

# Machine Learning

Learn Machine Learning with Microsoft Azure **ML Studio**

**Microsoft Student Partner**

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## ✓ Machine Learning 의 등장

- Limitations of Explicit Programming
- Explicit Programming : 개발자가 입력 조건과 프로그램 상태 조건에 따라 소프트웨어가 동작하는 방식을 직접 구현하는 프로그래밍 방식
- Complex Problem : Spam mail filtering, Autonomous vehicle?

## ✓ Machine Learning 이란?

- “Machine Learning is a field of computer science that gives computers the ability to learn without being explicitly programmed.” By. Arthur Samuel (1959)

# Machine Learning



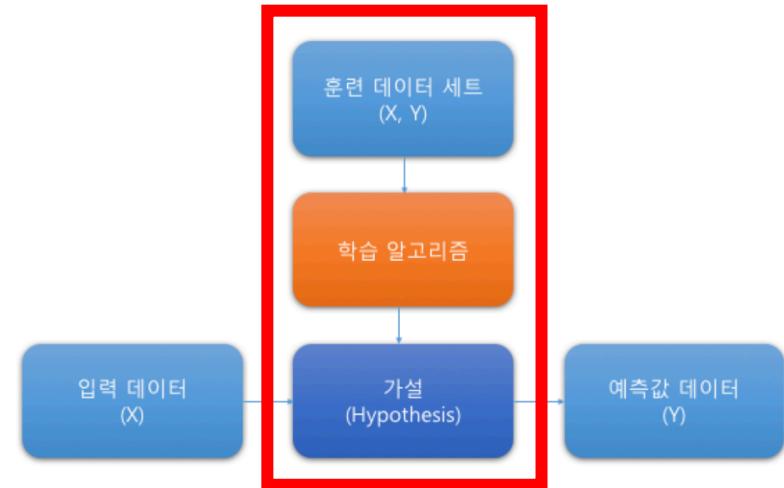
## ✓ 용어

- **Training Data Set**  
 $(X, Y)$  = (메일 내용, 스팸 여부)
- **Input Data**  
 $X$  = 새로운 메일 내용
- **Predicted Value**  
 $Y$  = 스팸 여부
- **Hypothesis (Model)**  
 $f(X) = Y$
- **Target function**

✓ 오차(Error) : 예측값과 목표값의 차이

✓ 오차를 수정하는 방식

- 최소제곱법(method of least squares—LSM or LMS)
- 최우추정법(maximum likelihood method—MLE)
- 최대사후확률추정(maximum a posterior estimation—MAP)
- 기울기 하강법(gradient descent—GD)
- 역전파(backpropagation—BP)



