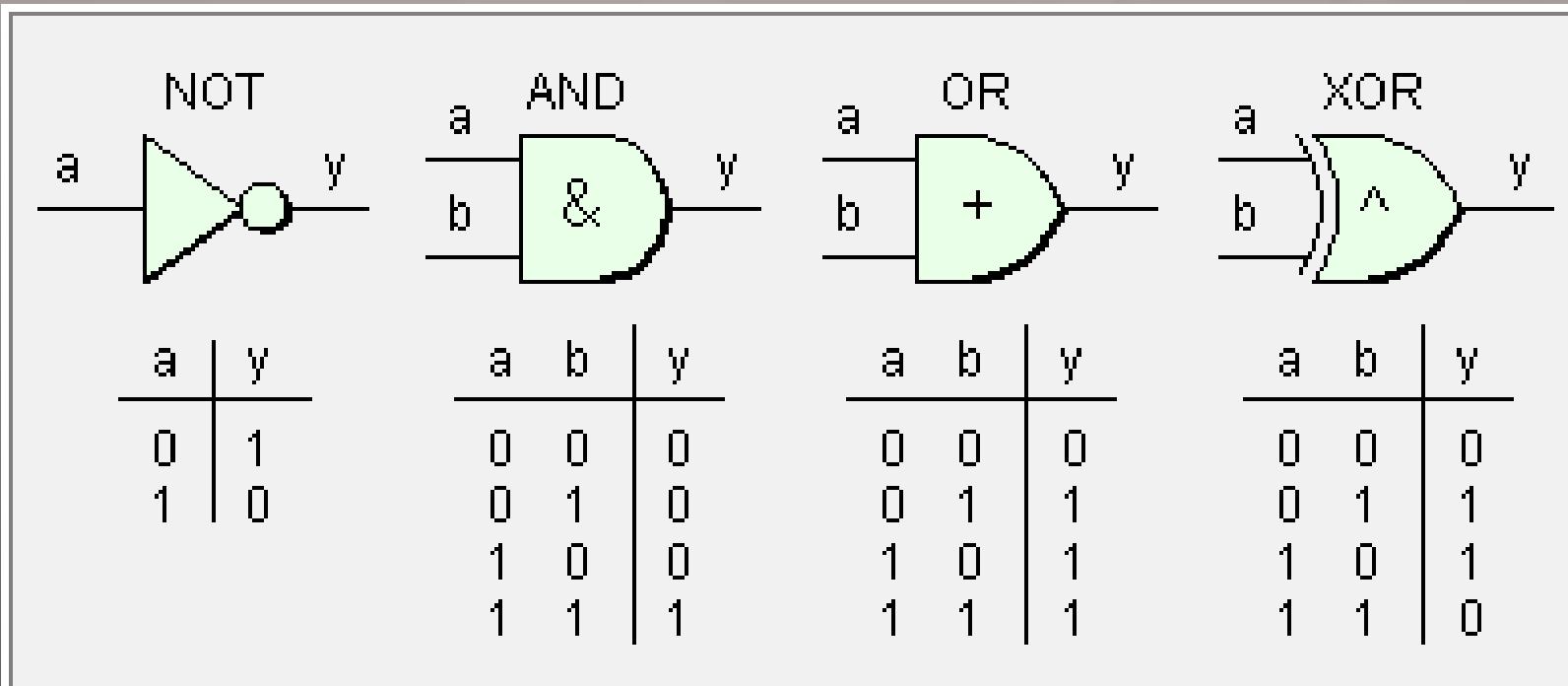
# AI Winter



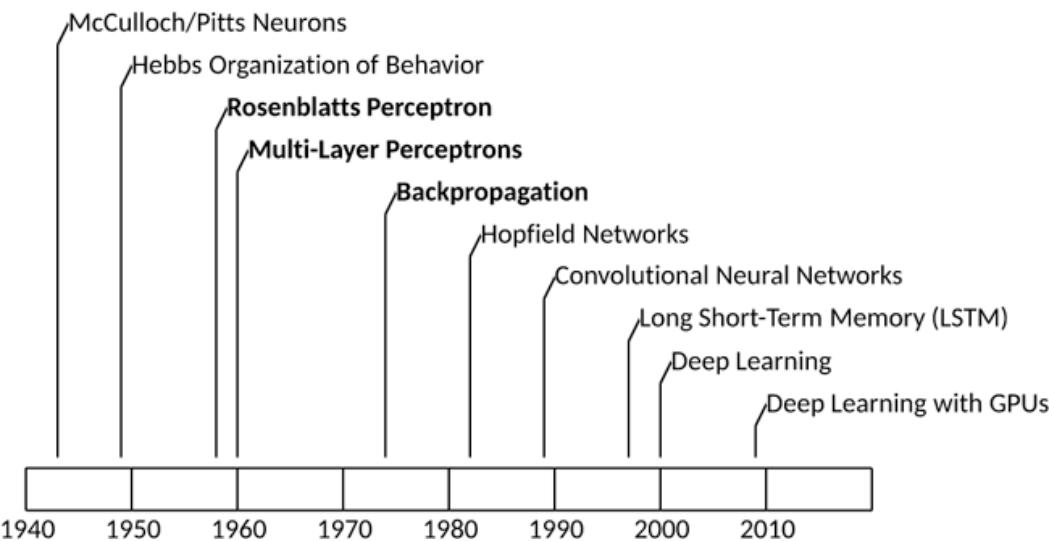
Marvin Minsky and Seymour Papert exploited this limitation in 1969 in their book *Perceptrons* in an attempt to return focus to symbolic methods for AI. The result was a decade-long decline in connectionist research funding. A



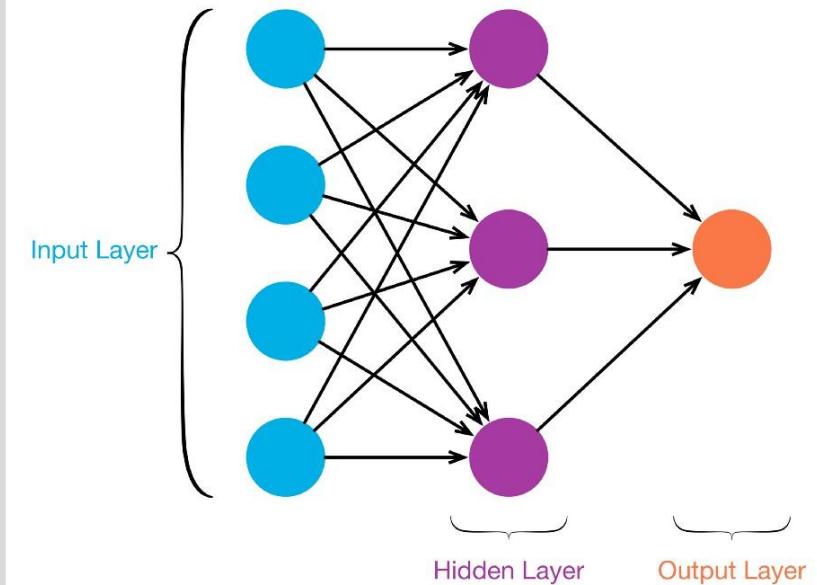
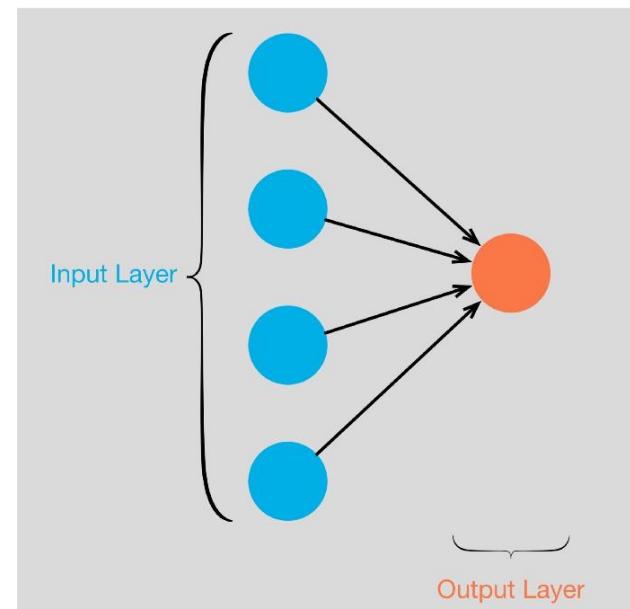
# Design XOR?



# Multi-Layer Perceptrons



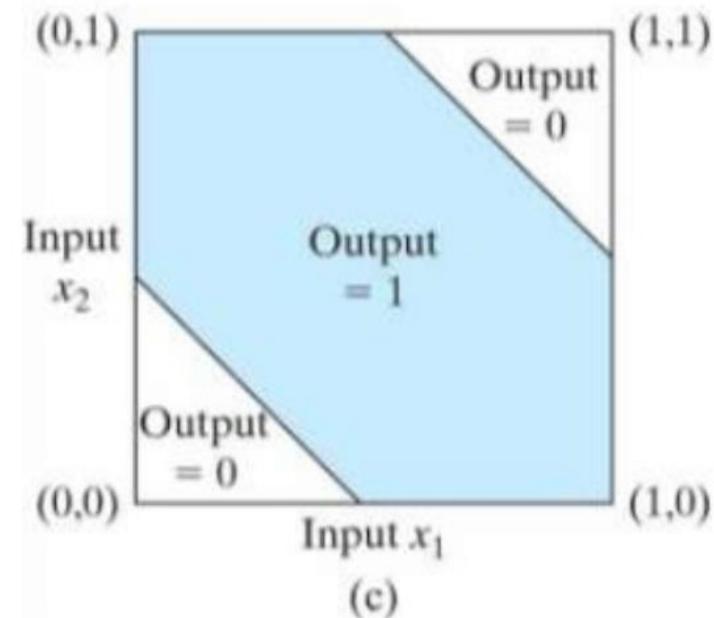
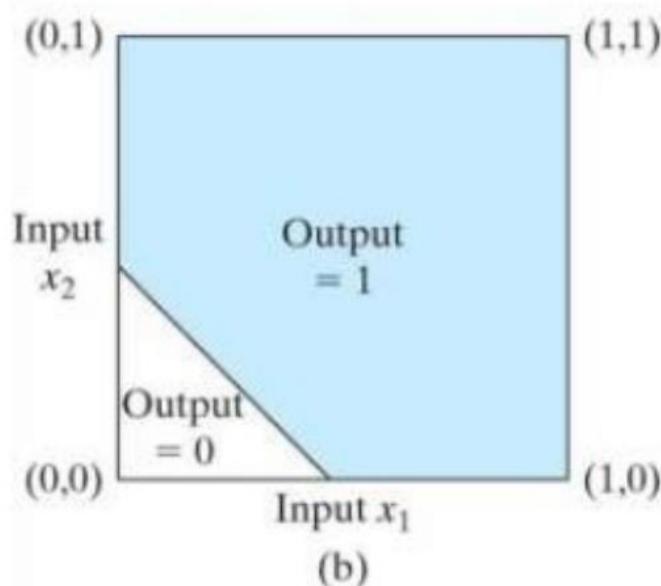
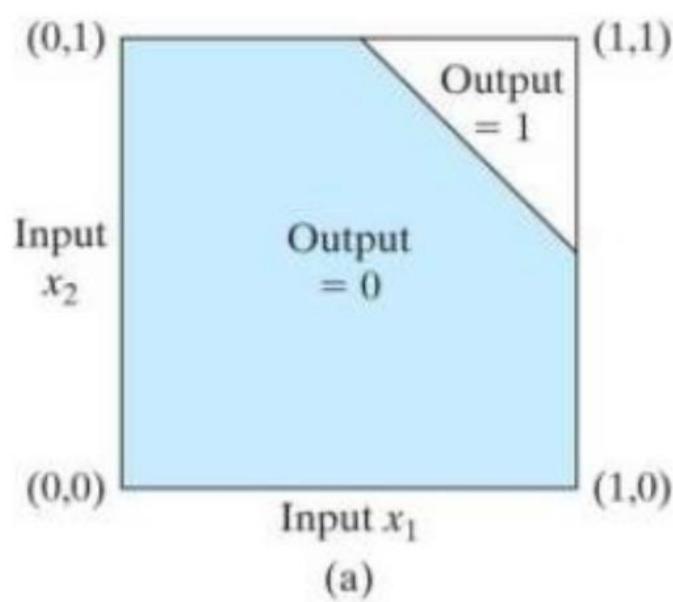
Idea is to add more layers



<https://towardsdatascience.com/multi-layer-neural-networks-with-sigmoid-function-deep-learning-for-rookies-2-bf464f9eb7f>

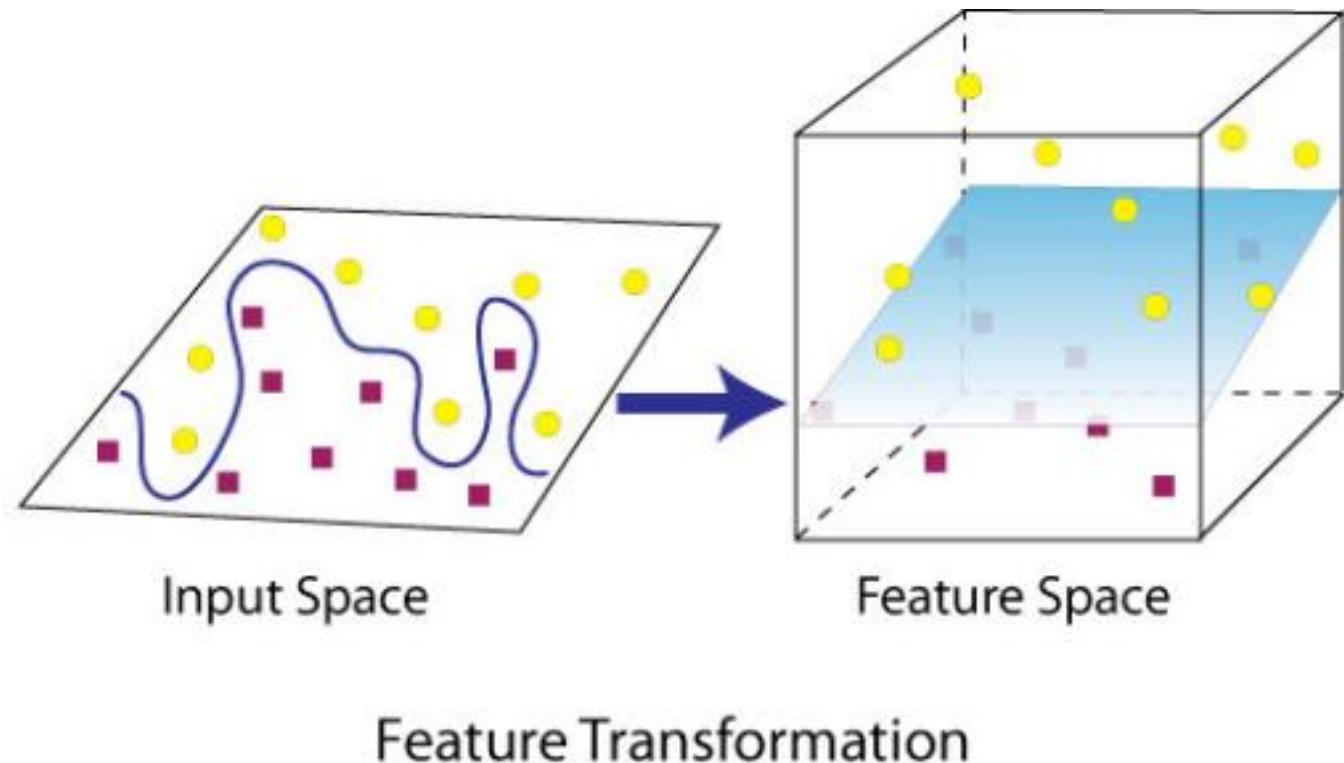
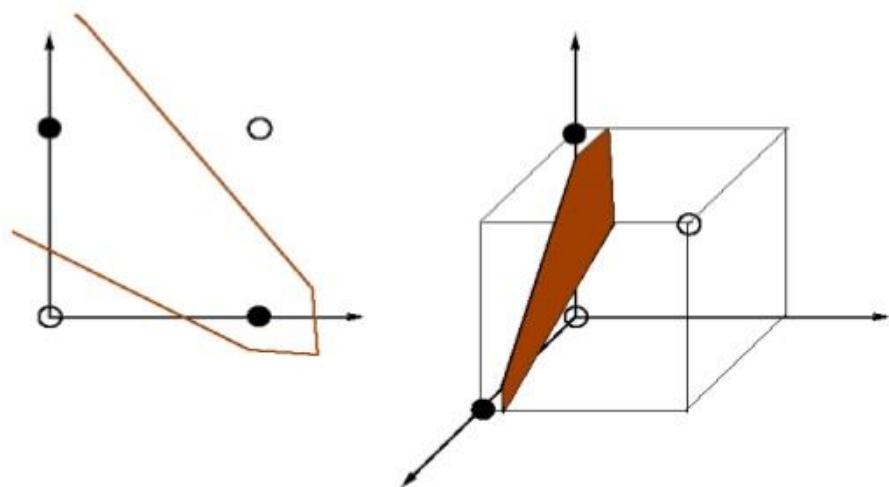
<https://www.ibm.com/developerworks/library/cc-cognitive-neural-networks-deep-dive/index.html>

# More Layers

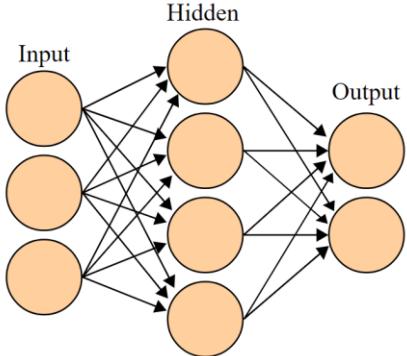


# Hyper Plane

Remember the XOR problem?

