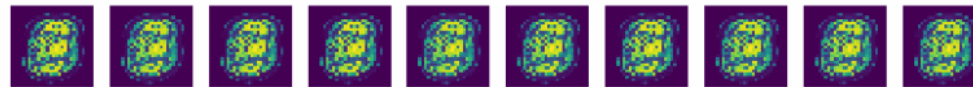




# Neural Network

July 11, 2017

Seung-Chan Kim, Ph. D





- 1. 머신러닝 개론 및 주요 개념의 이해. Tensorflow 시스템 설치 및 환경설정 (7/4 화)
- 2. Tensorflow 에 익숙해지기 실습 및 Regression의 이해 (7/6 목)
- **3. Neural Network 이해 및 tensorflow 를 이용한 구현 (7/11 화)**
- 4. 이미지 분류 이해 및 Tensorflow를 이용한 구현 (7/13 목)

<https://github.com/dalek7/DLWorkshop17Summer>



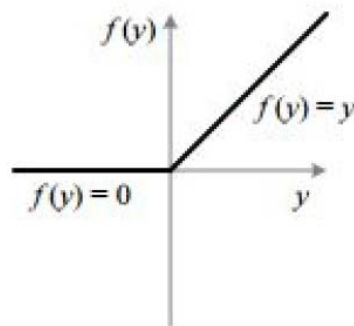
## 복습

- Optimization
- Regression
- Rectified Linear Unit (ReLU)



# Rectified Linear Unit (ReLU)

$$\begin{aligned} a_l &= \text{relu}(a_l) \\ b_l &= \text{relu}(b_l) \\ c_l &= \text{relu}(c_l) \end{aligned}$$



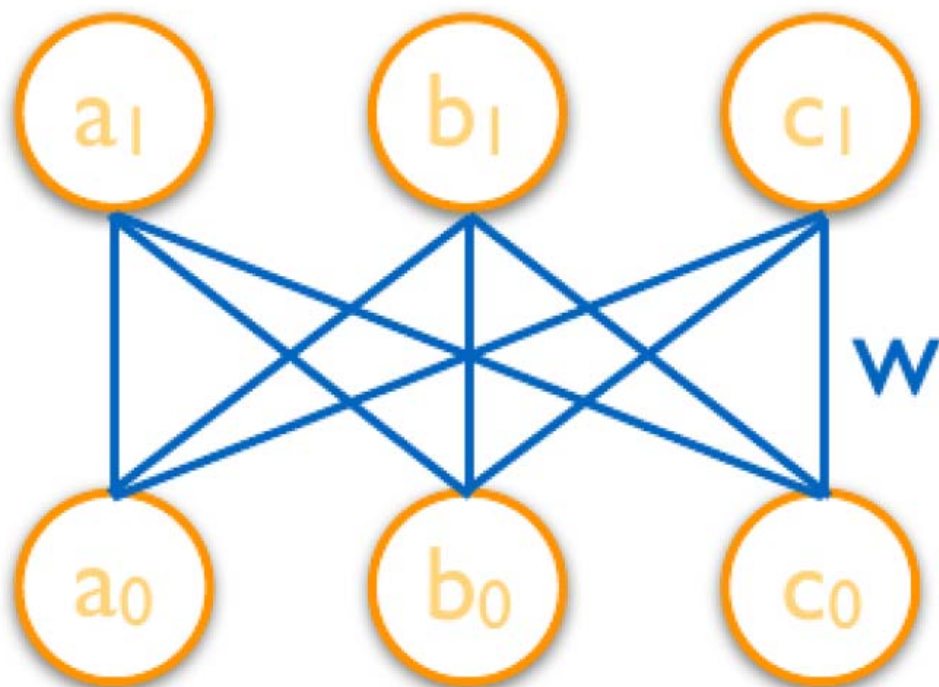
```
out = tf.nn.relu(y)
```

( -5 ,	'-->' ,	)
( -4 ,	'-->' ,	)
( -3 ,	'-->' ,	)
( -2 ,	'-->' ,	)
( -1 ,	'-->' ,	)
( 0 ,	'-->' ,	)
( 1 ,	'-->' ,	)
( 2 ,	'-->' ,	)
( 3 ,	'-->' ,	)
( 4 ,	'-->' ,	)

