

# Introduction to Deep Learning

## Ups and Downs of Deep Learning

- 1958, perceptron (linear model)
- 1969: Perceptron has limitation
- 1980: Multi-layer perceptron (X)
  - Do not have significant difference from DNN today
- 1986: Backpropagation
  - Usually more than 3 hidden layers is not helpful
- 1989: 1 hidden layer is 'good enough', why deep learning?
- 2006: RBM initialization (breakthrough)
- 2009: GPU
- 2011: Start to be popular in speech recognition
- 2012: win ILSVRC image competition

## Three Steps for Deep Learning

- Define a set of function (Neural Network)
- Evaluate the performance of function
- Pick up the best function

## Neural Network

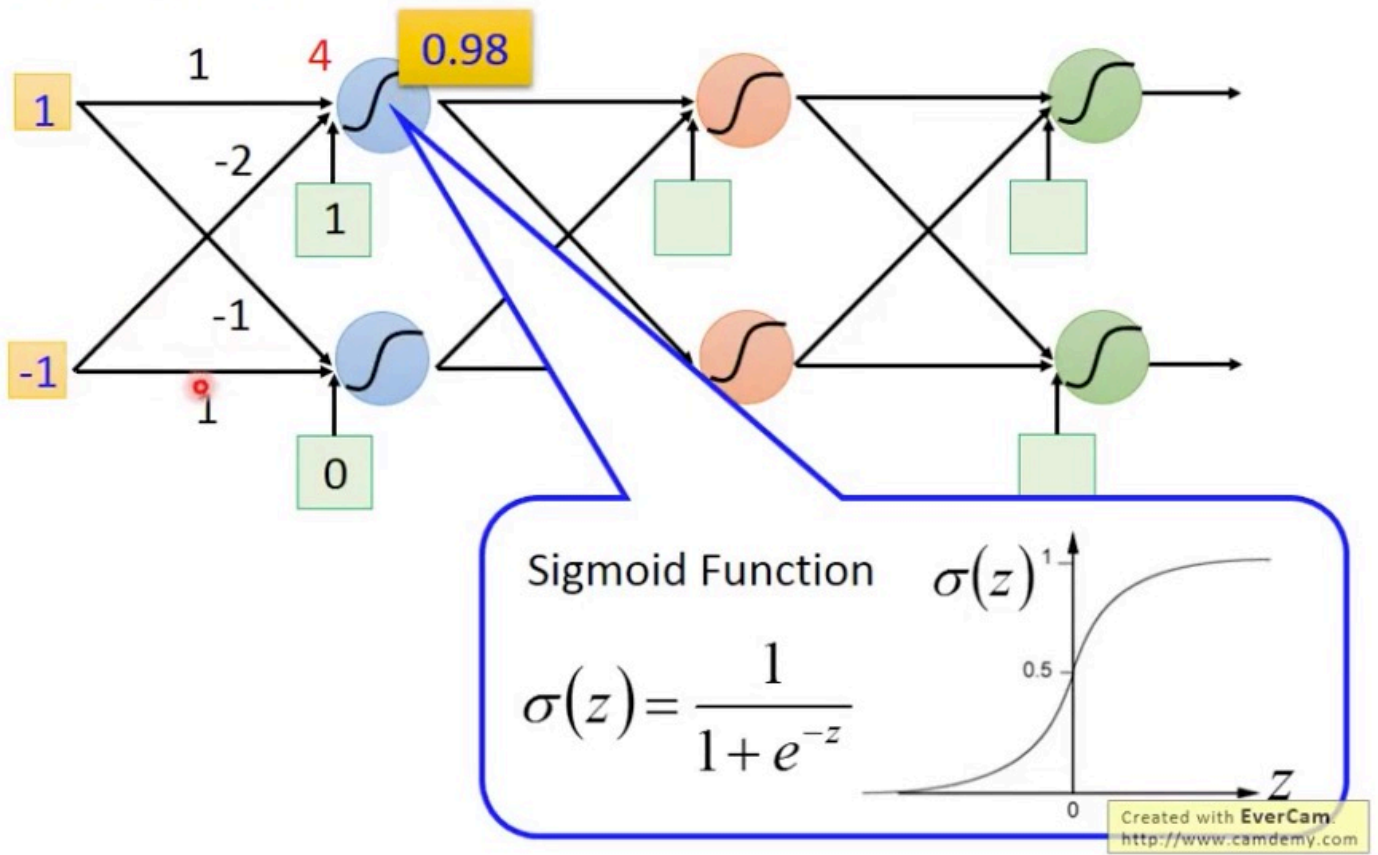
Concatenate different logistic regressions, each logistic regression has its own weight and bias.