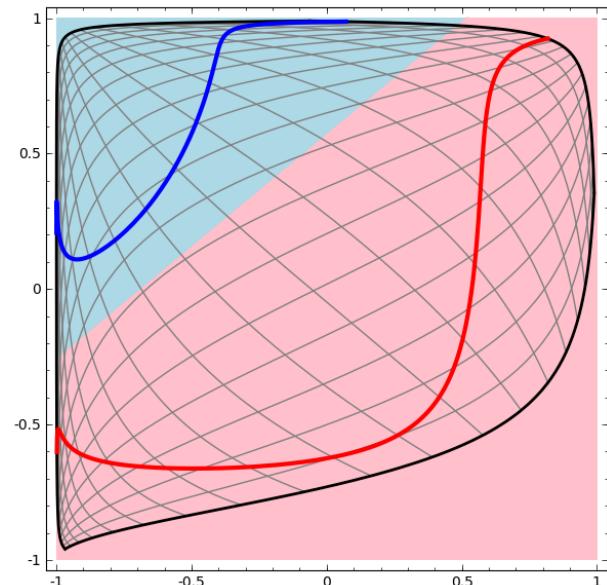
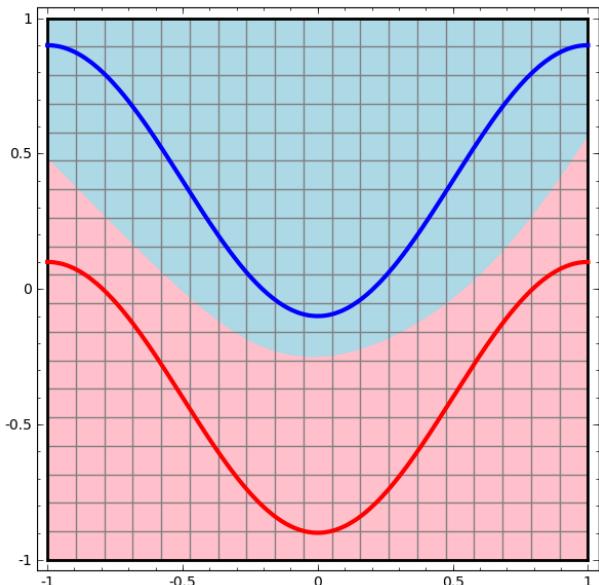
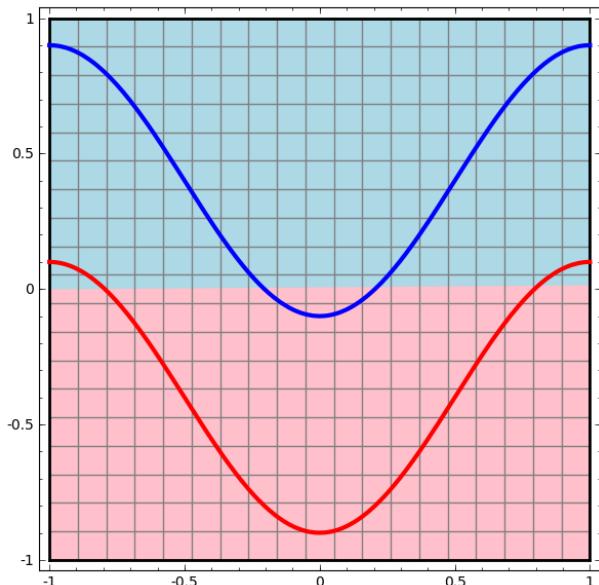
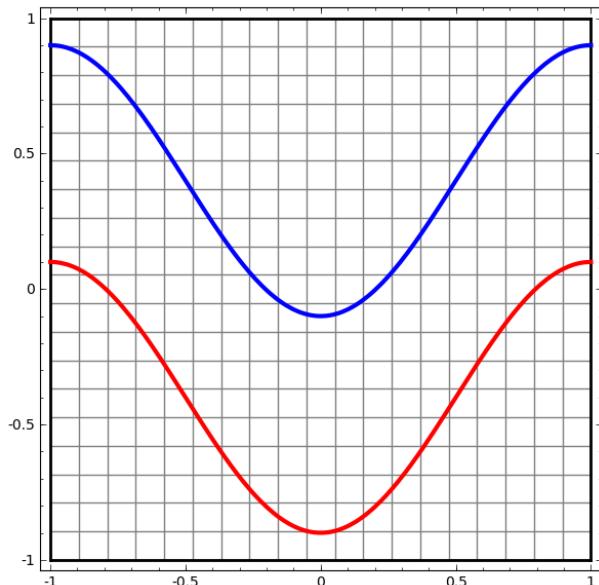


# Intuitive Understanding

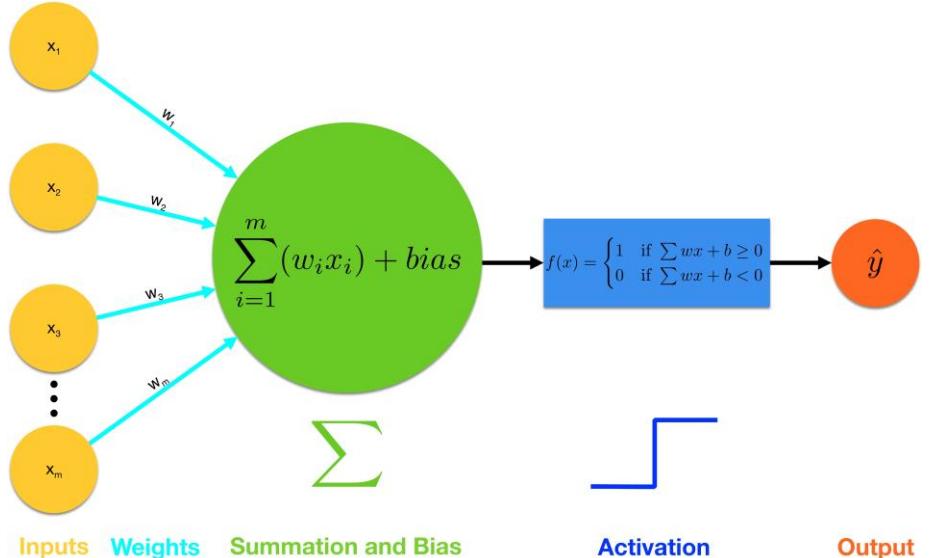


The hidden layer learns a representation so that the data is linearly separable

<http://colah.github.io/posts/2014-03-NN-Manifolds-Topology/>

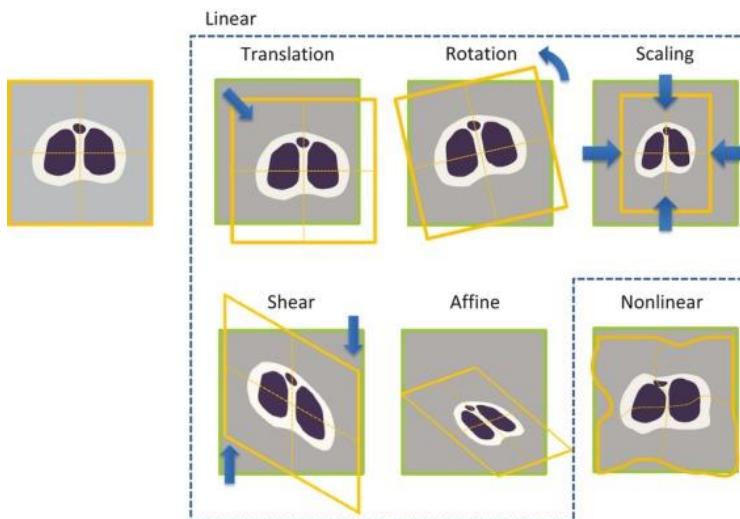


# From Spatial Transform Point



$$\hat{y} = \sigma(w^T x + b)$$

$w^T x + b$   
linear transformation  
(e.g. affine registration)



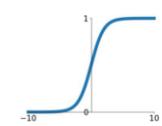
<http://nahuakang.com/page3/>

$$\sigma(\cdot)$$

Activation Functions

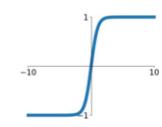
**Sigmoid**

$$\sigma(x) = \frac{1}{1+e^{-x}}$$



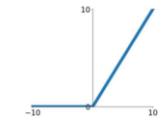
**tanh**

$$\tanh(x)$$



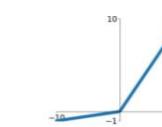
**ReLU**

$$\max(0, x)$$



**Leaky ReLU**

$$\max(0.1x, x)$$

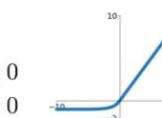


**Maxout**

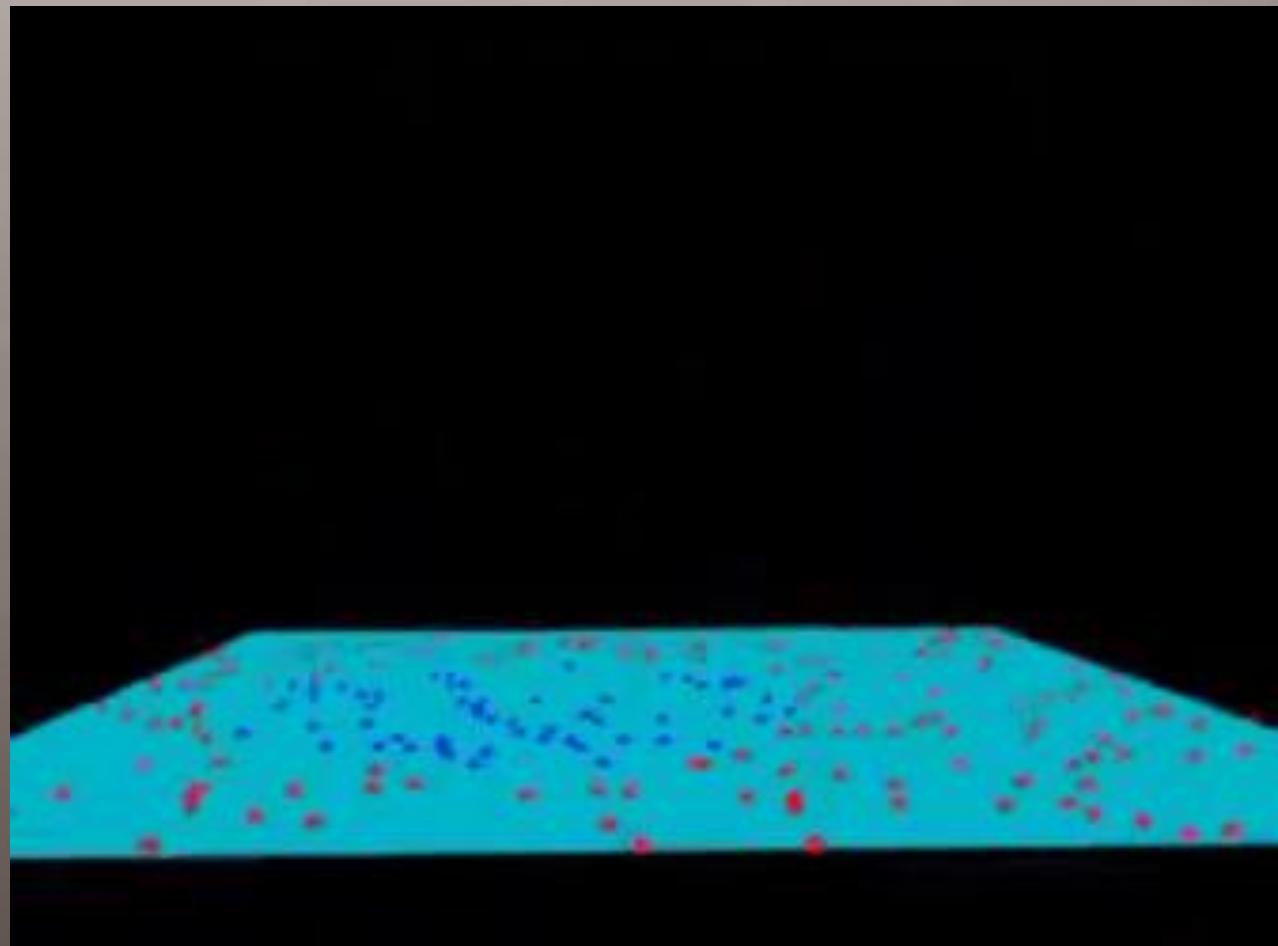
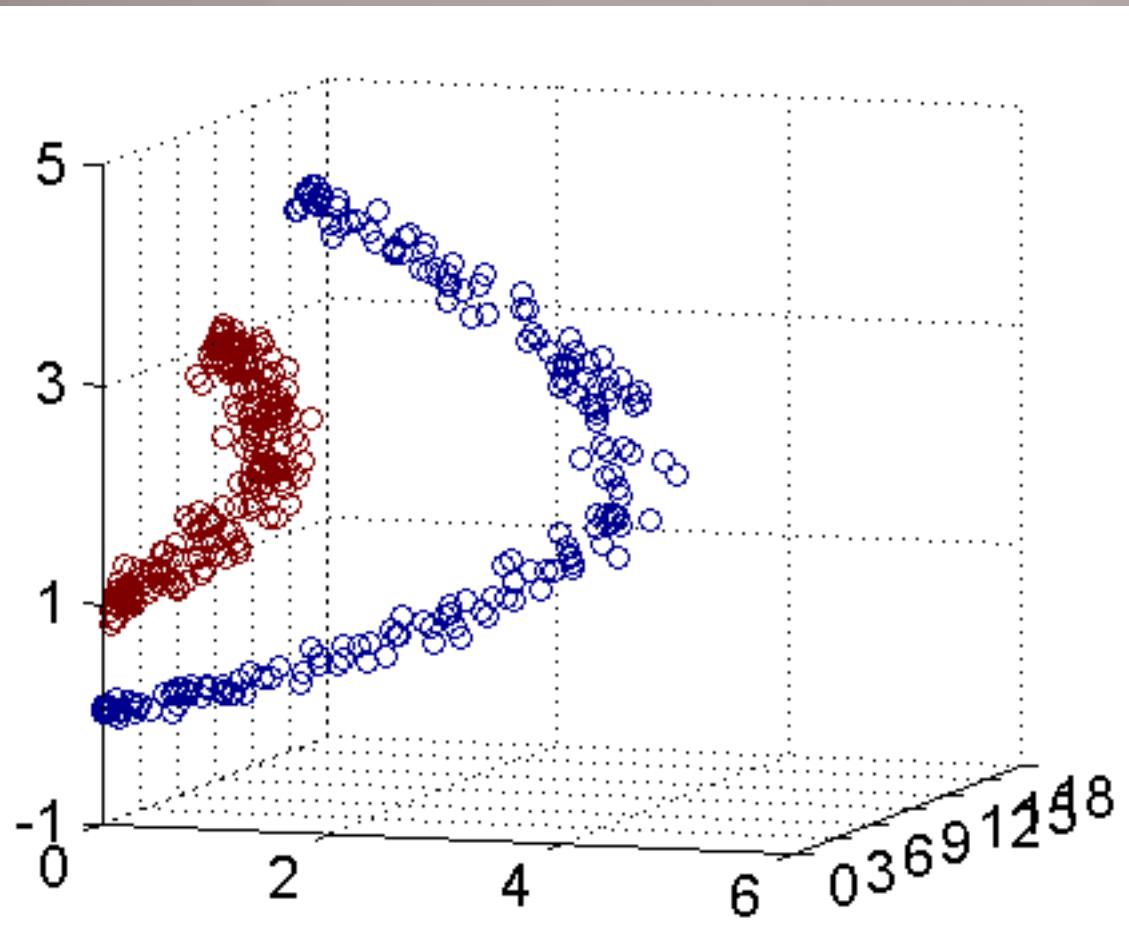
$$\max(w_1^T x + b_1, w_2^T x + b_2)$$