00-5-relutest.py 를 열어주세요



# A simple Rectified Linear Unit (ReLU) network

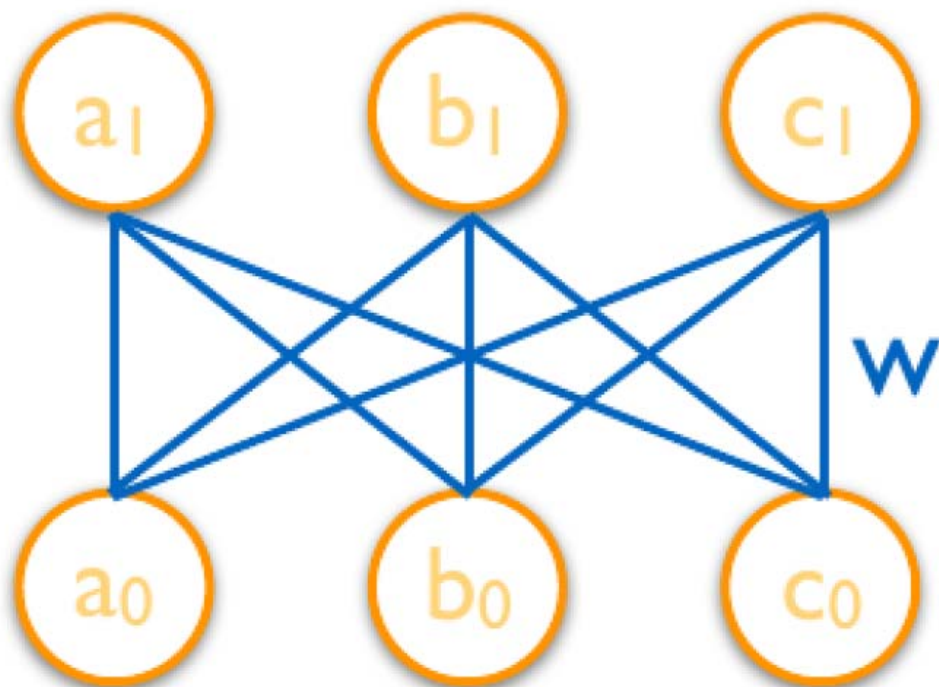


$$\begin{matrix} & \mathbf{x} & & \mathbf{w} & & \mathbf{y} \\ & \begin{bmatrix} a_0 & b_0 & c_0 \end{bmatrix} & \cdot & \begin{bmatrix} w_{a,a} & w_{a,b} & w_{a,c} \\ w_{b,a} & w_{b,b} & w_{b,c} \\ w_{c,a} & w_{c,b} & w_{c,c} \end{bmatrix} & = & \begin{bmatrix} a_1 & b_1 & c_1 \end{bmatrix} \end{matrix}$$

$y = \text{tf.matmul}(x, w)$



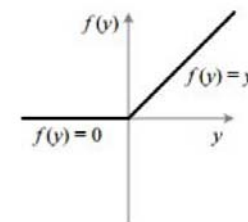
# A simple Rectified Linear Unit (ReLU) network



$$\begin{matrix} \mathbf{x} & & \mathbf{w} \\ \begin{bmatrix} a_0 & b_0 & c_0 \end{bmatrix} & \cdot & \begin{bmatrix} w_{a,a} & w_{a,b} & w_{a,c} \\ w_{b,a} & w_{b,b} & w_{b,c} \\ w_{c,a} & w_{c,b} & w_{c,c} \end{bmatrix} & = & \begin{bmatrix} a_1 & b_1 & c_1 \end{bmatrix} \end{matrix}$$

`y = tf.matmul(x, w)`

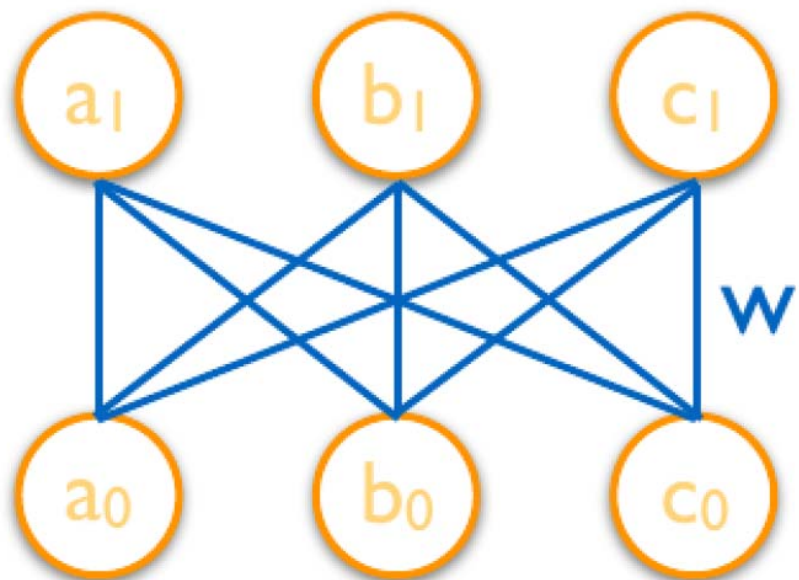
$$\begin{aligned} a_1 &= \text{relu}(a_1) \\ b_1 &= \text{relu}(b_1) \\ c_1 &= \text{relu}(c_1) \end{aligned}$$



`out = tf.nn.relu(y)`



# A simple Rectified Linear Unit (ReLU) network



```
import tensorflow as tf
sess = tf.Session()
x = tf.placeholder("float", [1, 3])
w = tf.Variable(tf.random_normal([3, 3]), name='w')
y = tf.matmul(x, w)
relu_out = tf.nn.relu(y)
```