Different connection leads to different network structures:

Network parameter theta: all the weights and biases in the 'neurons'

How to connect them?

# Fully Connect Feedforward Network



weight and bias are from training data.

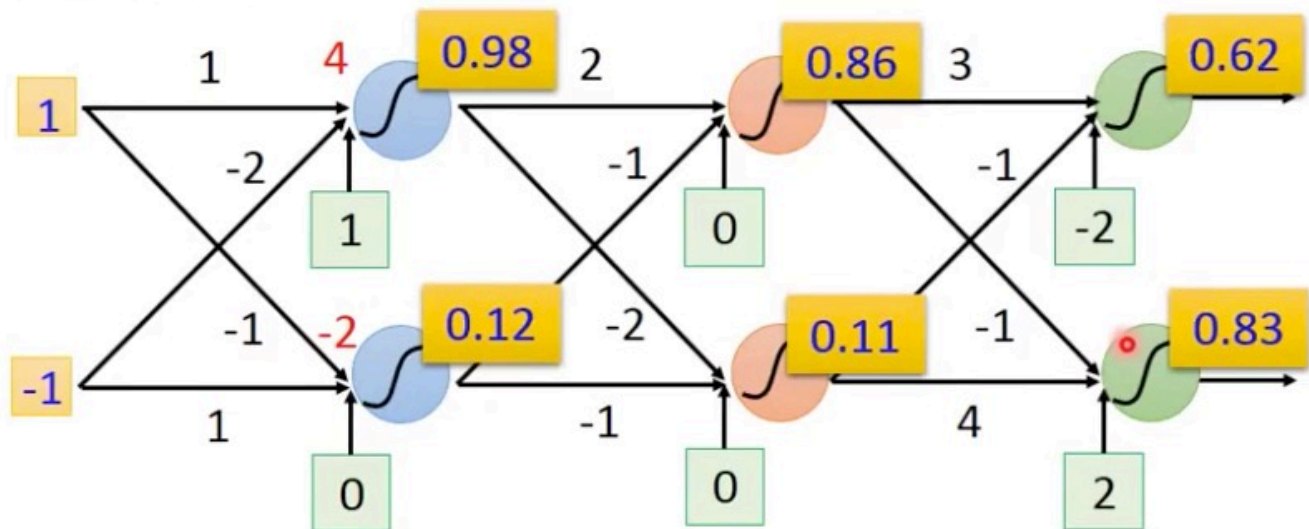
for example, blue top neuron, weight (1, -2), bias(1); blue bottom neuron, weight (-1, 1), bias(0)

top blue output:  $1 \times 1 + (-1) \times (-2) + 1$  (bias) = 4 >>> Pass the sigmoid function >>> 0.98

bottom blue output:  $1 \times (-1) + (-1) \times 1 + 0$  = -2 >>> Pass the sigmoid function >>> 0.12

Continuing ...