**ELU**

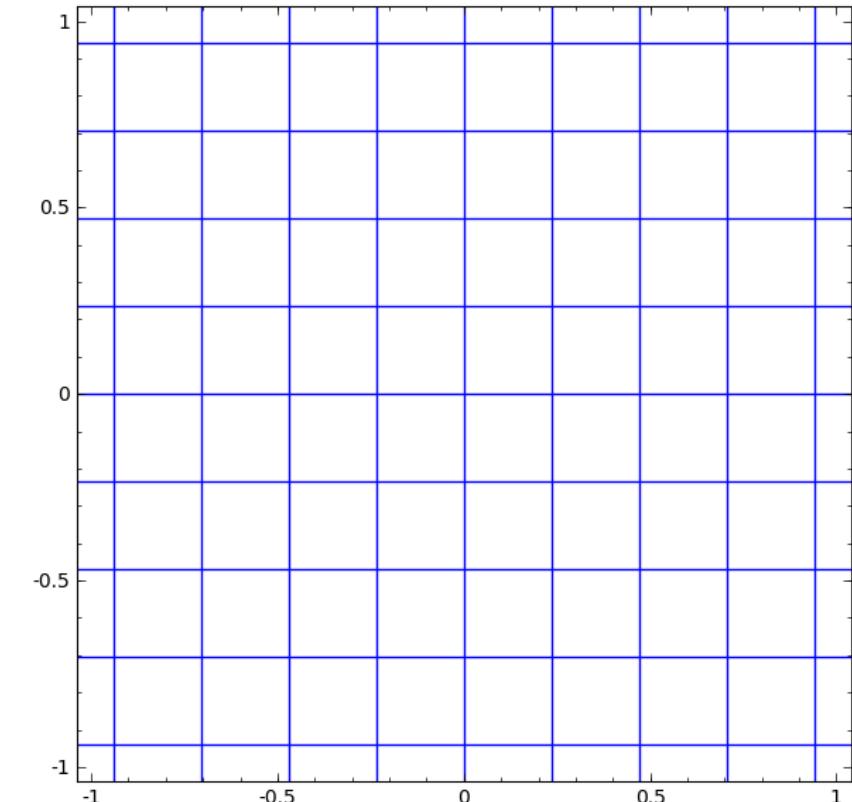
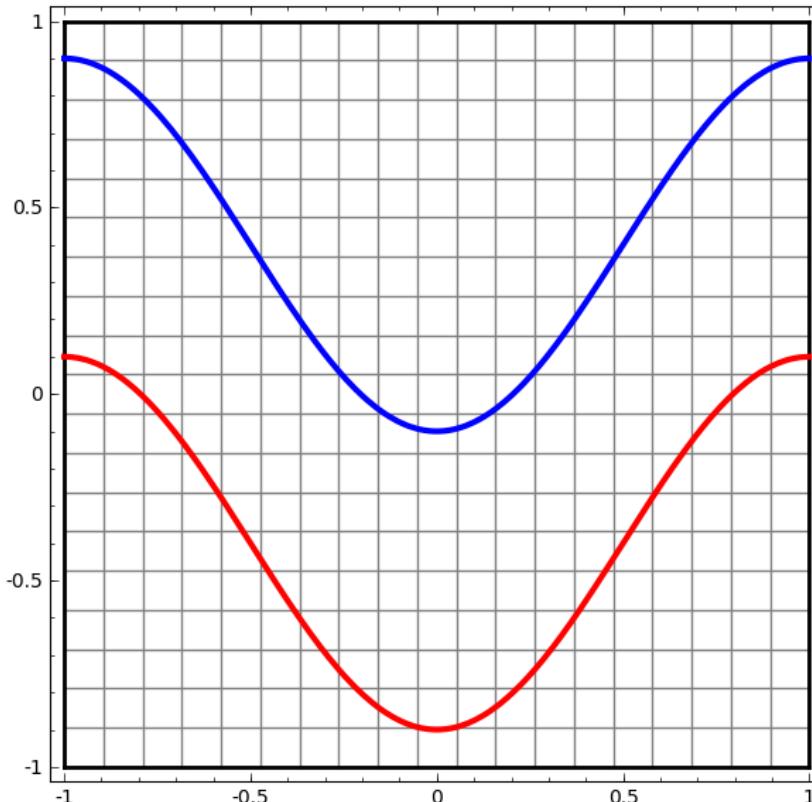
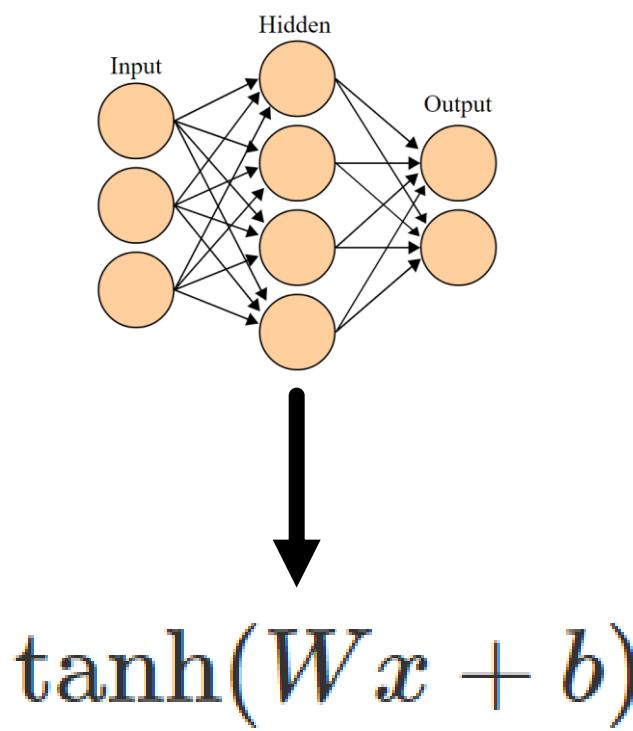
$$\begin{cases} x & x \geq 0 \\ \alpha(e^x - 1) & x < 0 \end{cases}$$



# Hyper Plane

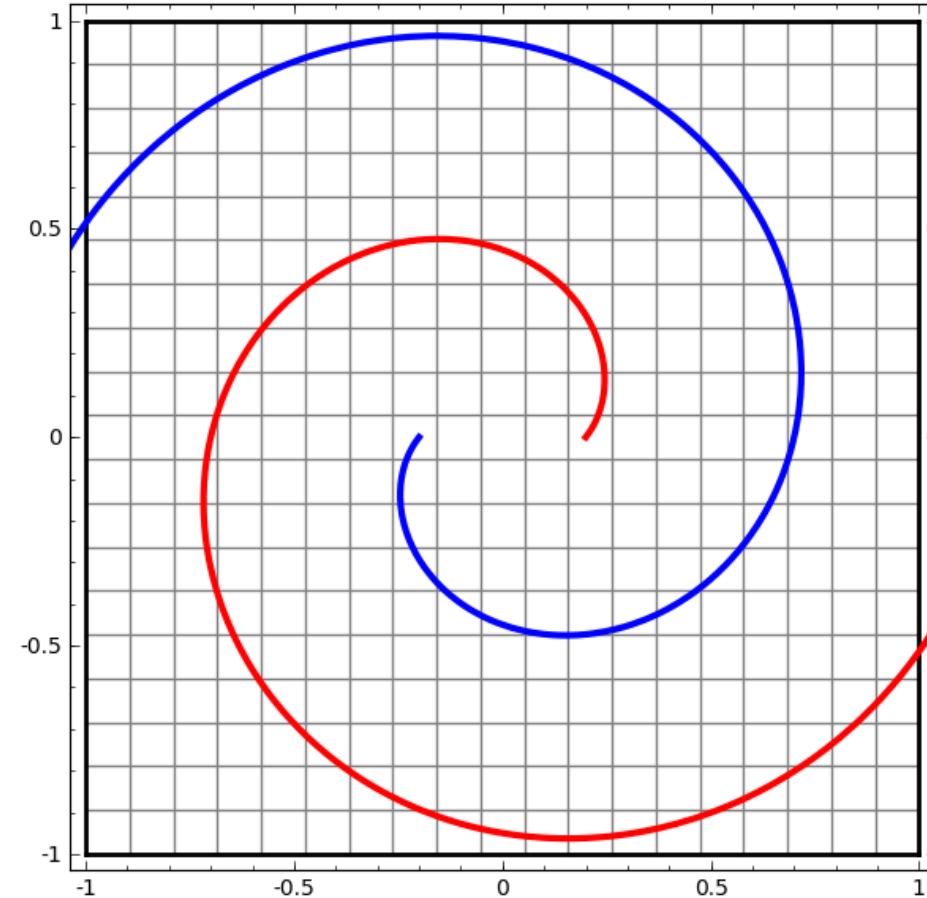
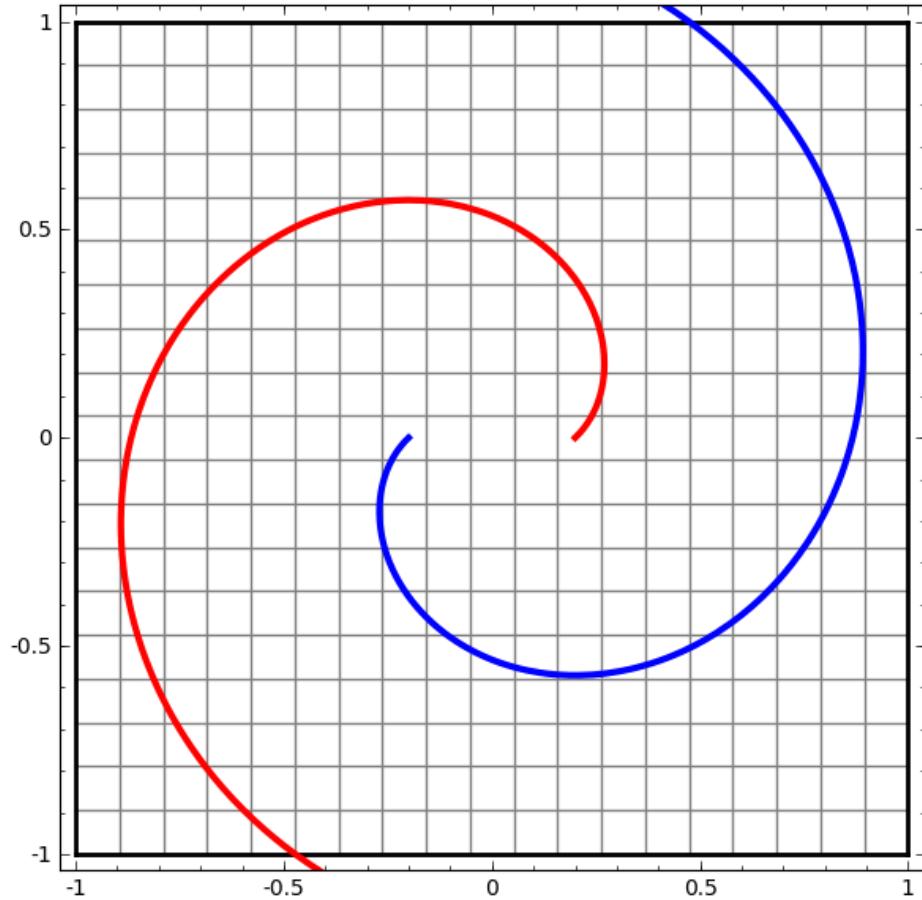