01-3-simplenetwork.py 를 열어주세요



# Why Neural Networks?

- Consider humans
  - Neuron switching time  $\sim .001$  second
  - Number of neurons  $\sim 10^{10}$
  - Connections per neuron  $\sim 10^{4\sim 5}$
  - Scene recognition time  $\sim .1$  second
  - 100 inference steps doesn't seem like enough  
 $\Rightarrow$  massively parallel computation
- Properties of artificial neural nets (ANN)
  - Many neuron-like threshold switching
  - Many weighted interconnections among units
  - Highly parallel, distributed process
  - Emphasis on tuning weights automatically
- Other names: connectionism, parallel distributed processing, neural computation





# When to Consider Neural Networks?

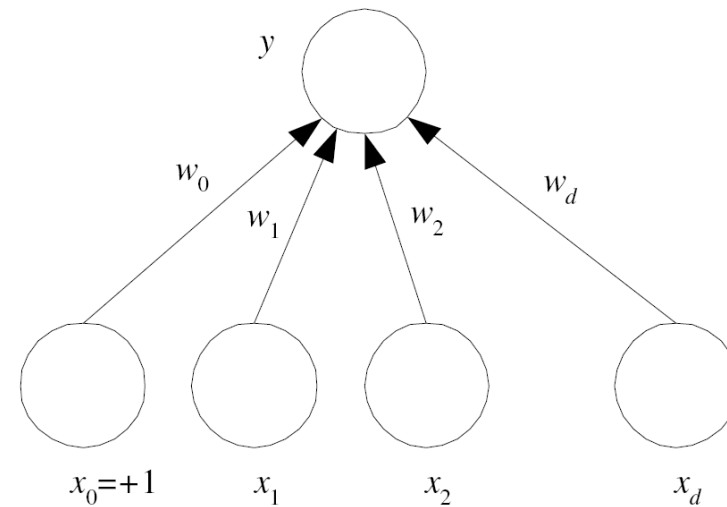
- Input is high-dimensional discrete or real-valued (e.g., raw sensor input)
- Output is discrete or real-valued
- Output is a vector of values
- Possibly noisy data
- Form of target function is unknown
- Human readability of result is unimportant
- Examples
  - Speech phoneme recognition
  - Image classification
  - Financial prediction