# Fully Connect Feedforward Network

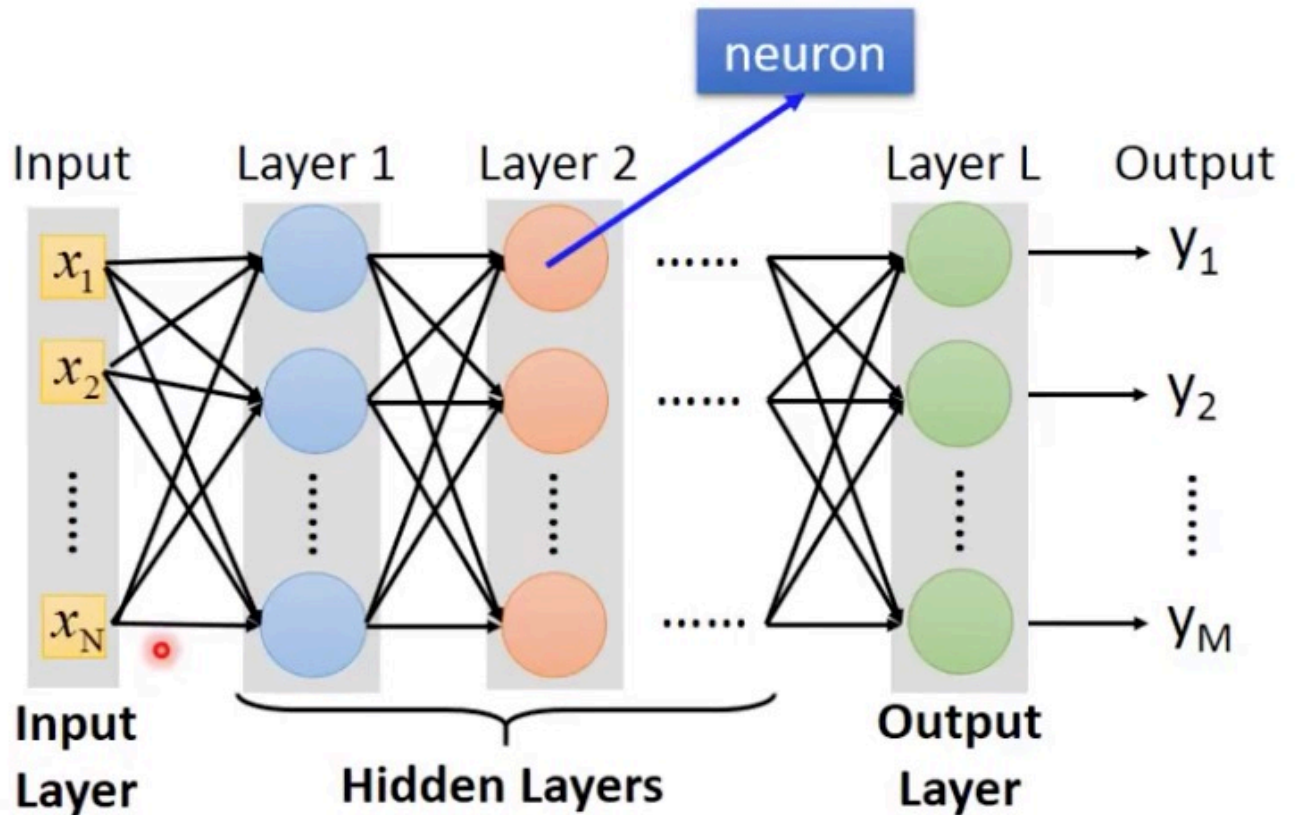


Input is a vector (1, -1), output is a vector (0.62, 0.83)

Given network structure, define a function set

**Types of connections**

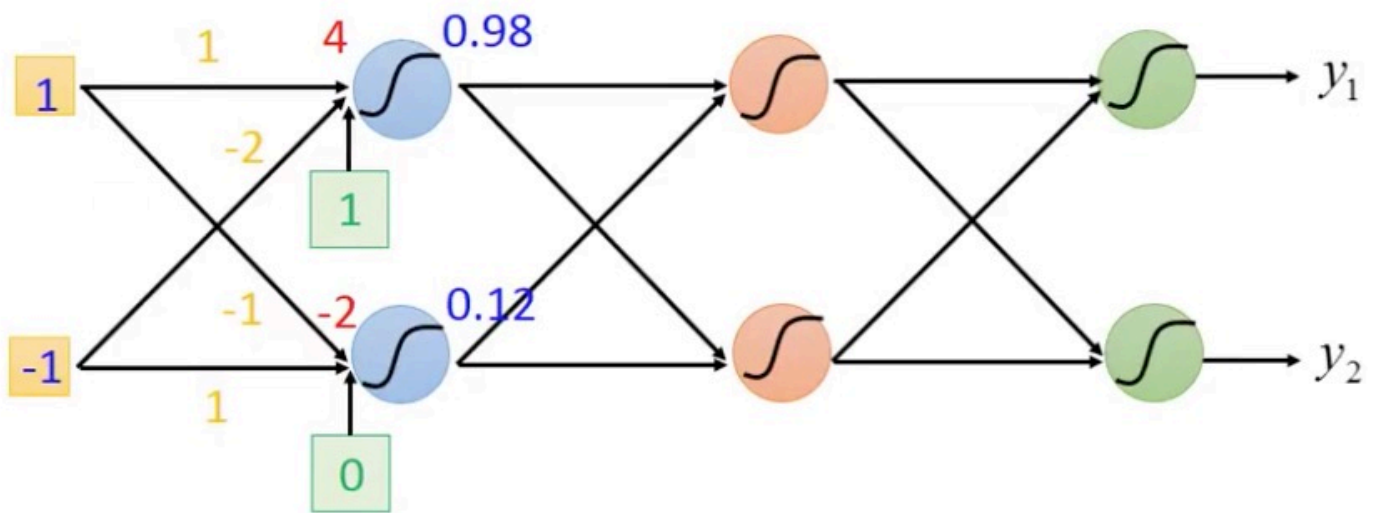
# Fully Connect Feedforward Network



Deep = many hidden layers

## Matrix Operation

# Matrix Operation



$$\sigma\left( \underbrace{\begin{bmatrix} 1 & -2 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix}}_{\begin{bmatrix} 4 \\ -2 \end{bmatrix}} \right) = \begin{bmatrix} 0.98 \\ 0.12 \end{bmatrix}$$

Feedforward Network Calculation Process:

(input multiply by matrix (weight) + bias ) >> into sigmoid (any activation function) >>> output