# (Deep) Neural Network = Hyper Plan



<http://colah.github.io/posts/2014-03-NN-Manifolds-Topology/>

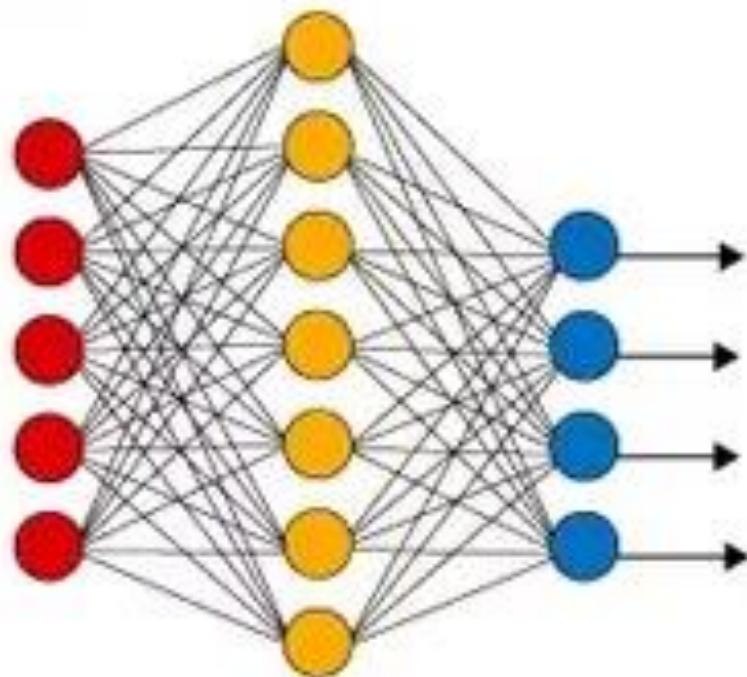
# Need Good Training



<http://colah.github.io/posts/2014-03-NN-Manifolds-Topology/>

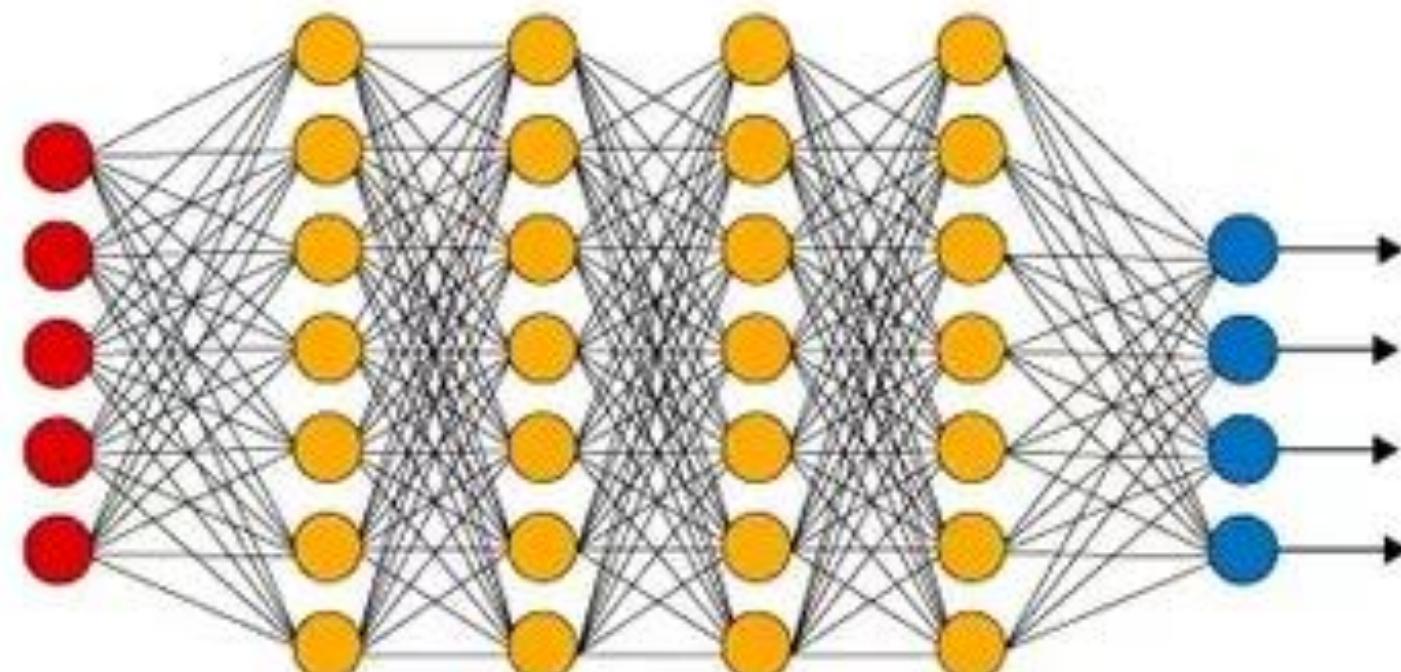
# Deep Neural Network

Simple Neural Network



● Input Layer

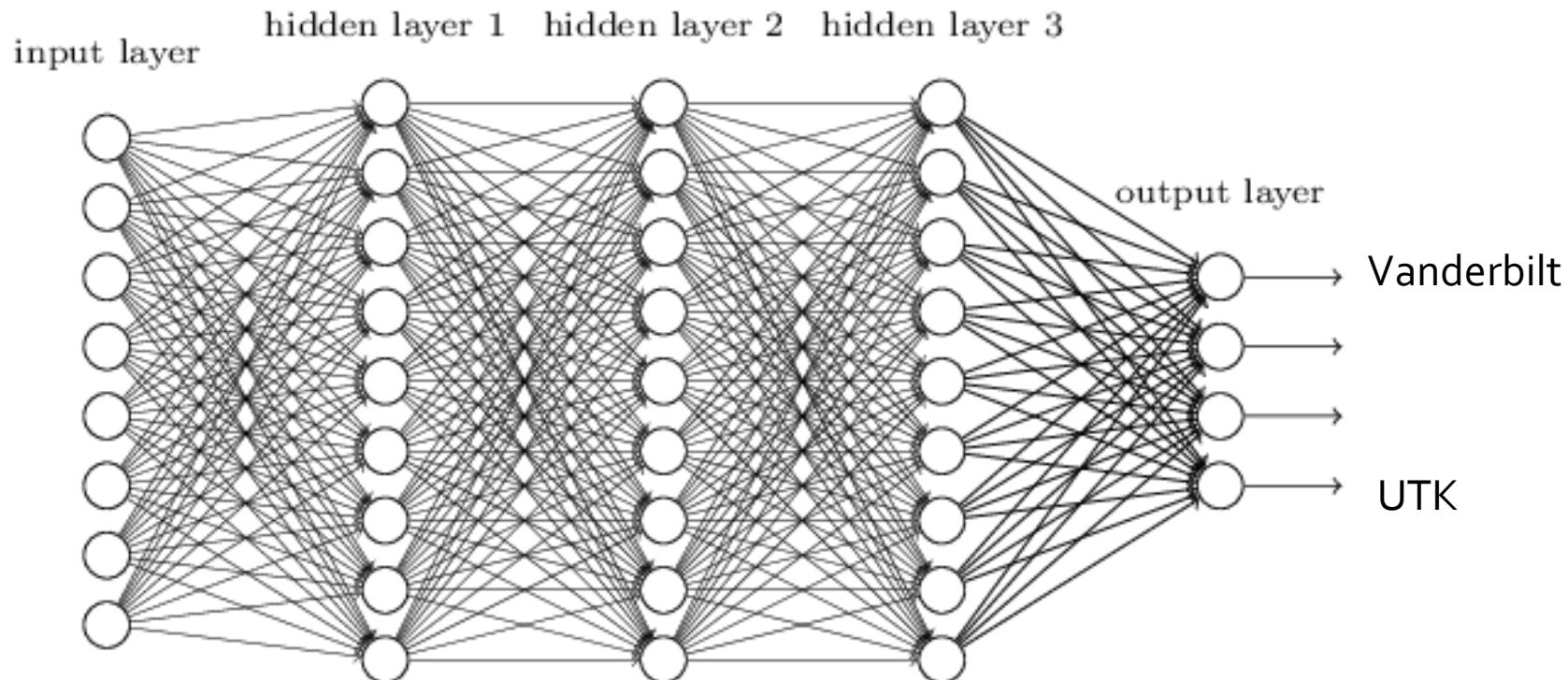
Deep Learning Neural Network



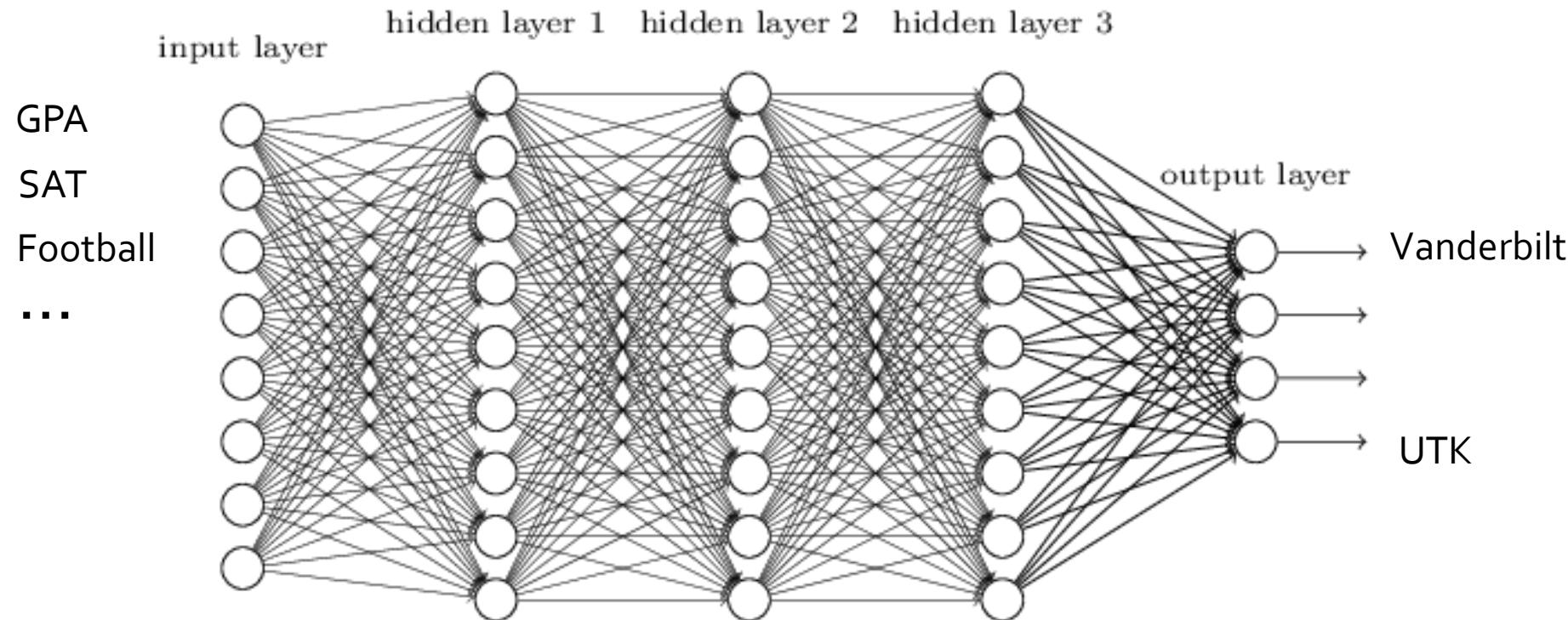
● Hidden Layer

● Output Layer

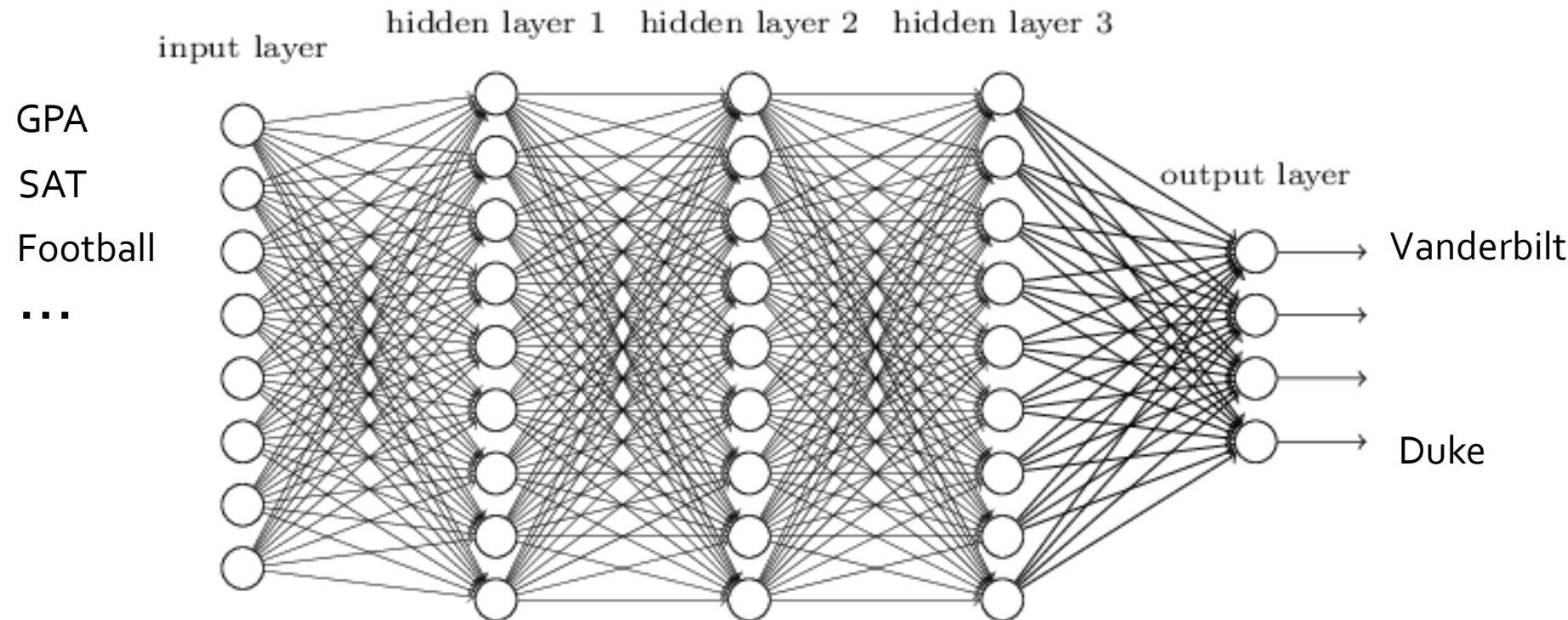
# 1D Data



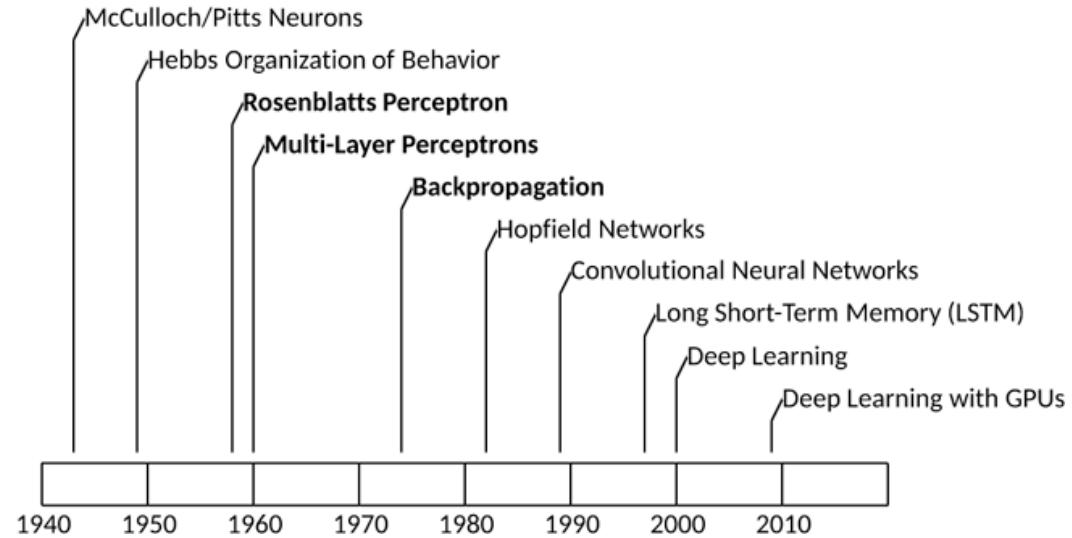
# 1D Data



# 1D Data

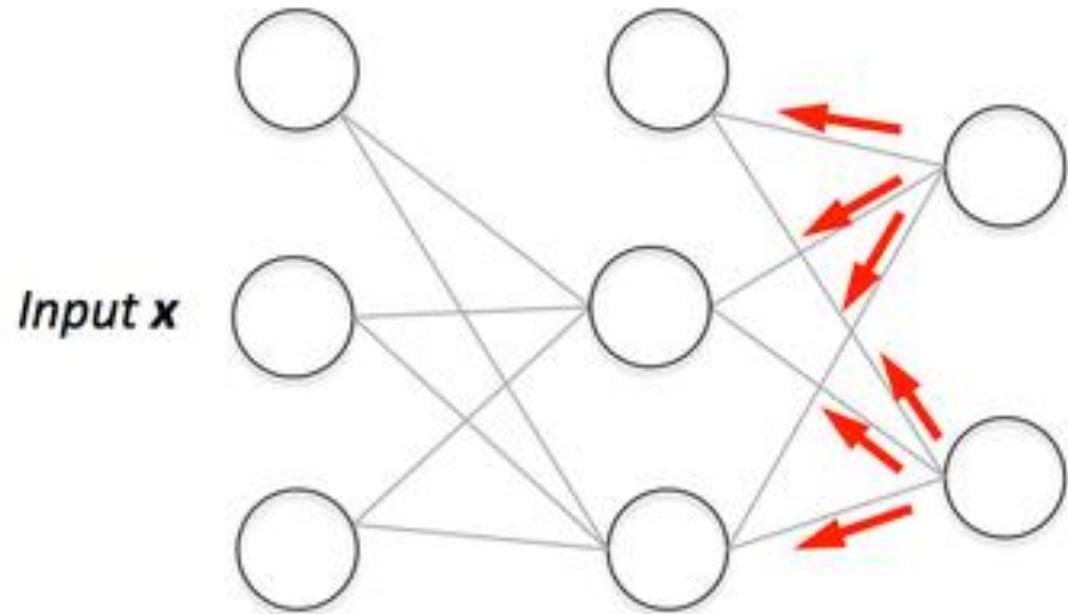


# Backpropagation



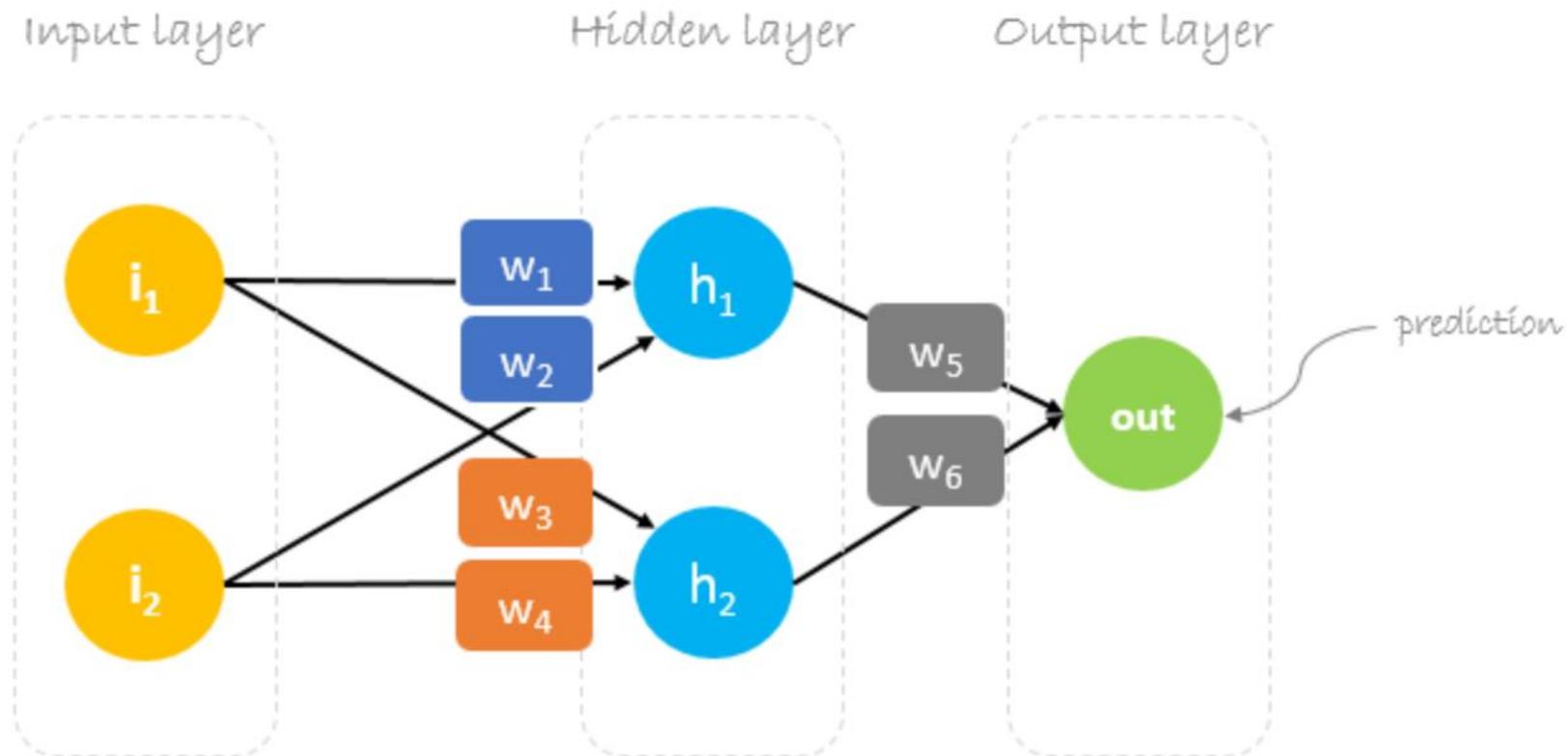
In 1975, Paul Werbos created the back-propagation algorithm, which could successfully train multilayer perceptrons and introduced various new applications of multilayer neural networks. This innovation led to a resurgence in neural network research and further popularized the method to solve real problems.

<https://www.ibm.com/developerworks/library/cc-cognitive-neural-networks-deep-dive/index.html>



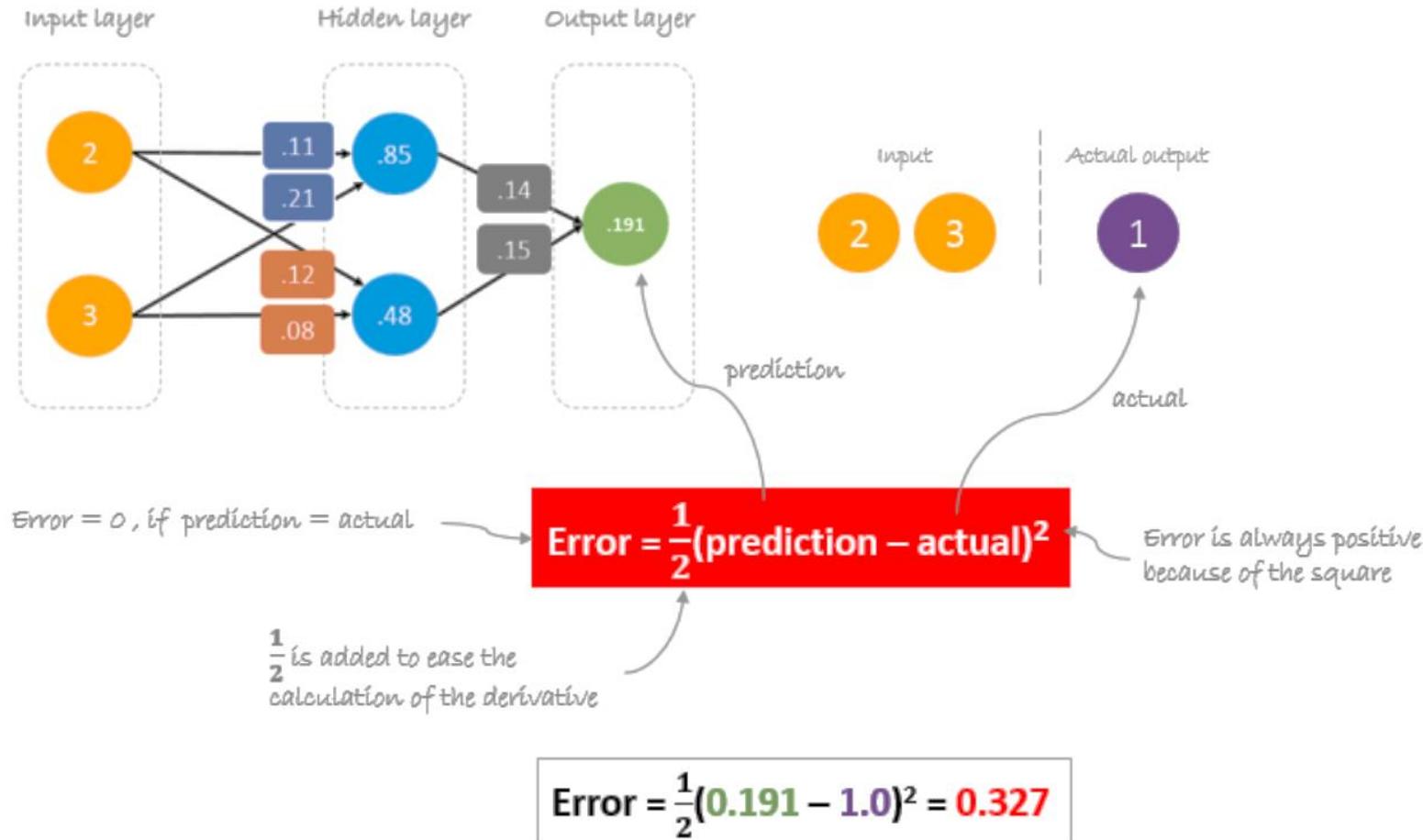
<http://hmkcode.github.io/ai/backpropagation-step-by-step/>

# Deduction



<http://hmkcode.github.io/ai/backpropagation-step-by-step/>

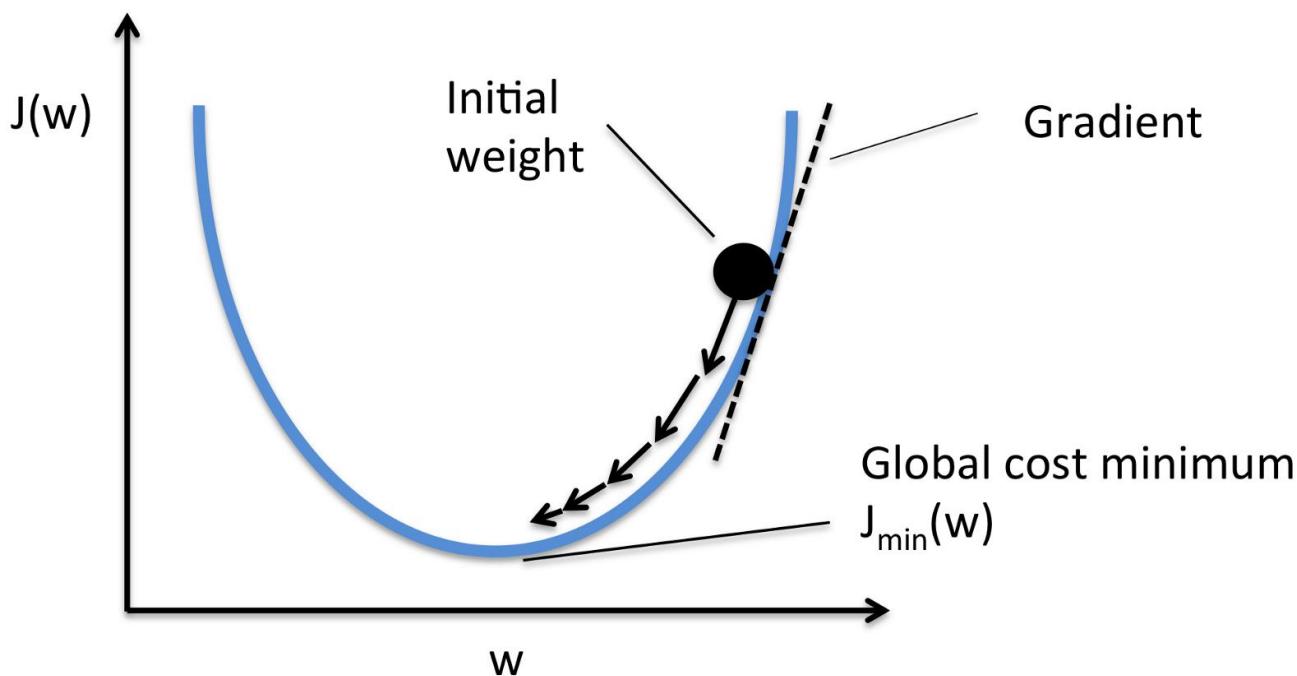
# Forward Path



The question now is  
How to change\update  
the weights value  
so that the error is  
reduced?  
The answer is  
Backpropagation!