# Single-Layer Perceptron

- Classification
  - Sigmoid activation function

$$y = \text{sigmoid}(in) = \frac{1}{1 + \exp[-\mathbf{w}^T \mathbf{x}]}$$



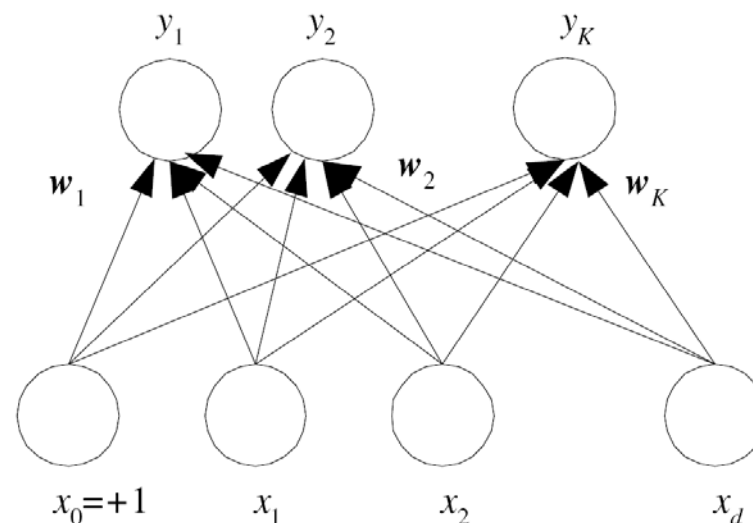


# Single-Layer Perceptron with K Outputs

- Classification
  - Sigmoid activation function

$$y_i = \text{sigmoid}(in_i) = \frac{1}{1 + \exp[-\mathbf{w}_i^T \mathbf{x}]}$$

choose  $C_i$  if  $y_i = \max_k y_k$



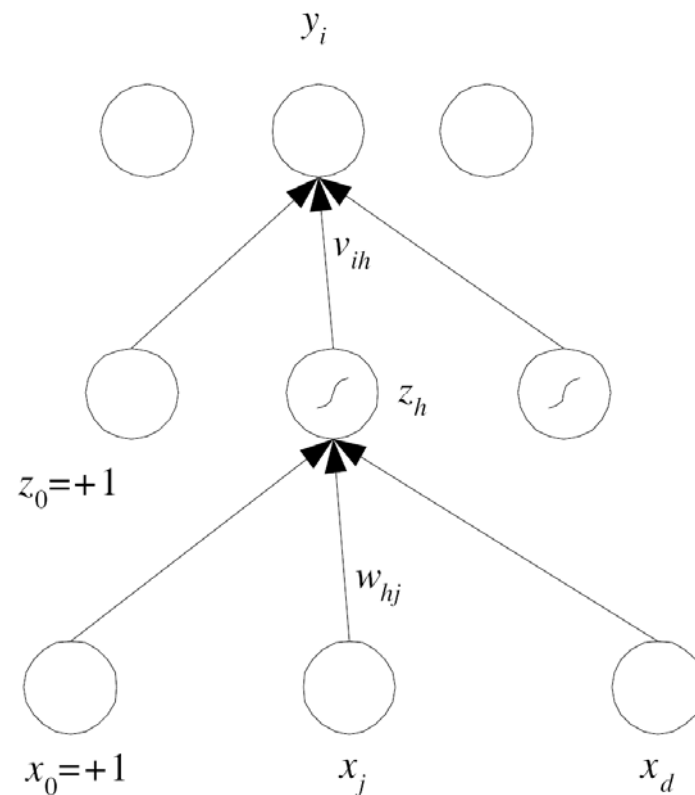


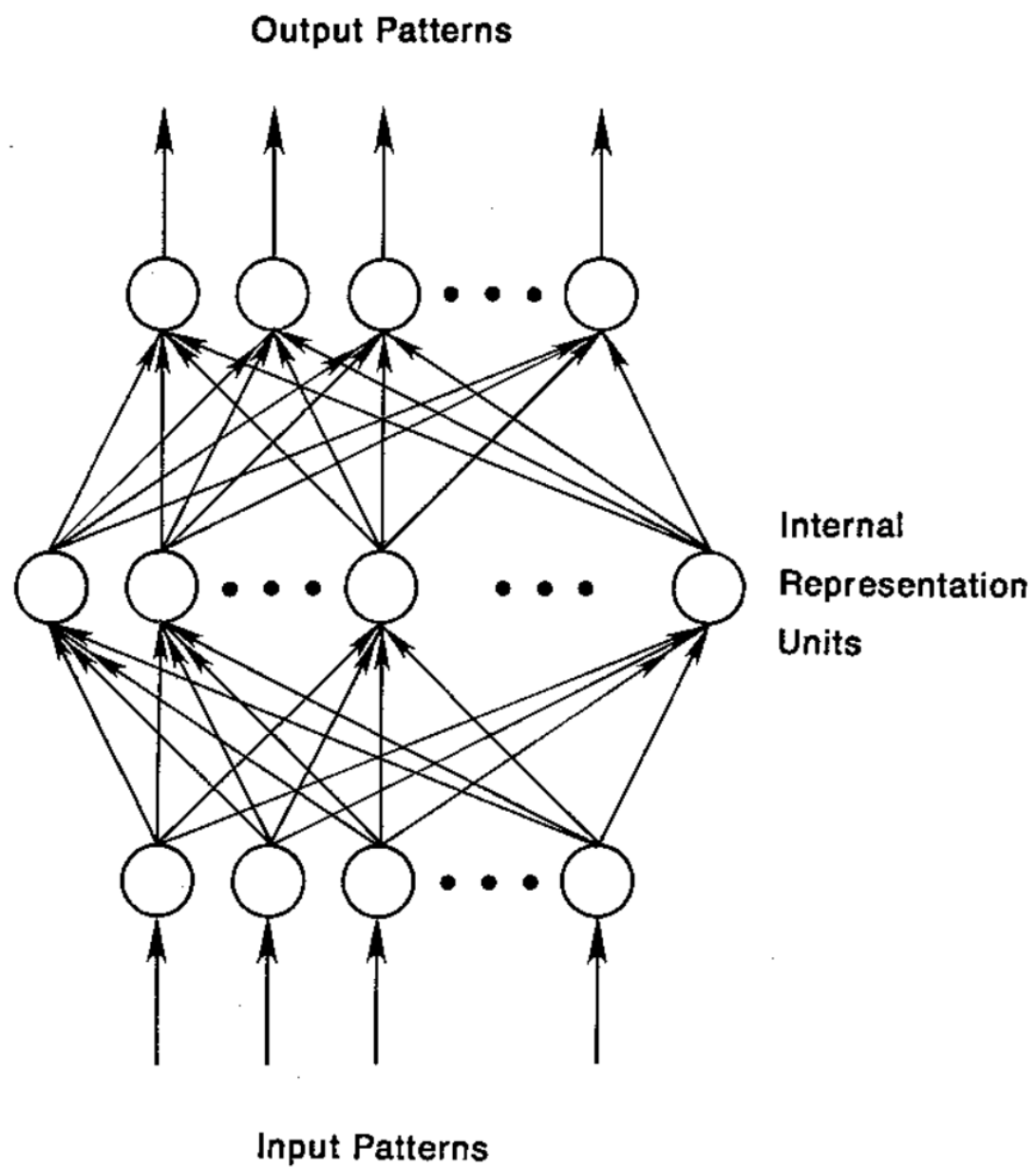
# Multi-layer Perceptrons

- [Rumelhart et al., 1986]