# Machine Learning

## ✓ Supervised Learning

- learn with labeled examples - training set

- Labeled data
- Direct feedback
- Predict outcome/future

## ✓ Unsupervised Learning

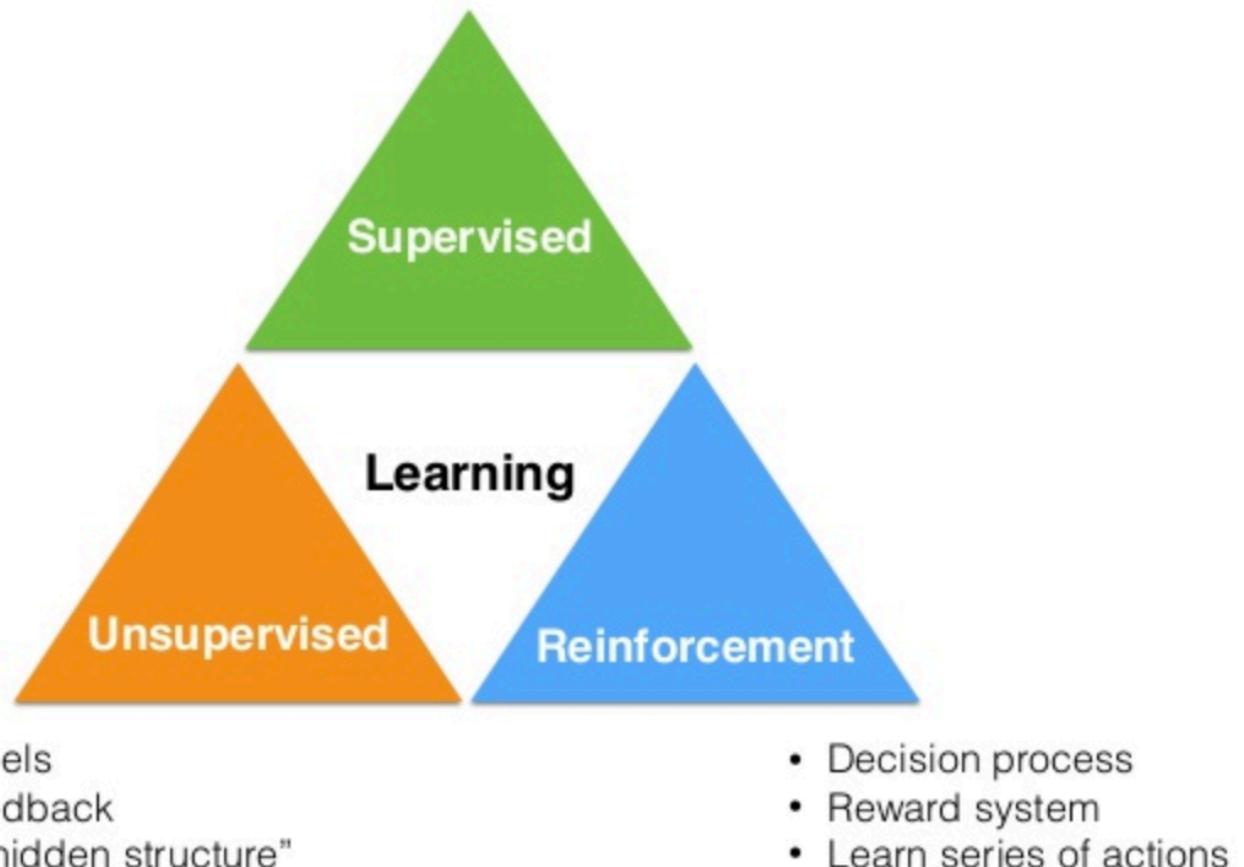
- learn with unlabeled data

## ✓ Semi-supervised Learning

- Supervised + Unsupervised

## ✓ Reinforcement Learning

- reward feedback required by each behavior



## ✓ Supervised Learning



A grid of handwritten digits from 0 to 9, arranged in 10 rows and 10 columns. Each digit is written in a different style, representing a sample of data for training a machine learning model.

0	0	0	0	0	0	0	0	0	0
1	1	1	1	1	1	1	1	1	1
2	2	2	2	2	2	2	2	2	2
3	3	3	3	3	3	3	3	3	3
4	4	4	4	4	4	4	4	4	4
5	5	5	5	5	5	5	5	5	5
6	6	6	6	6	6	6	6	6	6
7	7	7	7	7	7	7	7	7	7
8	8	8	8	8	8	8	8	8	8
9	9	9	9	9	9	9	9	9	9

✓ Classification (For Discrete Value)

✓ Regression (For Continuous Value)