# Back Propagation & Gradient Descent

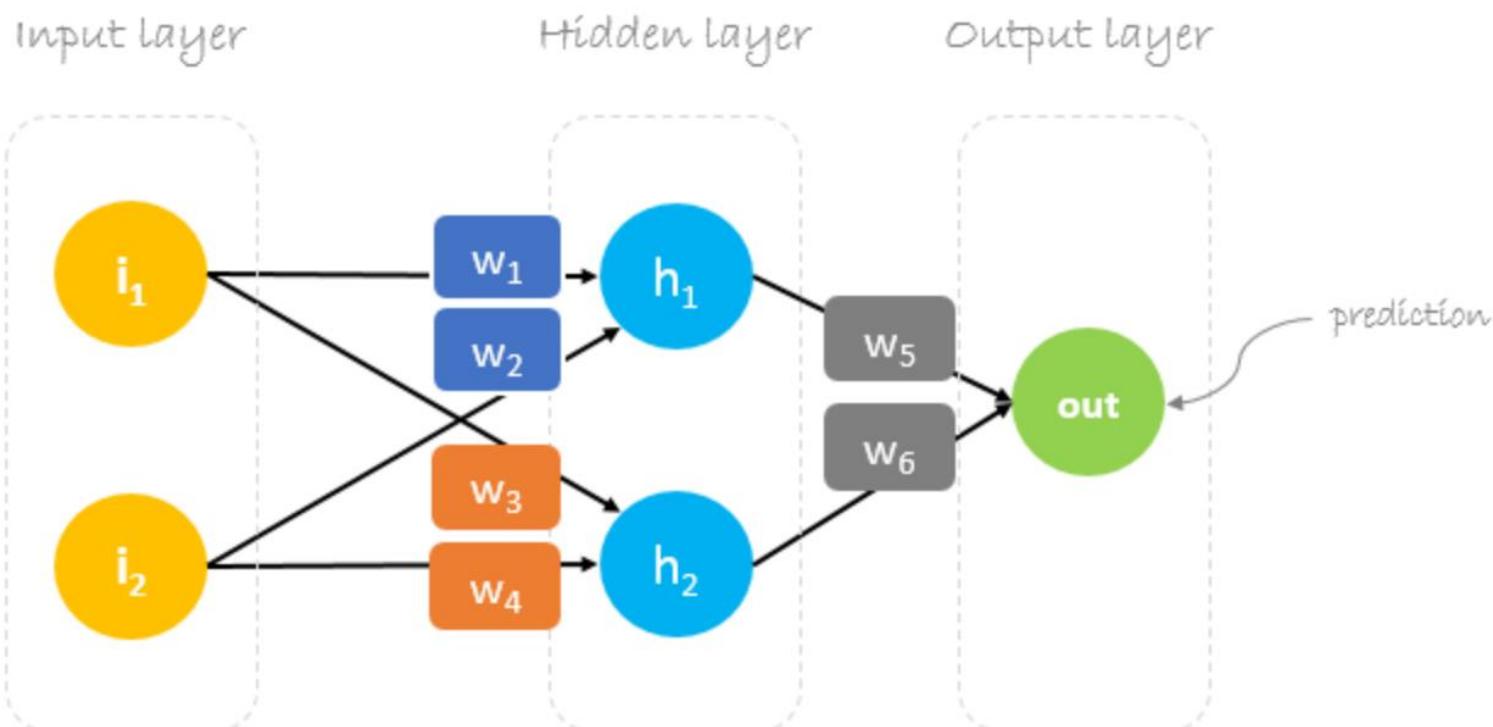
Backpropagation, short for “backward propagation of errors”, is a mechanism used to update the **weights** using gradient descent. It calculates the gradient of the error function with respect to the neural network’s weights. The calculation proceeds backwards through the network



$$*W_x = W_x - a \left( \frac{\partial \text{Error}}{\partial W_x} \right)$$

Derivative of Error  
with respect to weight  
↓  
Old weight  
↓  
New weight  
↑  
Learning rate

# Get One W



Derivative of Error  
with respect to weight

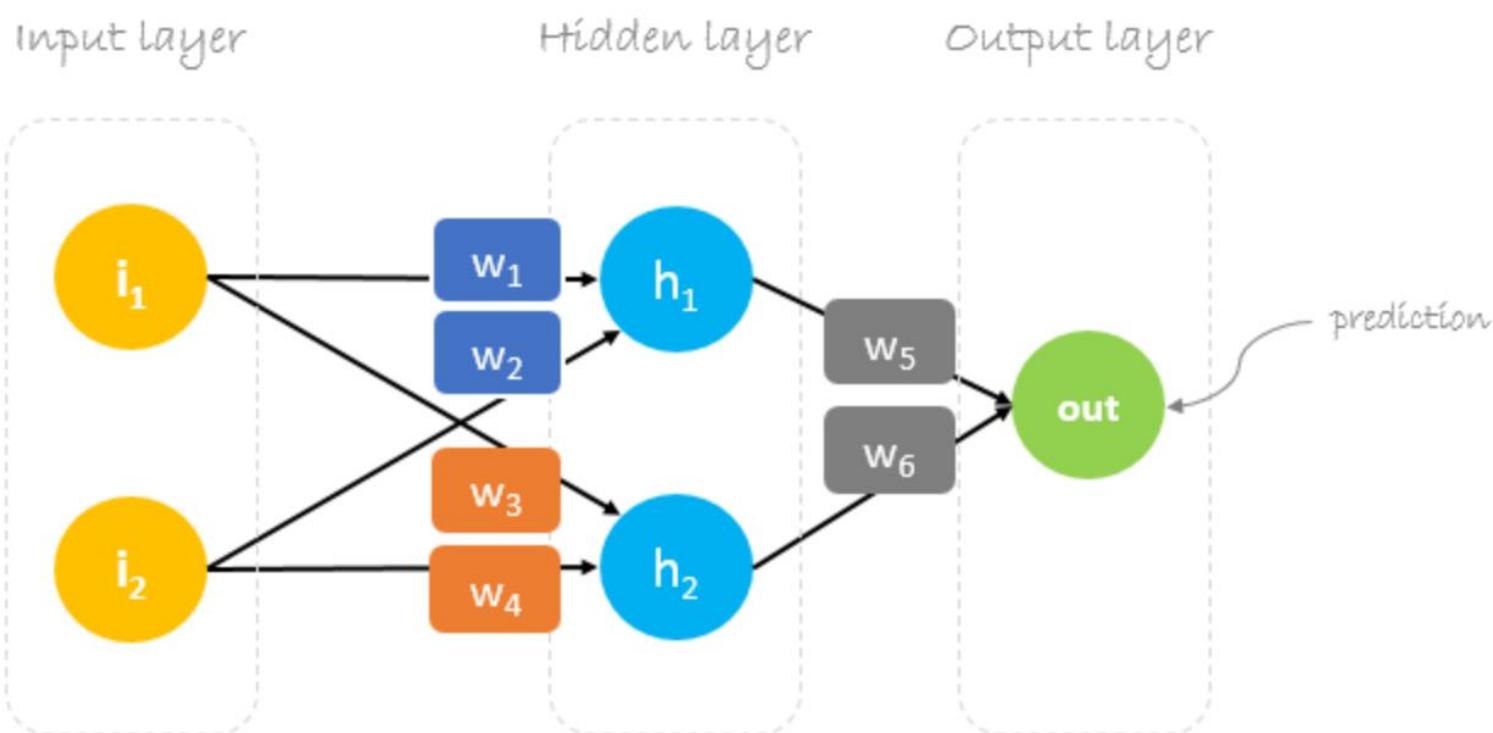
old weight

New weight

Learning rate

$$*W_x = W_x - \text{a} \left( \frac{\partial \text{Error}}{\partial W_x} \right)$$

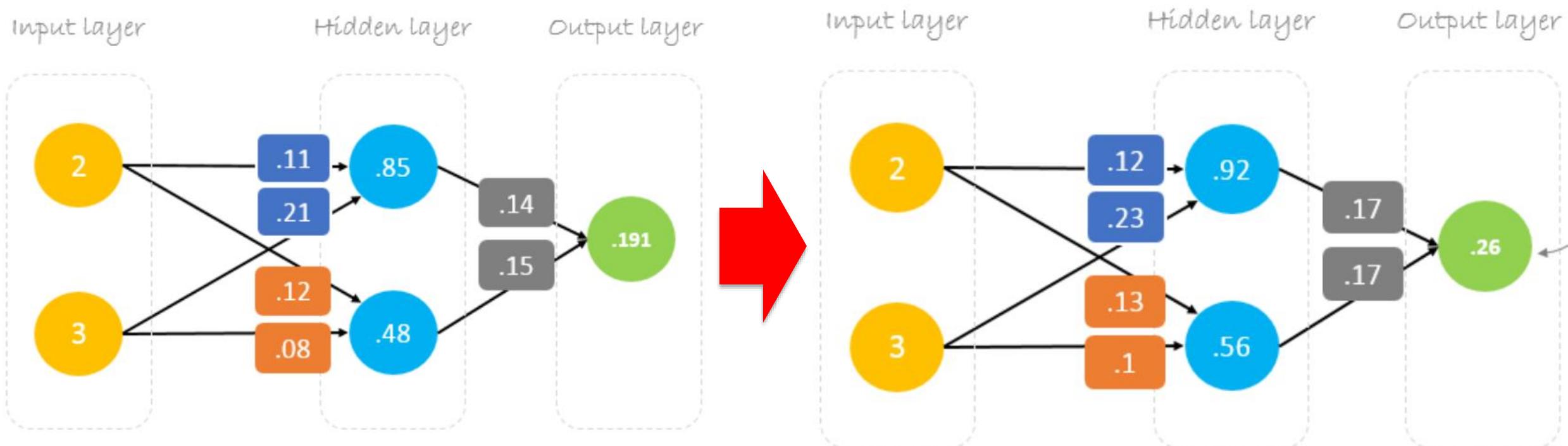
# Results



$\Delta = \text{prediction} - \text{actual}$

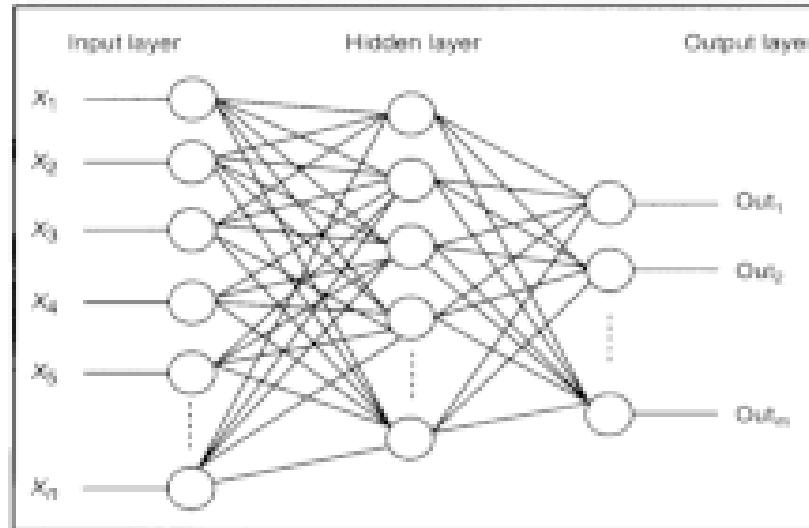
- \*  $w_6 = w_6 - a (h_2 \cdot \Delta)$
- \*  $w_5 = w_5 - a (h_1 \cdot \Delta)$
- \*  $w_4 = w_4 - a (i_2 \cdot \Delta w_6)$
- \*  $w_3 = w_3 - a (i_1 \cdot \Delta w_6)$
- \*  $w_2 = w_2 - a (i_2 \cdot \Delta w_5)$
- \*  $w_1 = w_1 - a (i_1 \cdot \Delta w_5)$

# Backward Path

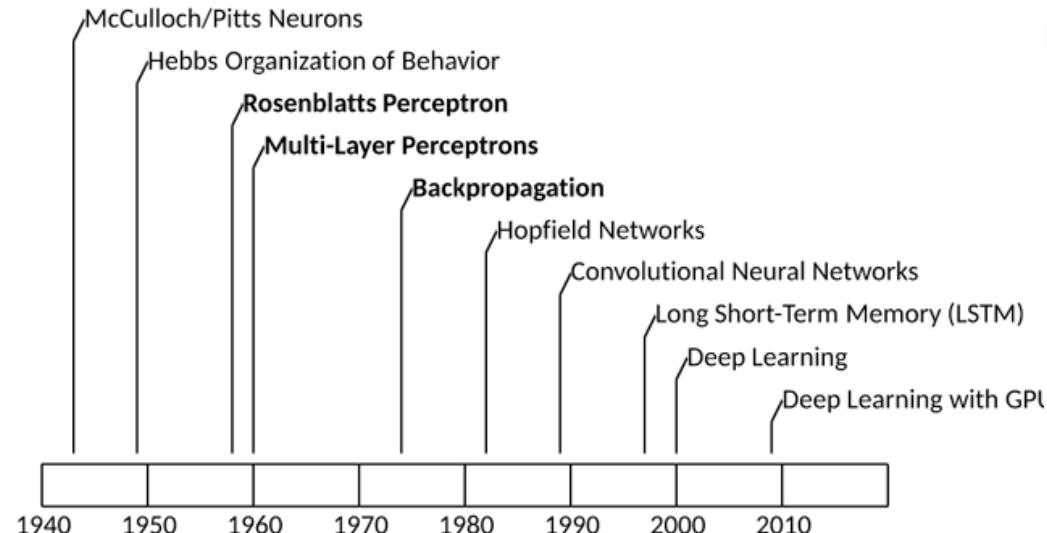


# Summarize on How To Train Neural Network

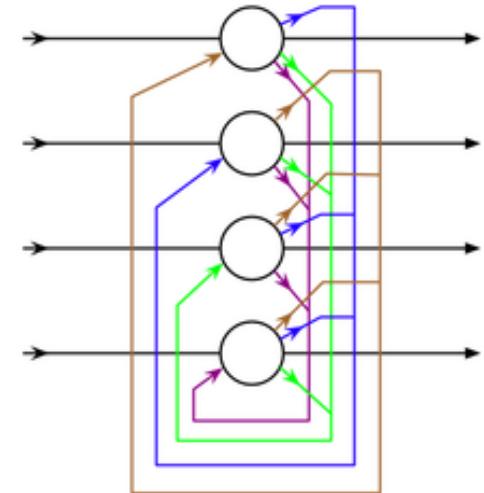
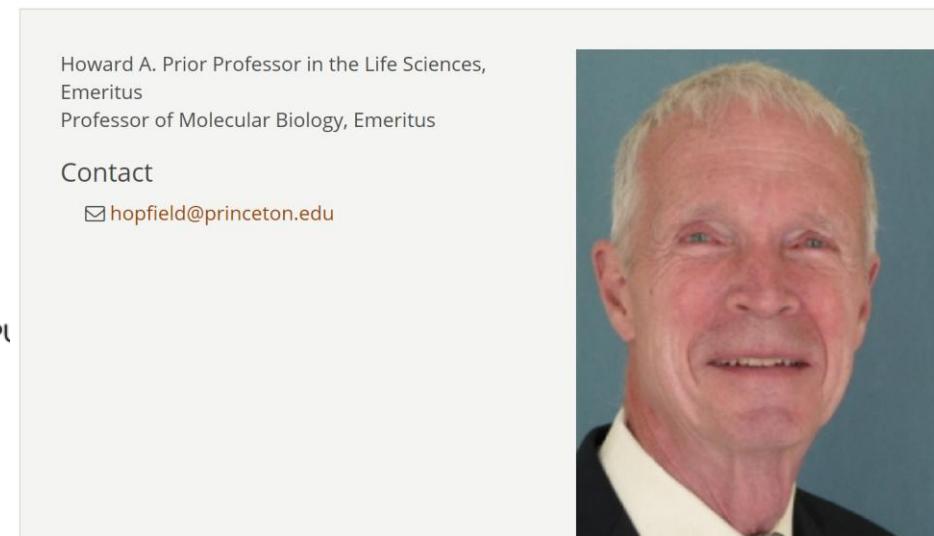
1