$$y_i = \mathbf{w}_i^T \mathbf{z} = \sum_{h \in H} w_{hi} z_h + w_{0i}$$

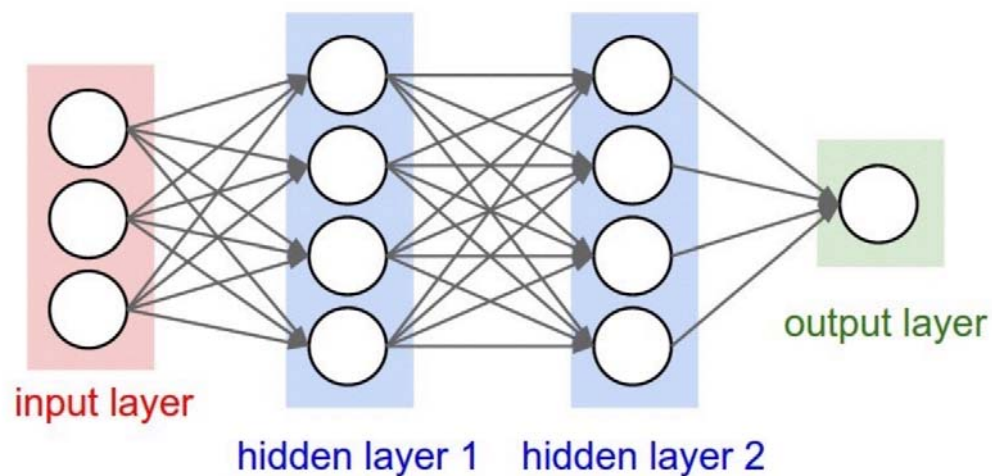
$$z_h = a_h(\mathbf{w}_h^T \mathbf{x})$$
$$= \frac{1}{1 + \exp[-(\sum_{j=1}^d w_{jh} x_j + w_{0h})]}$$







# No one on earth had found a viable way to train !



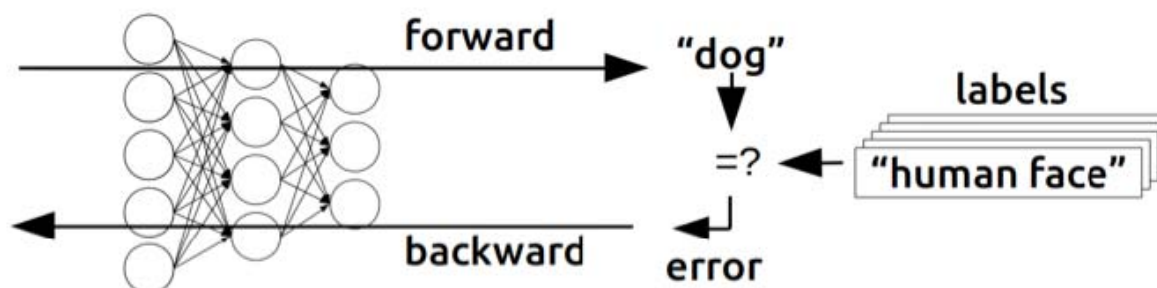
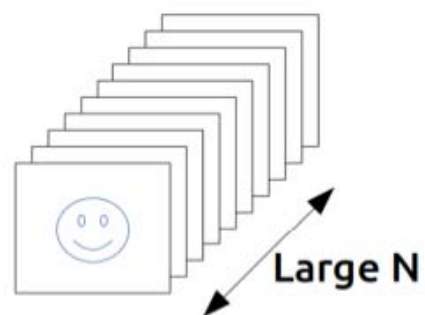
\*Marvin Minsky, 1969

<http://cs231n.github.io/convolutional-networks/>

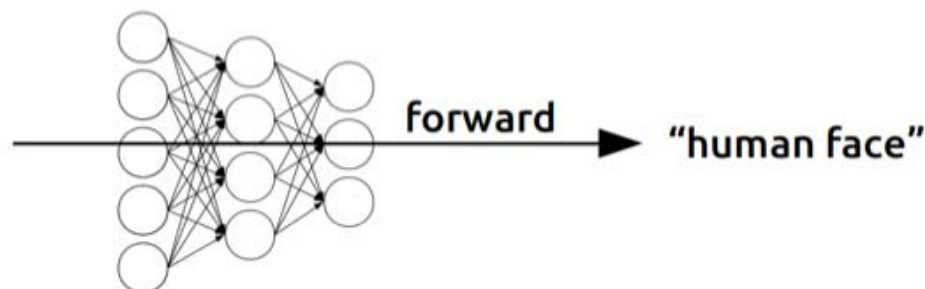
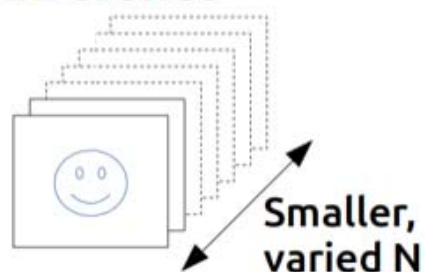


# Backpropagation (1974, 1982 by Paul Werbos, 1986 by Hinton)

## Training



## Inference



<https://devblogs.nvidia.com/parallelforall/inference-next-step-gpu-accelerated-deep-learning/>



$$\text{cost} = \frac{1}{m} \sum_{i=1}^m (wx^{(i)} - y^{(i)})^2$$

- W는 어떻게 업데이트 하는가?
  - learning rate  $\alpha$ 에 cost function을  $w$ 에 대하여 미분한 값, 즉 기울기를 곱한 값을 빼는 것이다.