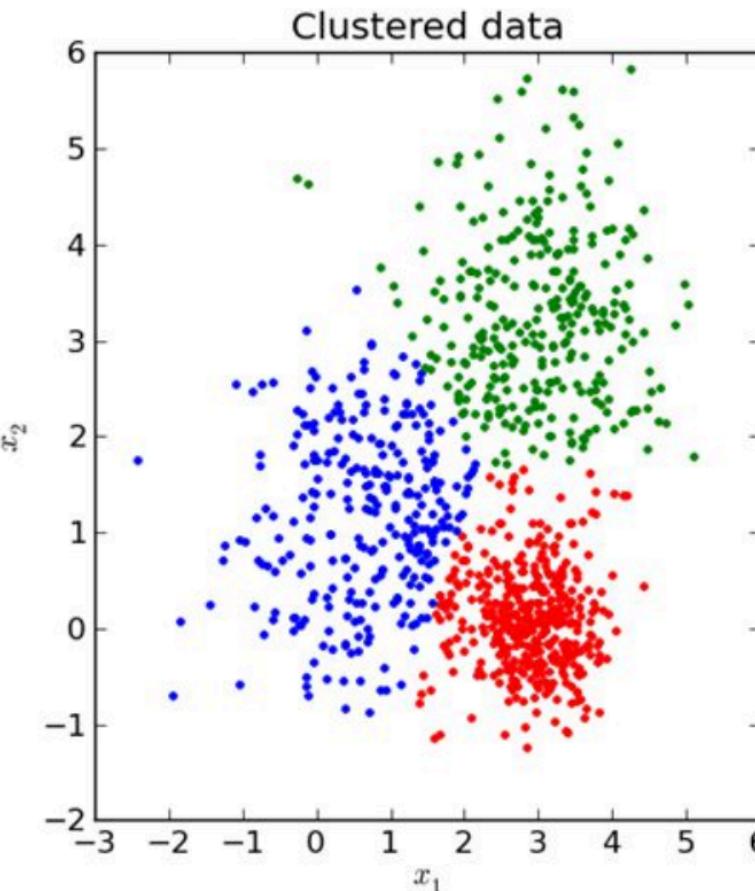
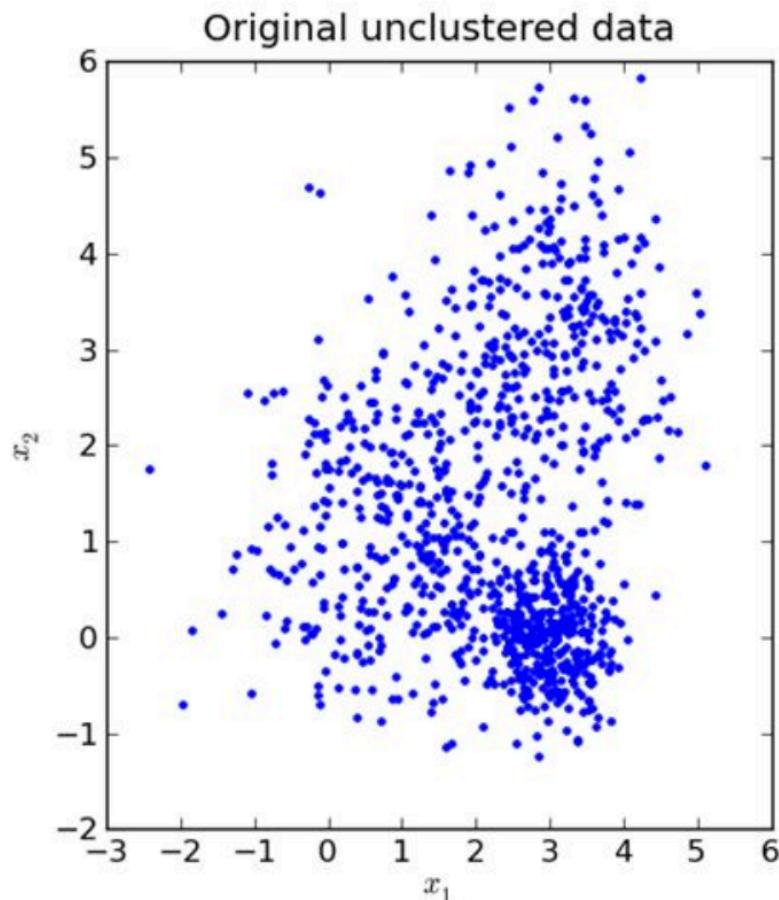
✓ Deep Learning Algorithms

- Convolutional Neural Network(CNN)
- Recurrent Neural Network(RNN)