# 1D+1D = “2D” Temporal Modal



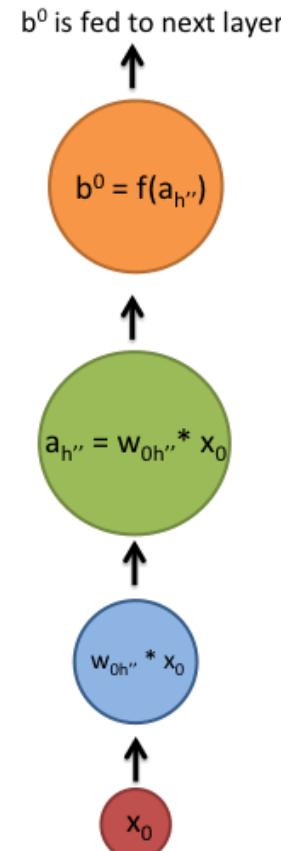
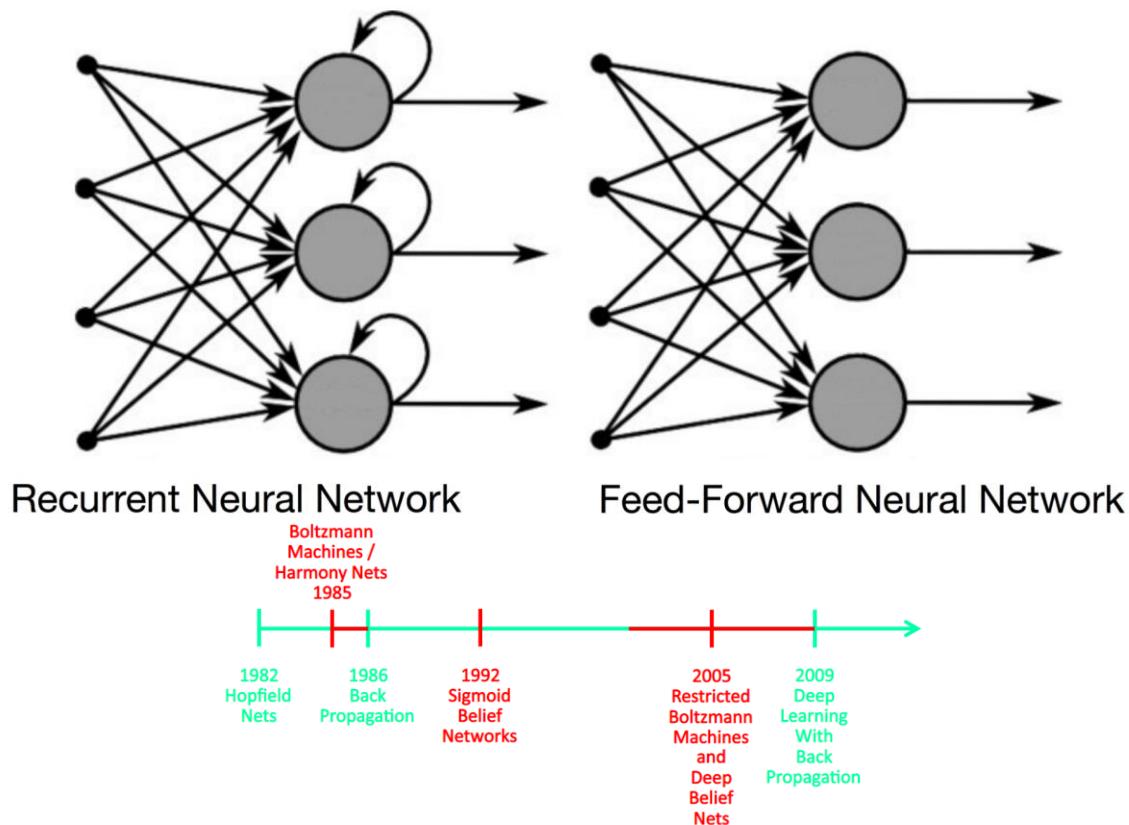
John J. Hopfield



Hopfield, John J., David I. Feinstein, and Richard G. Palmer. "Unlearning" has a stabilizing effect in collective memories. *Nature* (1983).

Recurrent Neural Network (RNN)

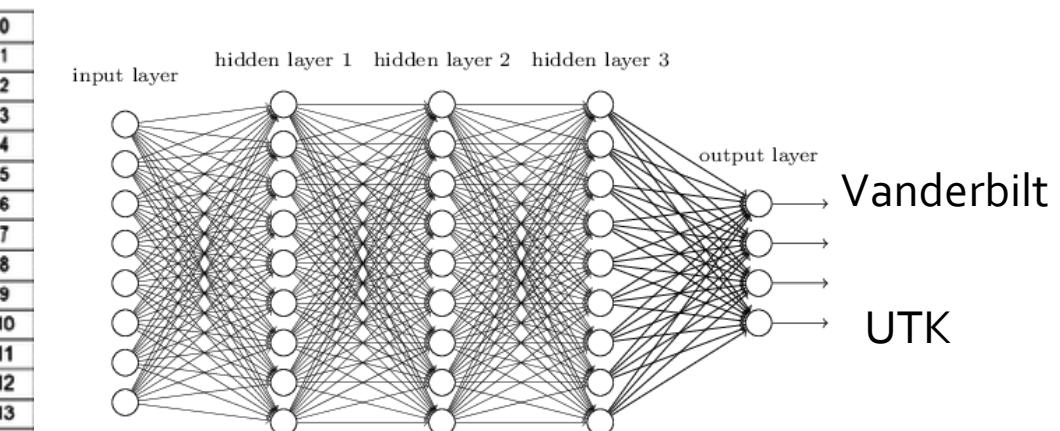
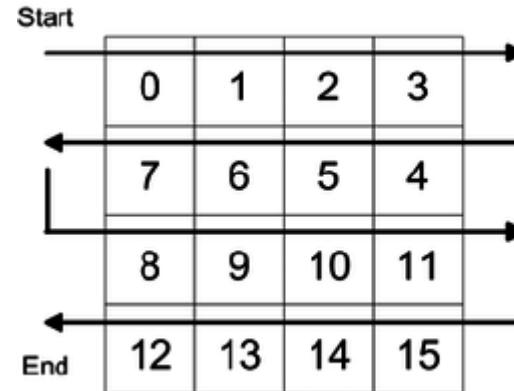
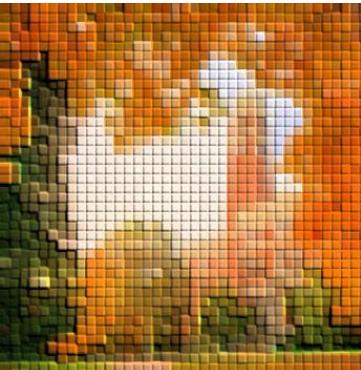
# Feed-forward and Recurrent Network



<https://towardsdatascience.com/recurrent-neural-networks-and-lstm-4b601dd822a5>

<https://www.leiphone.com/news/201608/syAwLNx4bGPyFYI1.html> <https://slideplayer.com/slide/3383596/>

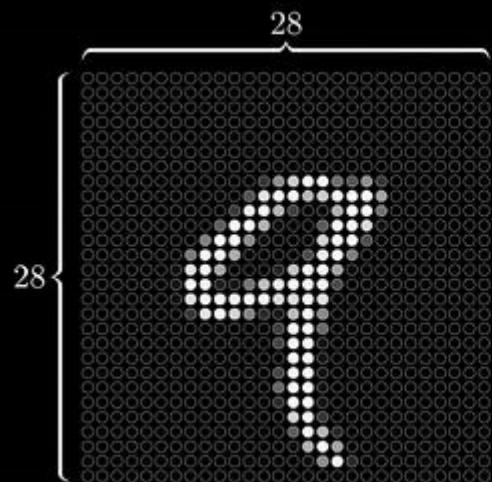
# 2D Data (Image) ?



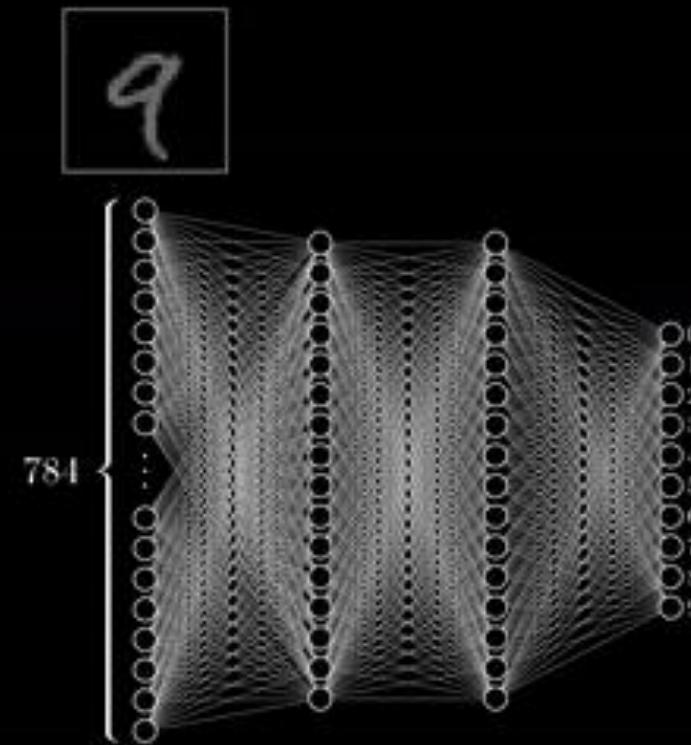
<https://www.usnews.com/best-colleges/vanderbilt-3535>

[https://www.researchgate.net/figure/Scanning-2D-blocks-left-into-1D-vector-as-input-into-logical-transform-middle-the\\_fig4\\_228345705](https://www.researchgate.net/figure/Scanning-2D-blocks-left-into-1D-vector-as-input-into-logical-transform-middle-the_fig4_228345705)

# Dense Neural Network

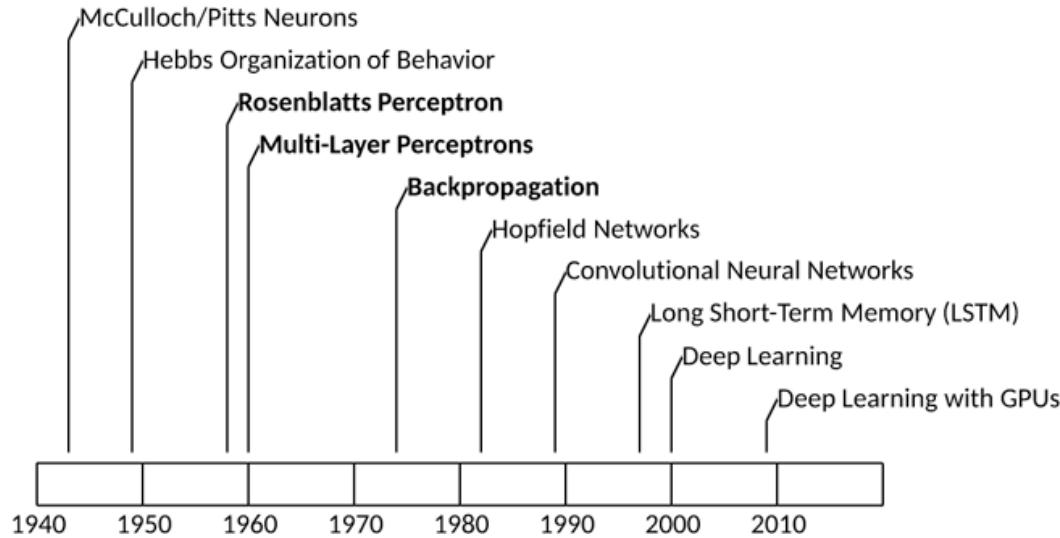


<https://gfycat.com/gifs/search/deep+neural+networks>



<http://homdor.com/search/deep-neural-networks>

# Convolutional Neural Network

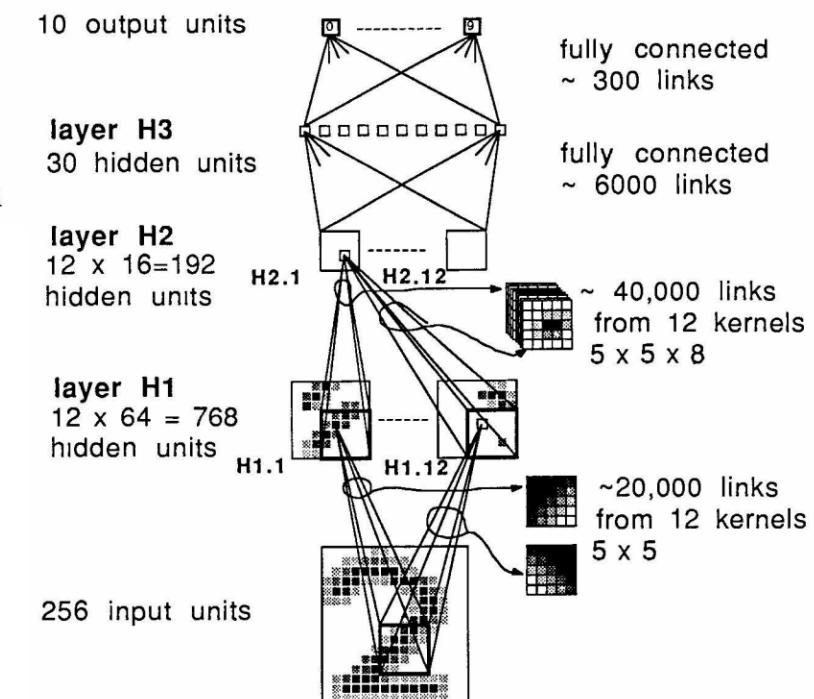


Modern convNets were proposed in an article by Yann LeCun et al of 1989. He used backprop to train a Neocognitron-like architecture and refined the architecture trying different variations of the model.

<https://www.ibm.com/developerworks/library/cc-cognitive-neural-networks-deep-dive/index.html>

## Backpropagation Applied to Handwritten Zip Code Recognition

Y. LeCun  
B. Boser  
J. S. Denker  
D. Henderson  
R. E. Howard  
W. Hubbard  
L. D. Jackel  
AT&T Bell Laboratories Holmdel, NJ 07733 USA



<http://espiritureresafidalgo.blogspot.com/2017/03/teresa-fidalgo-girl-of-road.html>

# Convolutional Network Demo from 1993



<https://gfycat.com/gifs/detail/AchingCluelessEchidna>

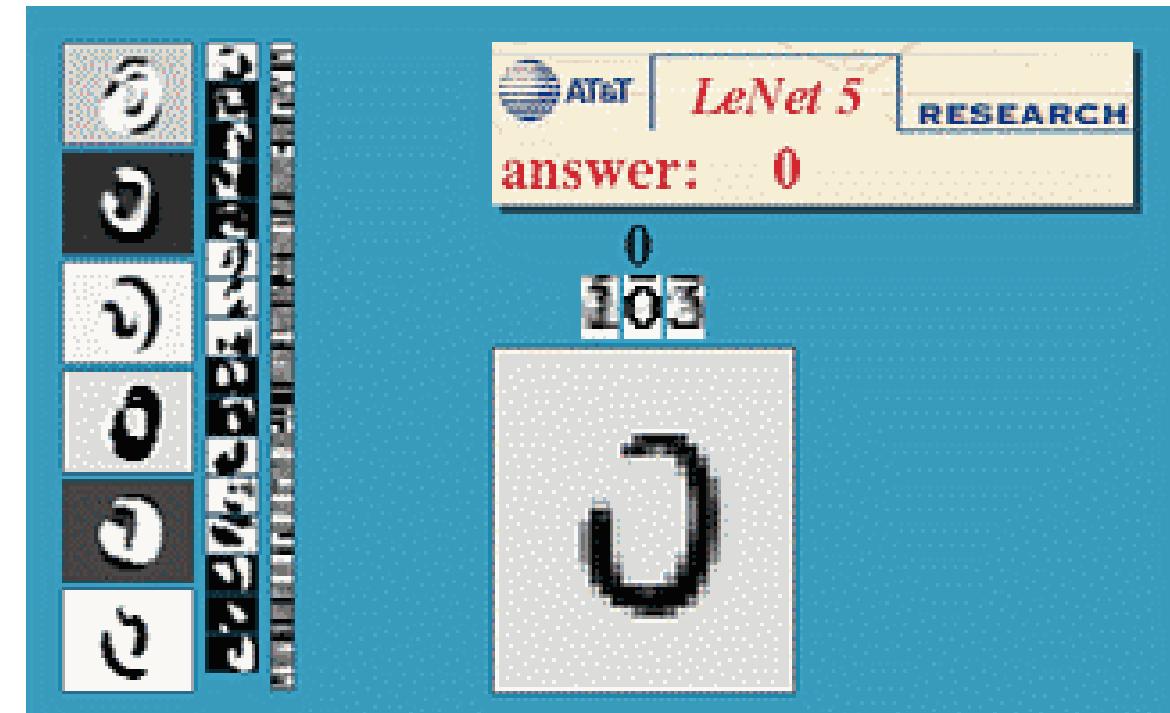
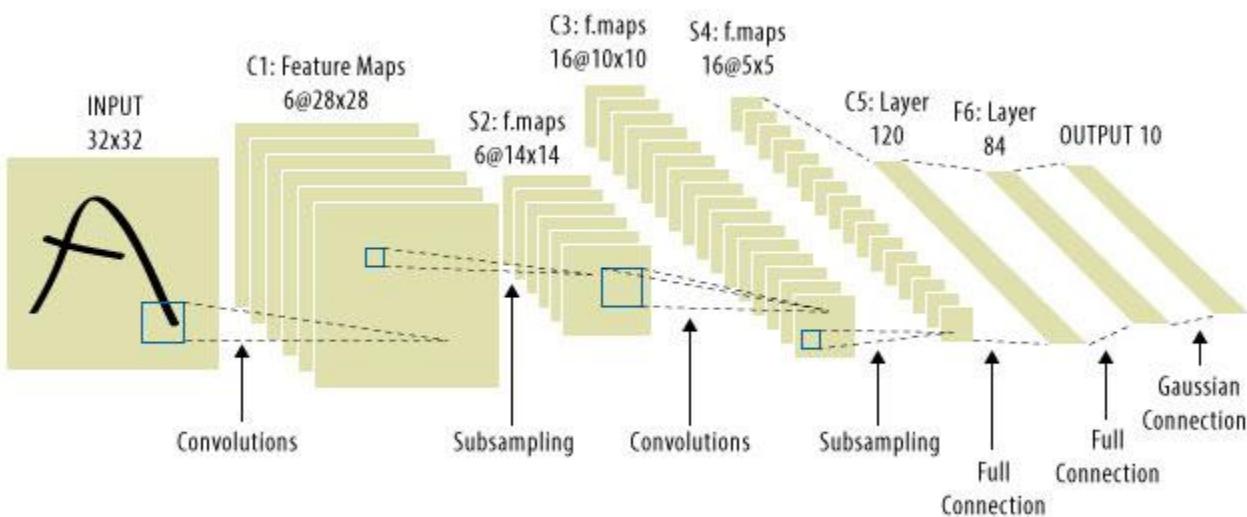
# Convolutional Neural Network