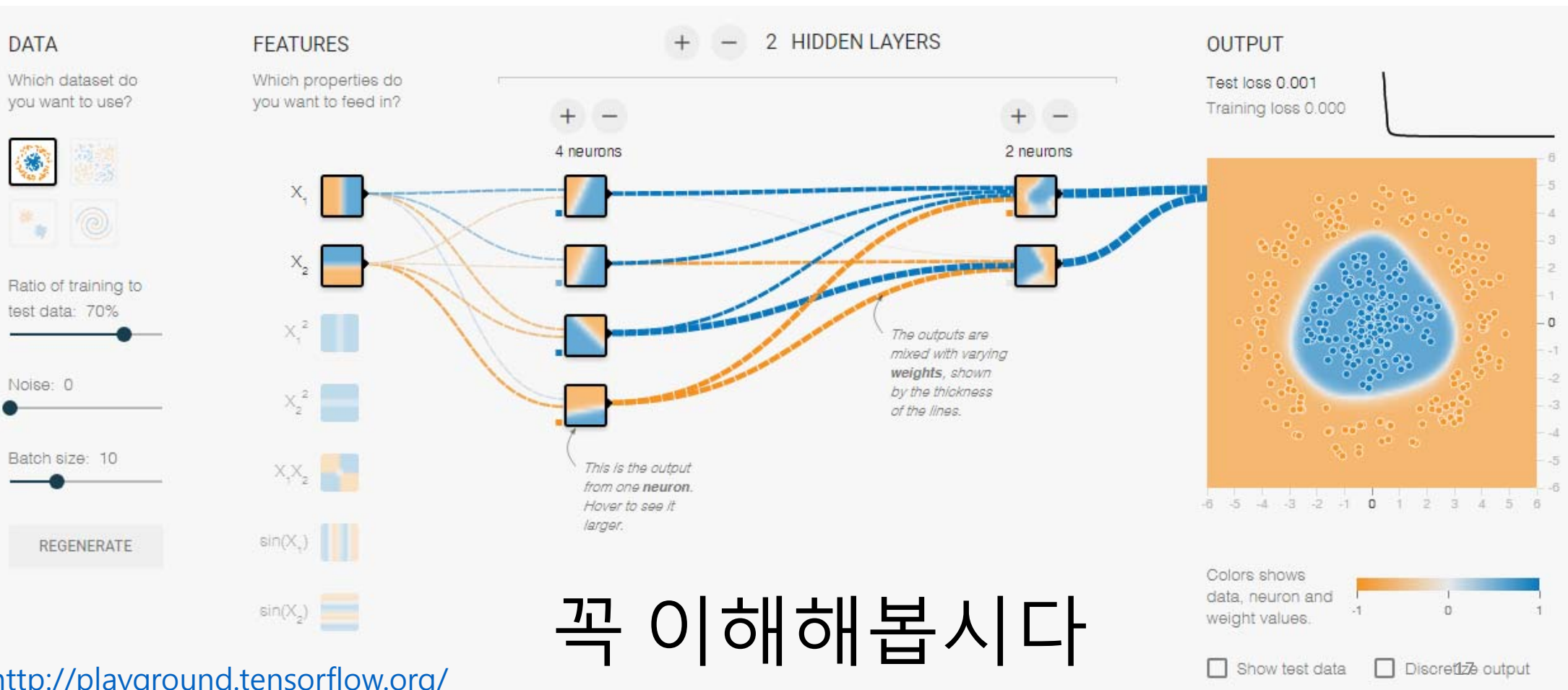
$$\begin{aligned} w &:= w - \alpha \frac{\partial}{\partial w} \text{cost}(w) = w - \alpha \frac{\partial}{\partial w} \frac{1}{2m} \sum_{i=1}^m (wx^{(i)} - y^{(i)})^2 \\ &= w - \alpha \frac{1}{2m} \sum_{i=1}^m 2 (wx^{(i)} - y^{(i)}) x^{(i)} \\ &= w - \alpha \frac{1}{m} \sum_{i=1}^m (wx^{(i)} - y^{(i)}) x^{(i)} \end{aligned}$$

자세한 계산 방법은 2학기 "인공지능 개론" 수업에서 다룸





# Neural Network : playground.tensorflow.org



<http://playground.tensorflow.org/>



## Recap - XOR 문제 !!



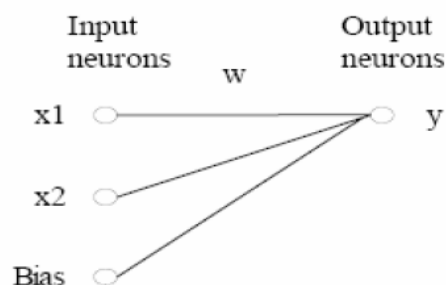
Input Patterns		Output Patterns
00	→	0
01	→	1
10	→	1
11	→	0