✓ Examples

- Number Recognition
- Spam mail filtering

## ✓ Unsupervised Learning

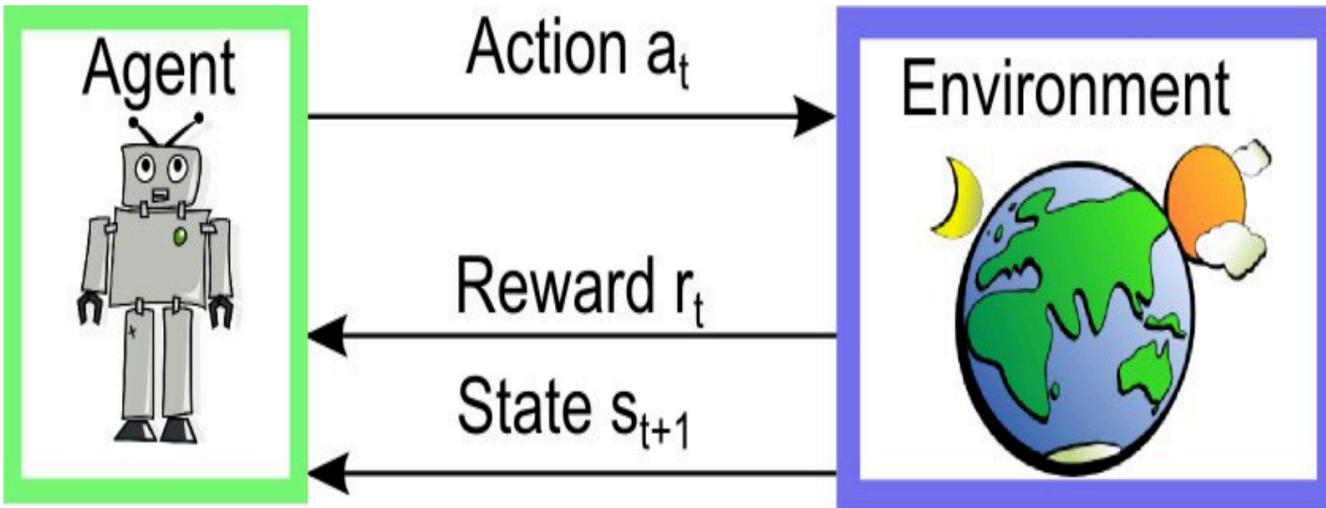


### ✓ Clustering

- ✓ Deep Learning Algorithms
  - AutoEncoders
- ✓ Examples
  - Google News Clustering
  - Word Clustering

# Machine Learning

## ✓ Reinforcement Learning



Reinforcement Learning Setup

## ✓ Deep Learning Algorithms

- Q-learning
- Deep-Q-Network(DQN)

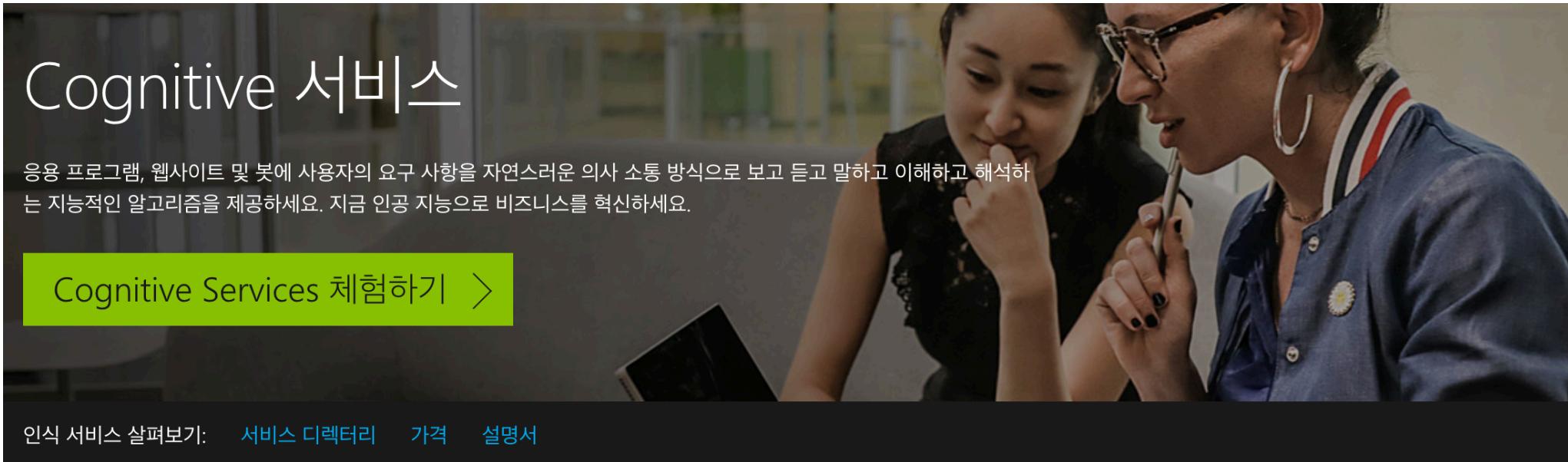


## ✓ Examples

- “AlphaGo”
- Autonomous Car driving

# Machine Learning

## ✓ Microsoft Cognitive API



Cognitive 서비스

응용 프로그램, 웹사이트 및 봇에 사용자의 요구 사항을 자연스러운 의사 소통 방식으로 보고 듣고 말하고 이해하고 해석하는 지능적인 알고리즘을 제공하세요. 지금 인공 지능으로 비즈니스를 혁신하세요.