PROC. OF THE IEEE, NOVEMBER 1998

1

## Gradient-Based Learning Applied to Document Recognition

Yann LeCun, Léon Bottou, Yoshua Bengio, and Patrick Haffner



<http://yann.lecun.com/exdb/lenet/>

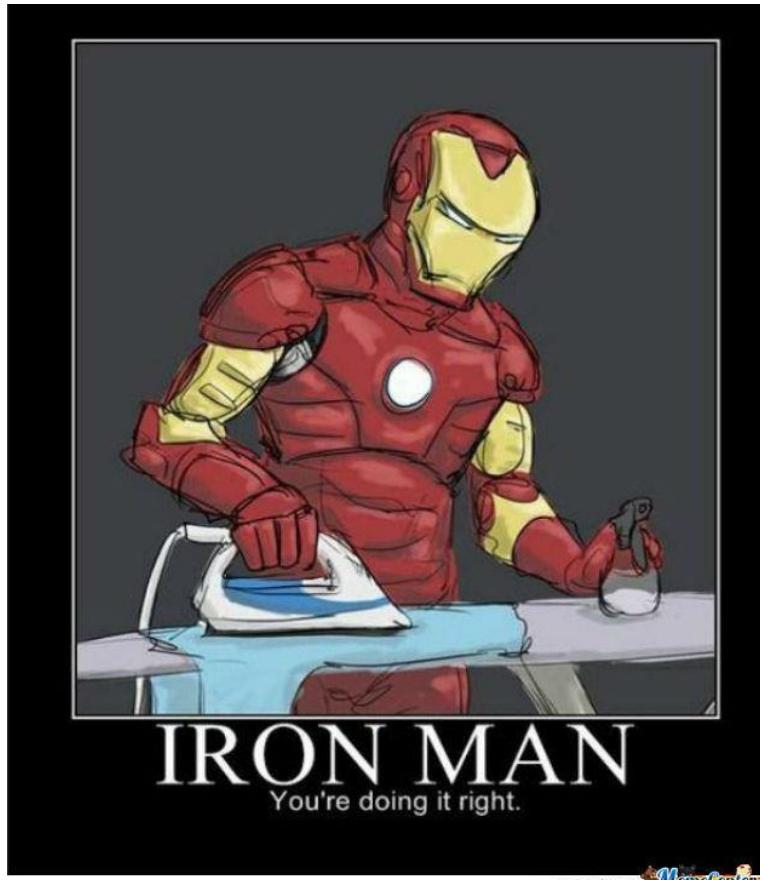
# Computer Vision



<https://www.empireonline.com/shopping/movies/top-iron-man-merch/>



<https://www.playstation.com/en-us/games/marvels-iron-man-vr-ps4/>



<https://www.memecenter.com/fun/1136662/stupid-iron-man>

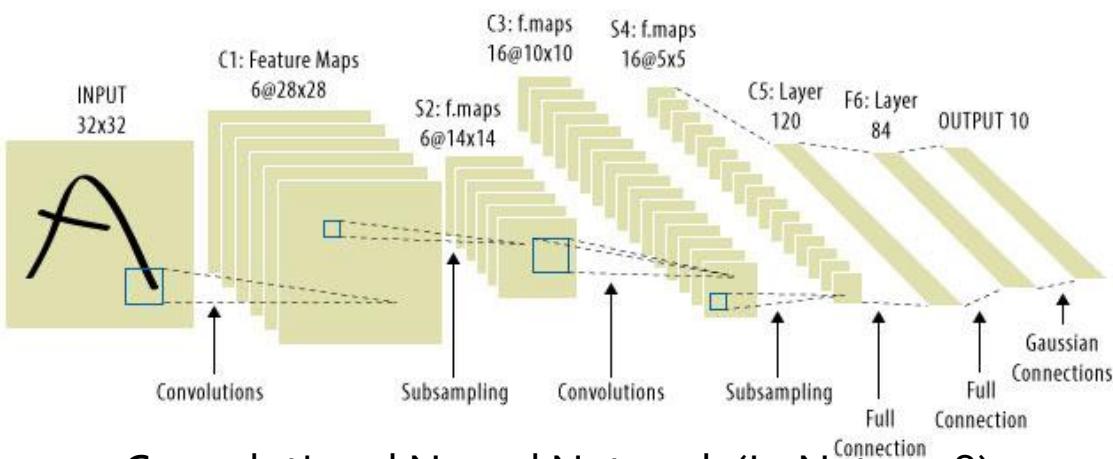
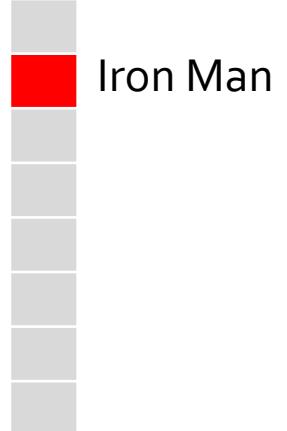


<https://www.stupid.com/iron-man-comics-bust-bank.html>

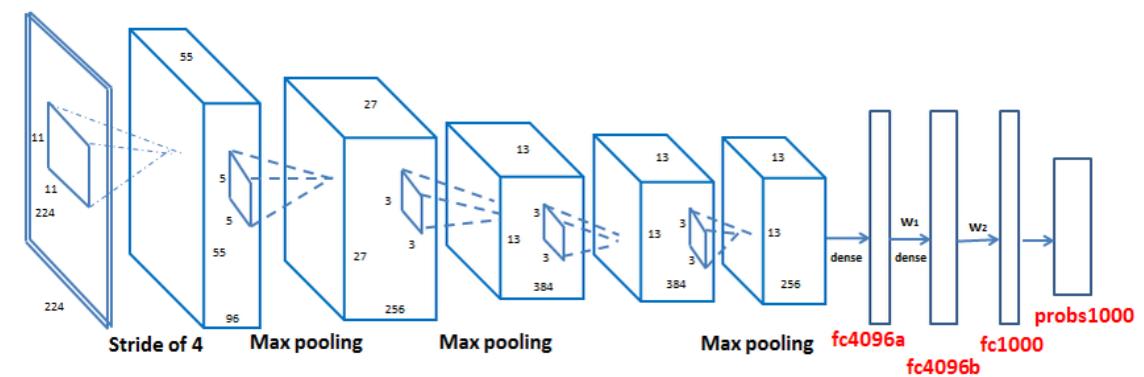
# What can we do?



?



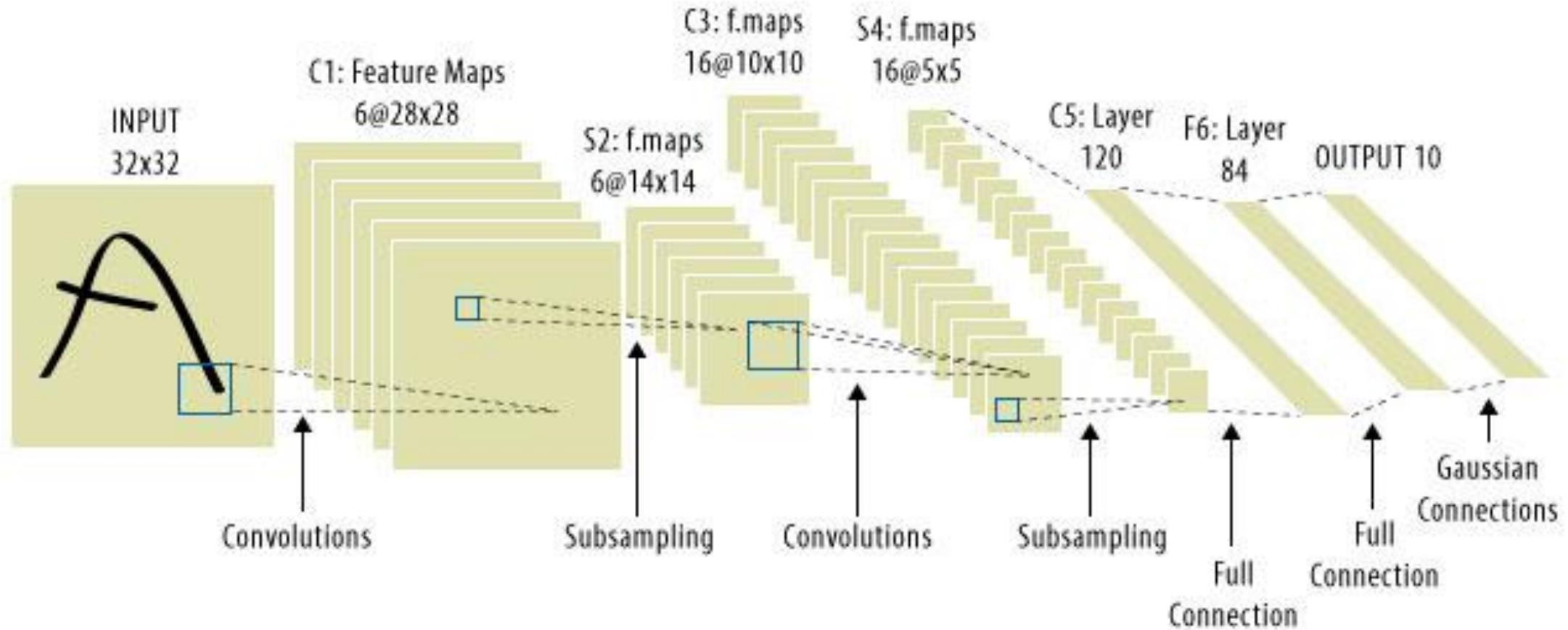
Convolutional Neural Network (LeNet 1998)



Deep Convolutional Neural Network (AlexNet 2012)

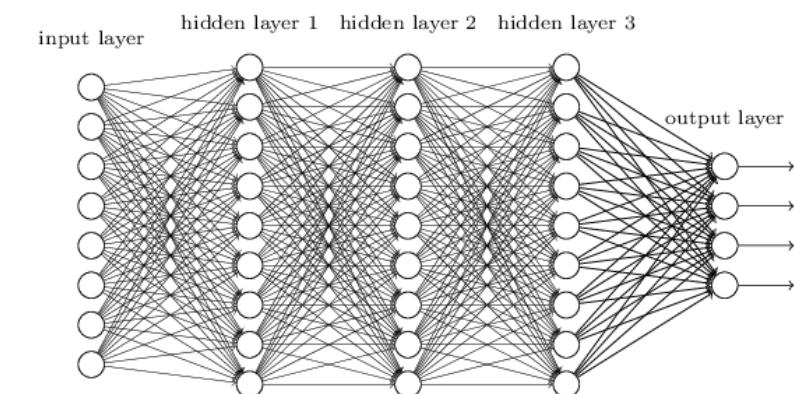
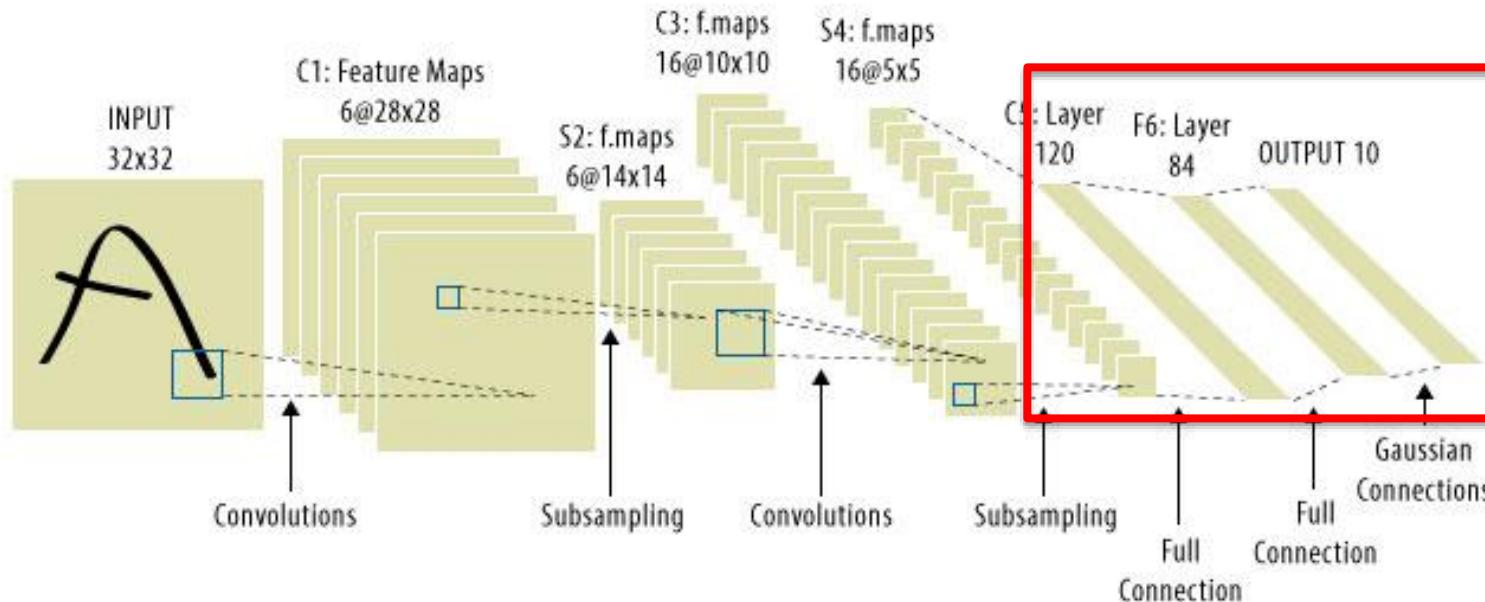
[https://www.researchgate.net/figure/AlexNet-Architecture-To-be-noted-is-copied-2\\_fig1\\_300412100](https://www.researchgate.net/figure/AlexNet-Architecture-To-be-noted-is-copied-2_fig1_300412100)

# CNN for Image



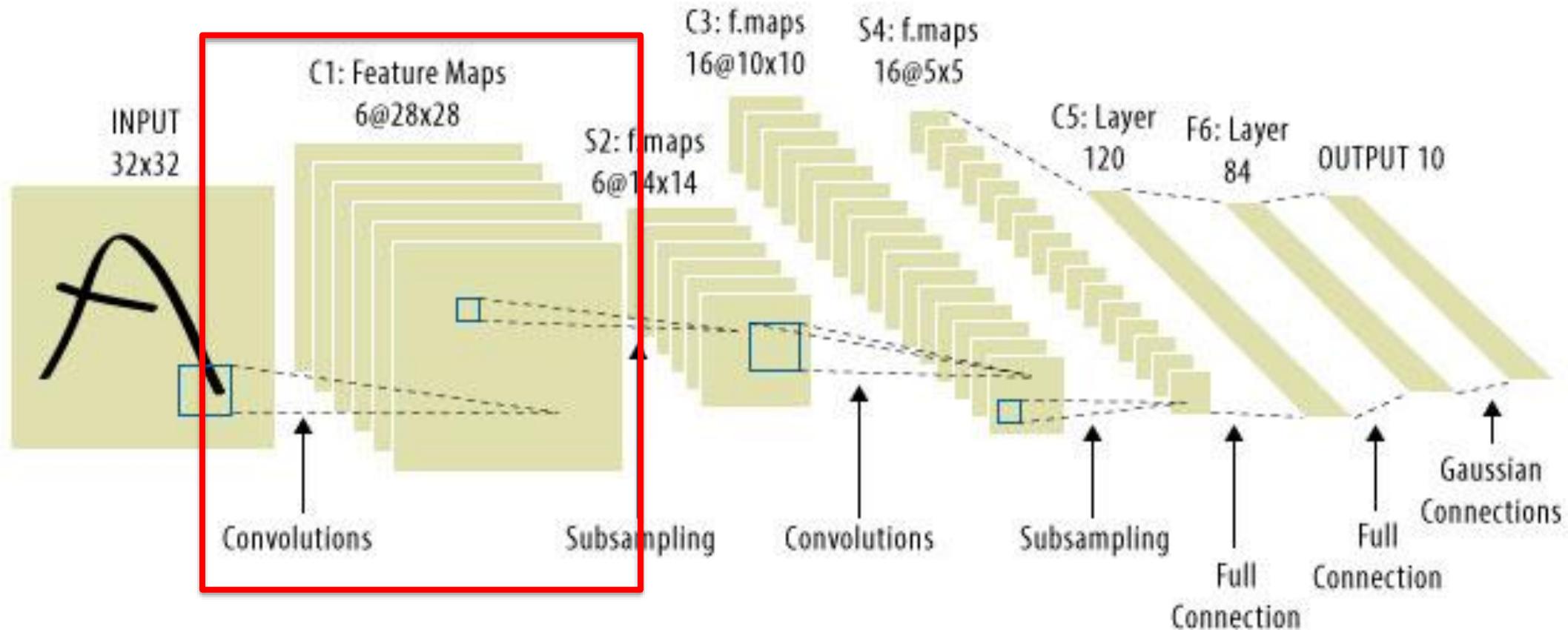
Convolutional Neural Network (LeNet 1998)

# Fully Connection



Convolutional Neural Network (LeNet 1998)

# Convolutions



Convolutional Neural Network (LeNet 1998)