XOR classification using single and two layer neural networks

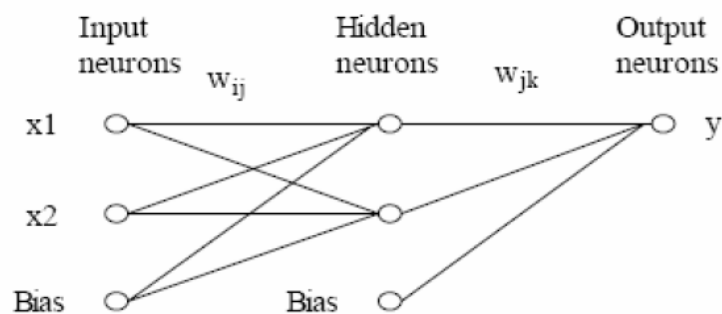


# XOR classification using single and two layer neural networks

## ◆ Single layer networks (ADALINE)

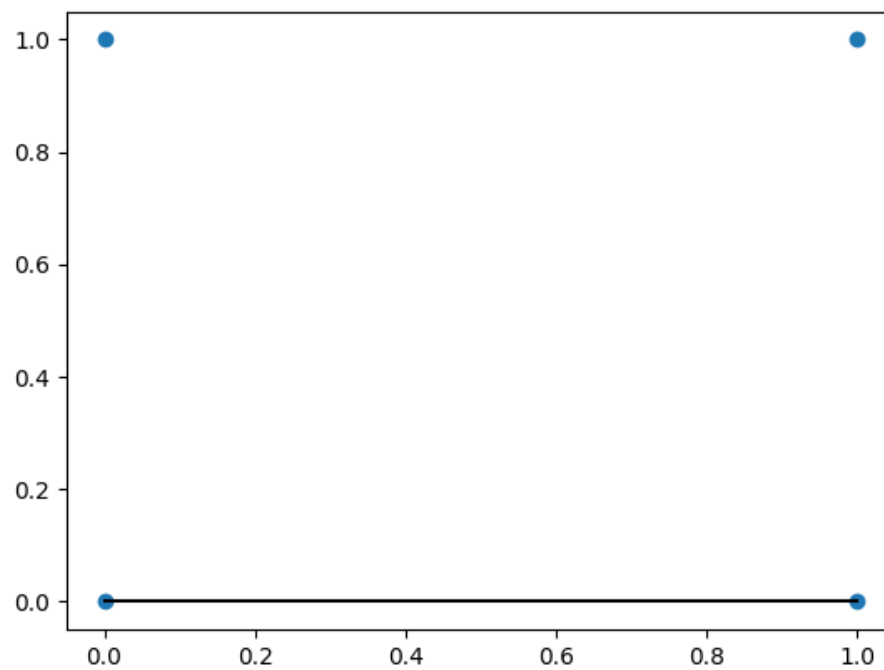


## ◆ An example of two layer neural networks: 2-2-1 structure

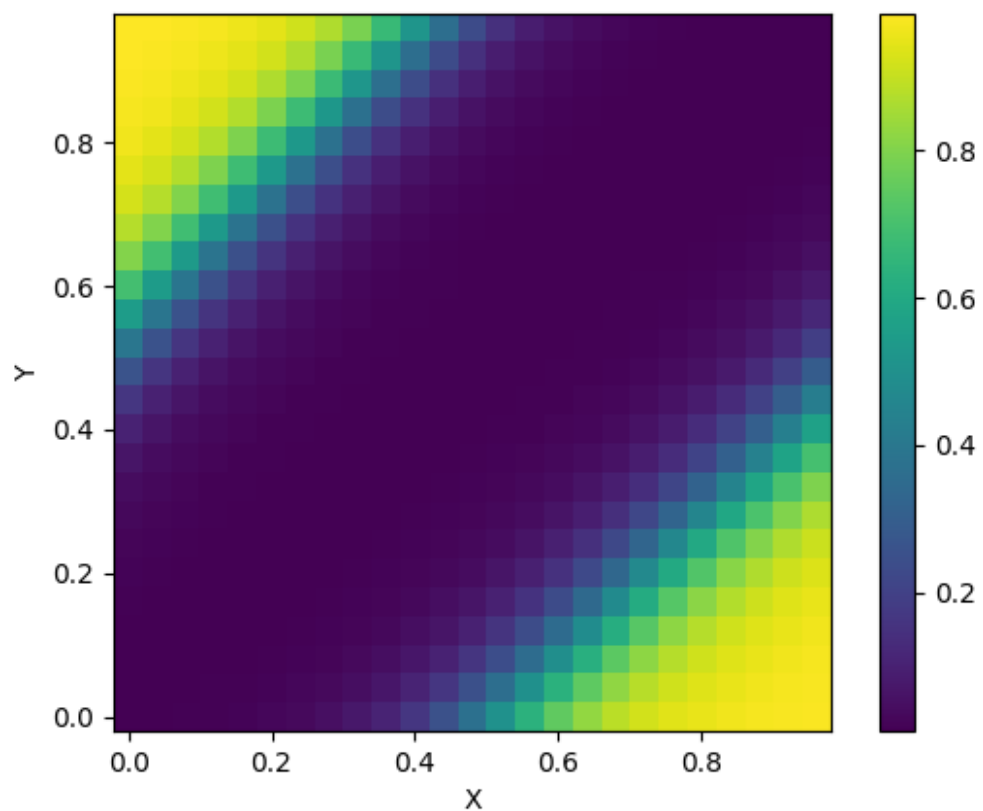




Input Patterns		Output Patterns
00	→	0
01	→	1
10	→	1
11	→	0



03-1-xor-simple.py 를 열어주세요



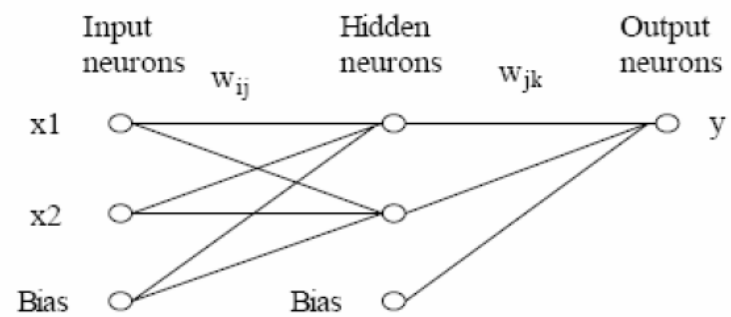
```
0 [[ 0.01338216]] [[ 0.]]  
1 [[ 0.98166394]] [[ 1.]]  
2 [[ 0.98809403]] [[ 1.]]  
3 [[ 0.01135799]] [[ 0.]]
```



03-2-xor-layers.py 를 열어주세요



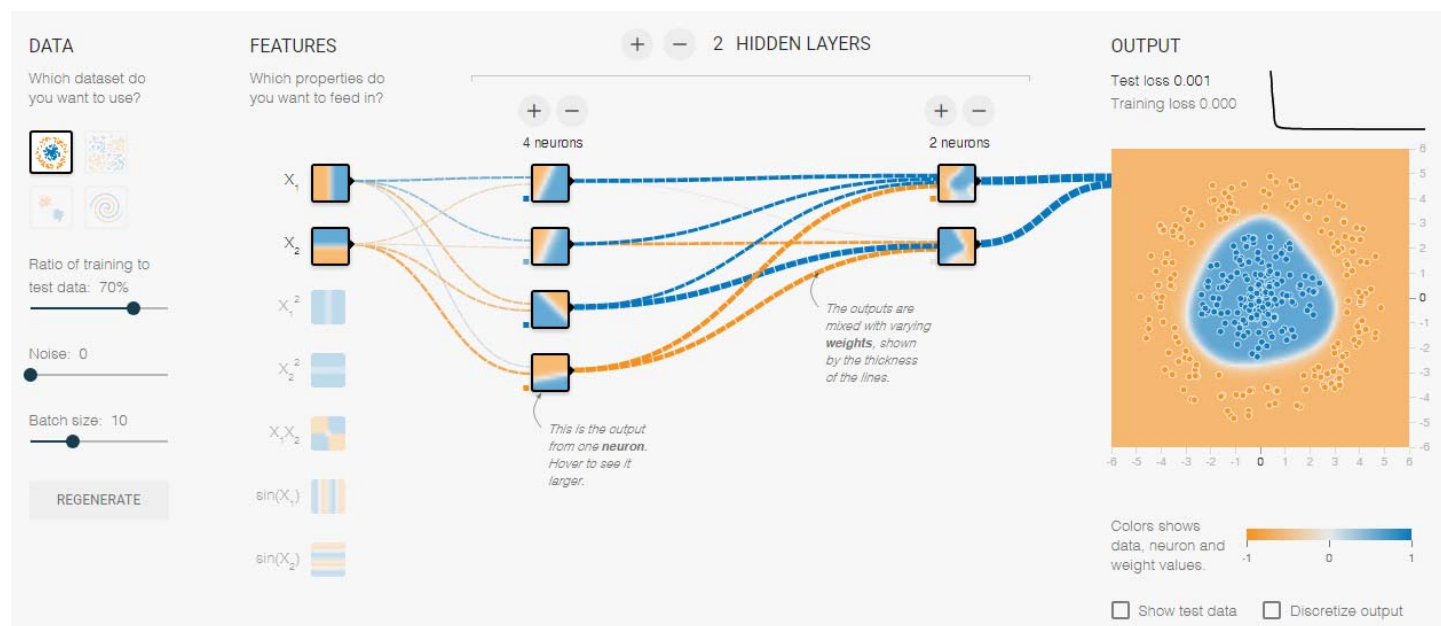
◆ An example of two layer neural networks: 2-2-1 structure





# HW

- Neural Network (playground.tensorflow.org) 사이트에서 문제 하나를 고른후 layer 개수, hidden state 개수를 자유롭게 선택하여, 문제 해결





# Acknowledgement



모두를 위한 머신러닝/딥러닝 강의



Seung-han

Seung-Chan

Jeung-Chan

감사합니다.