# Image Representation 2: Convolution

|   |   |   |   |   |
|---|---|---|---|---|
| 1<br><small><math>\times 1</math></small> | 1<br><small><math>\times 0</math></small> | 1<br><small><math>\times 1</math></small> | 0 | 0 |
| 0<br><small><math>\times 0</math></small> | 1<br><small><math>\times 1</math></small> | 1<br><small><math>\times 0</math></small> | 1 | 0 |
| 0<br><small><math>\times 1</math></small> | 0<br><small><math>\times 0</math></small> | 1<br><small><math>\times 1</math></small> | 1 | 1 |
| 0   | 0   | 1   | 1 | 0 |
| 0   | 1   | 1   | 0 | 0 |

Image

|   |  |  |
|---|--|--|
| 4 |  |  |
|   |  |  |
|   |  |  |
|   |  |  |

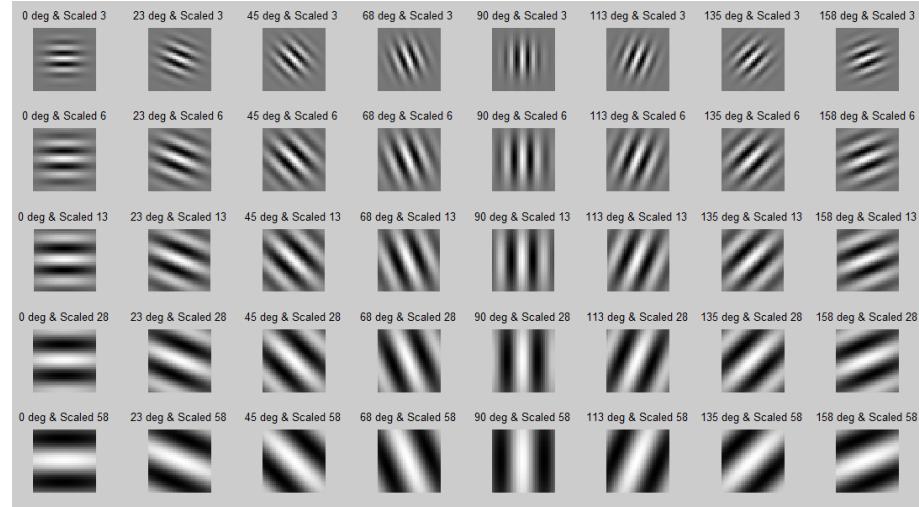
Convolved  
Feature



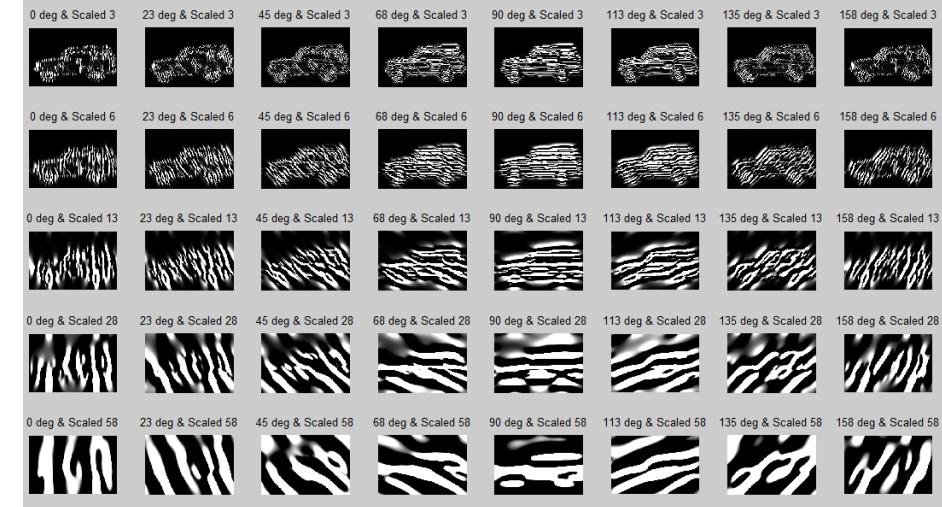
[http://ufldl.stanford.edu/wiki/index.php/Feature\\_extraction\\_using\\_convolution](http://ufldl.stanford.edu/wiki/index.php/Feature_extraction_using_convolution)

[https://leonardoaraujosantos.gitbooks.io/artificial-inteligence/content/convolutional\\_neural\\_networks.html](https://leonardoaraujosantos.gitbooks.io/artificial-inteligence/content/convolutional_neural_networks.html)

# Handcrafted Feature



Gabor Filter

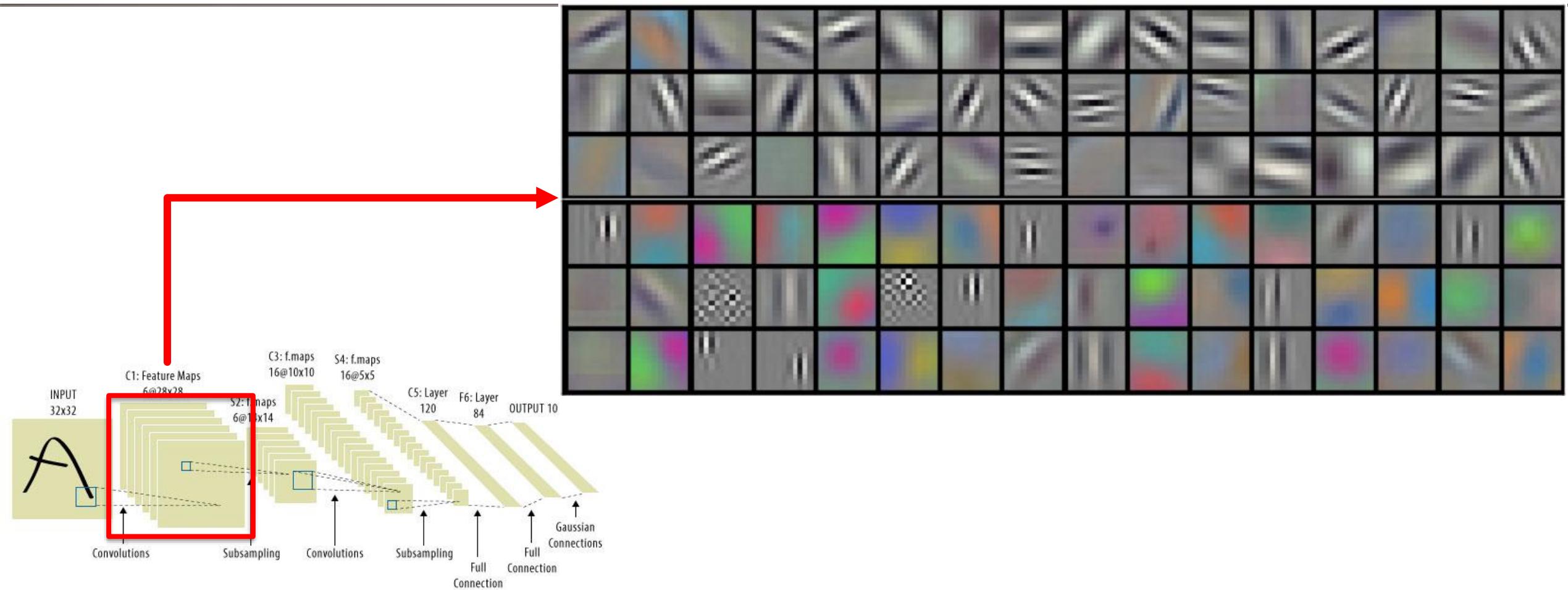


Features

Not as good as human's performance

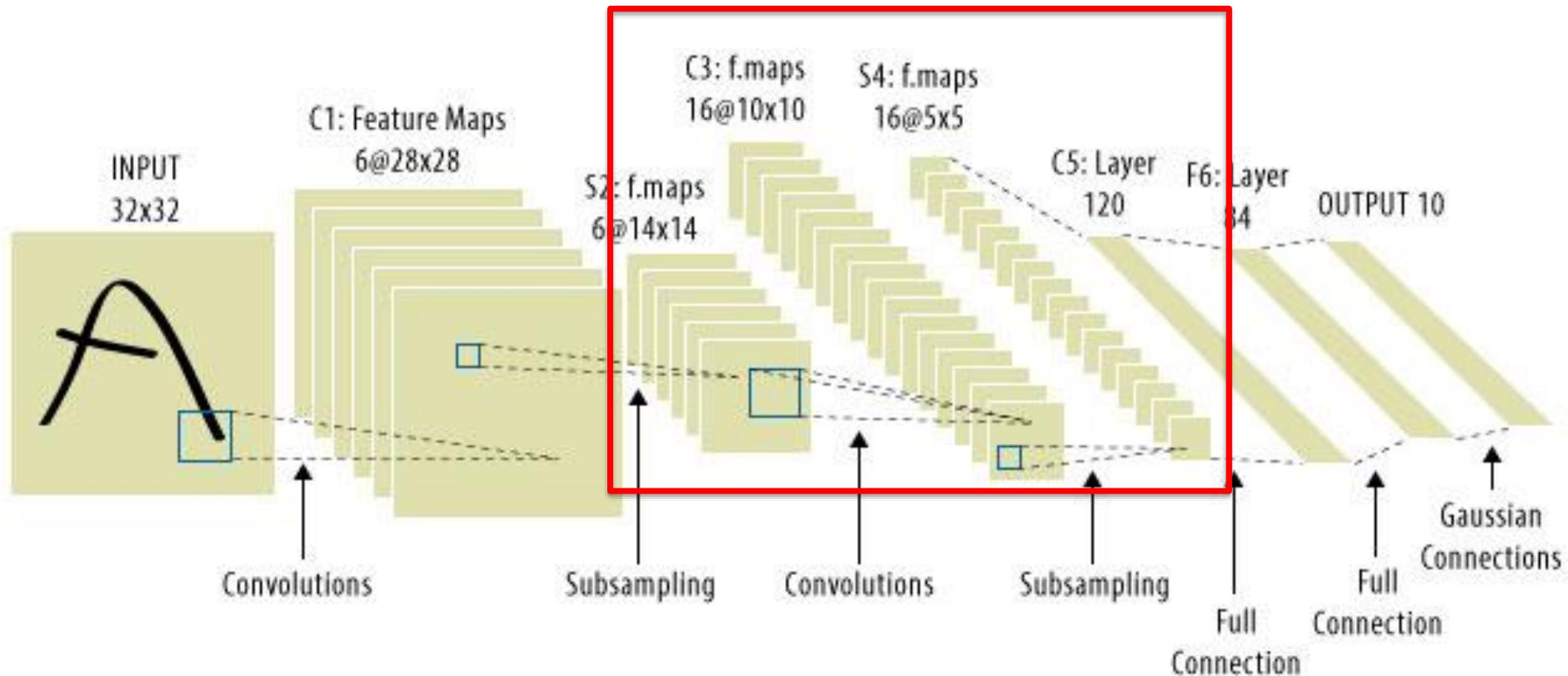
<https://stackoverflow.com/questions/20608458/gabor-feature-extraction>

# Convolutions



<https://arstechnica.com/science/2018/12/how-computers-got-shockingly-good-at-recognizing-images/3/>

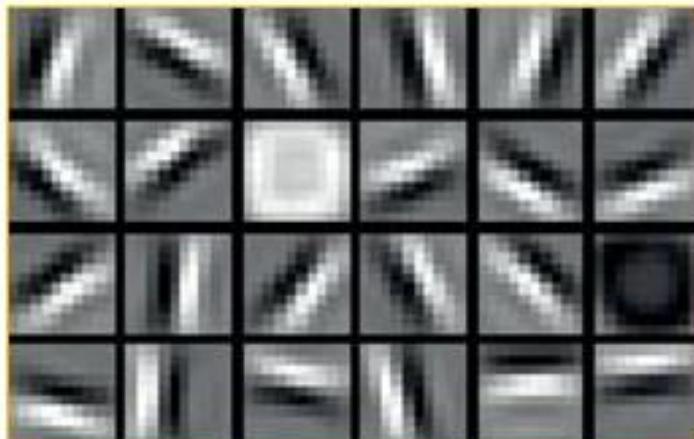
# Why ?



Convolutional Neural Network (LeNet 1998)

# Different Layers

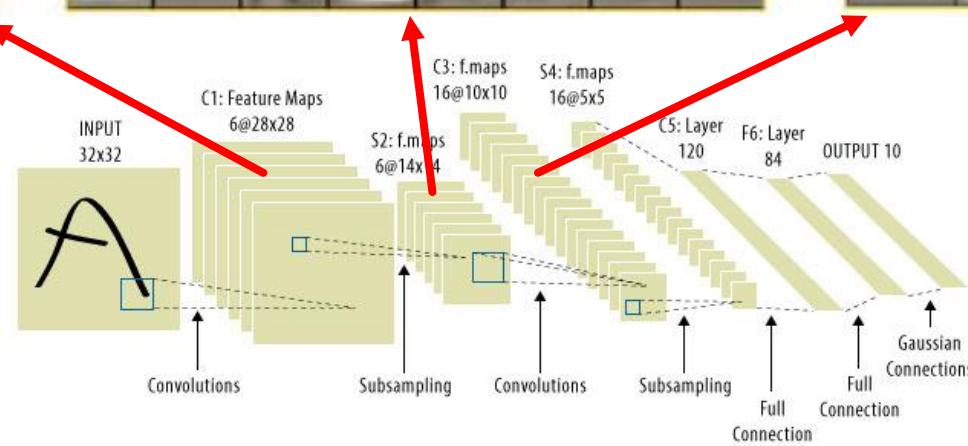
Low-level features



Mid-level features



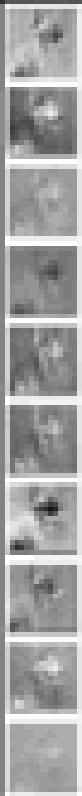
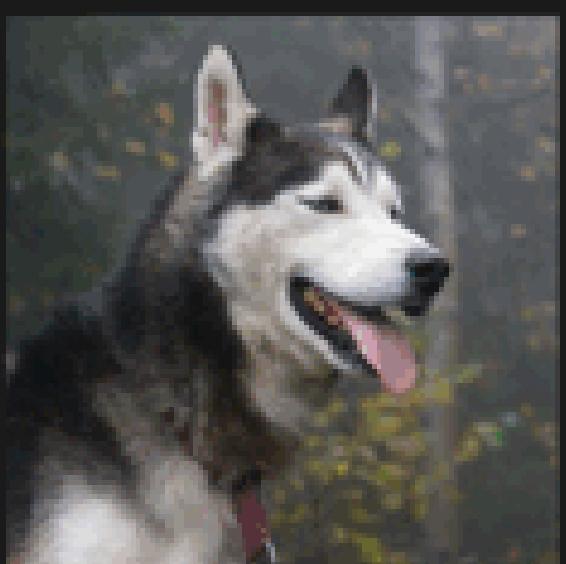
High-level features



<https://www.all-electronics.de/wp-content/uploads/2010/05/AudiA7.jpg>

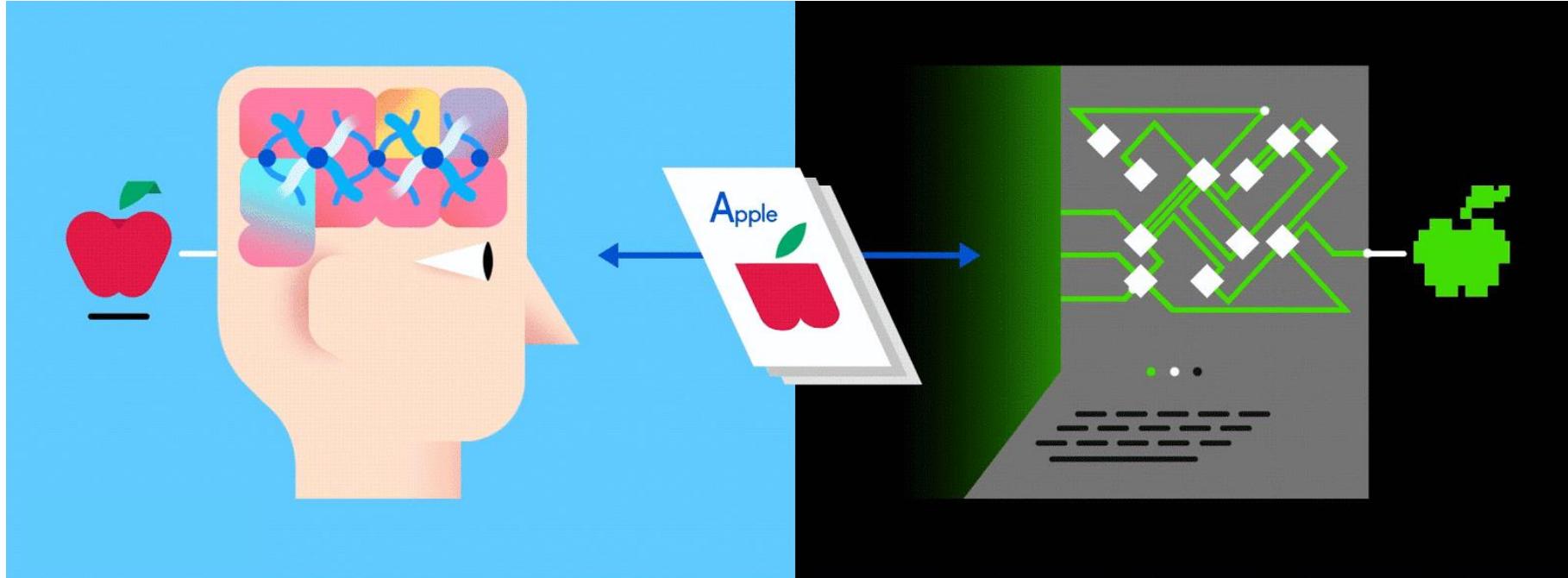
“Audi A7”

# CNN Features



<http://karpathy.github.io/2015/10/25/selfie/>

# Summarize



<https://techcrunch.com/2017/04/13/neural-networks-made-easy/>

Neural network is able to perform  $X \rightarrow Y$  non-linear fitting.  
More layers will have stronger fitting ability.  
CNN provides an efficient way of training a neural network for images.

# In this class



<https://www.empireonline.com/shopping/movies/top-iron-man-merch/>