[Cognitive Services 체험하기 >](#)

인식 서비스 살펴보기: [서비스 딕렉터리](#) [가격](#) [설명서](#)

## 인공지능을 사용하여 비즈니스 문제 해결



### 시각

이미지 처리 알고리즘을 통해 사진을 스마트하게 인식, 캡션 작성, 조정합니다.

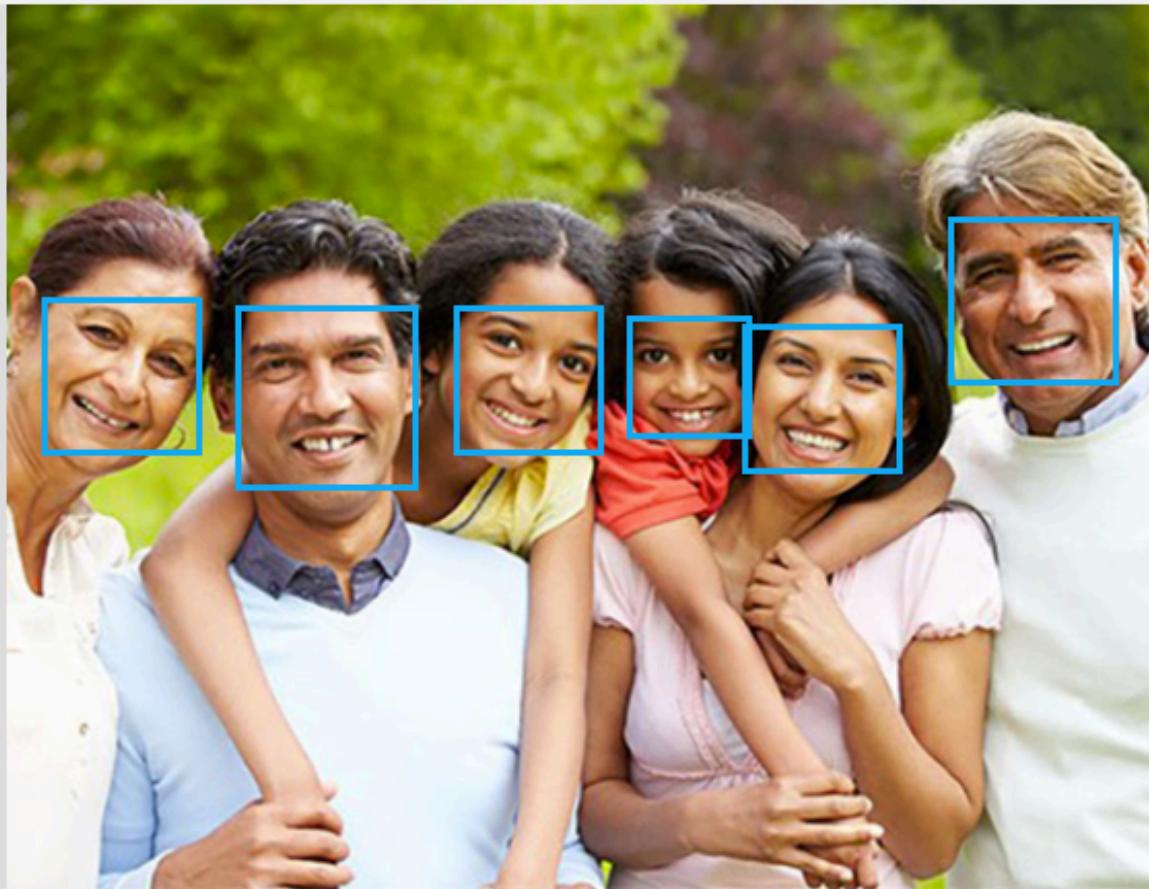


### 음성

음성을 텍스트로 변환하고 신원 확인에 음성을 사용하거나 앱에 화자 인식 기능을 추가합니다.

# Machine Learning

## ✓ Microsoft Cognitive Face API



Detection result:

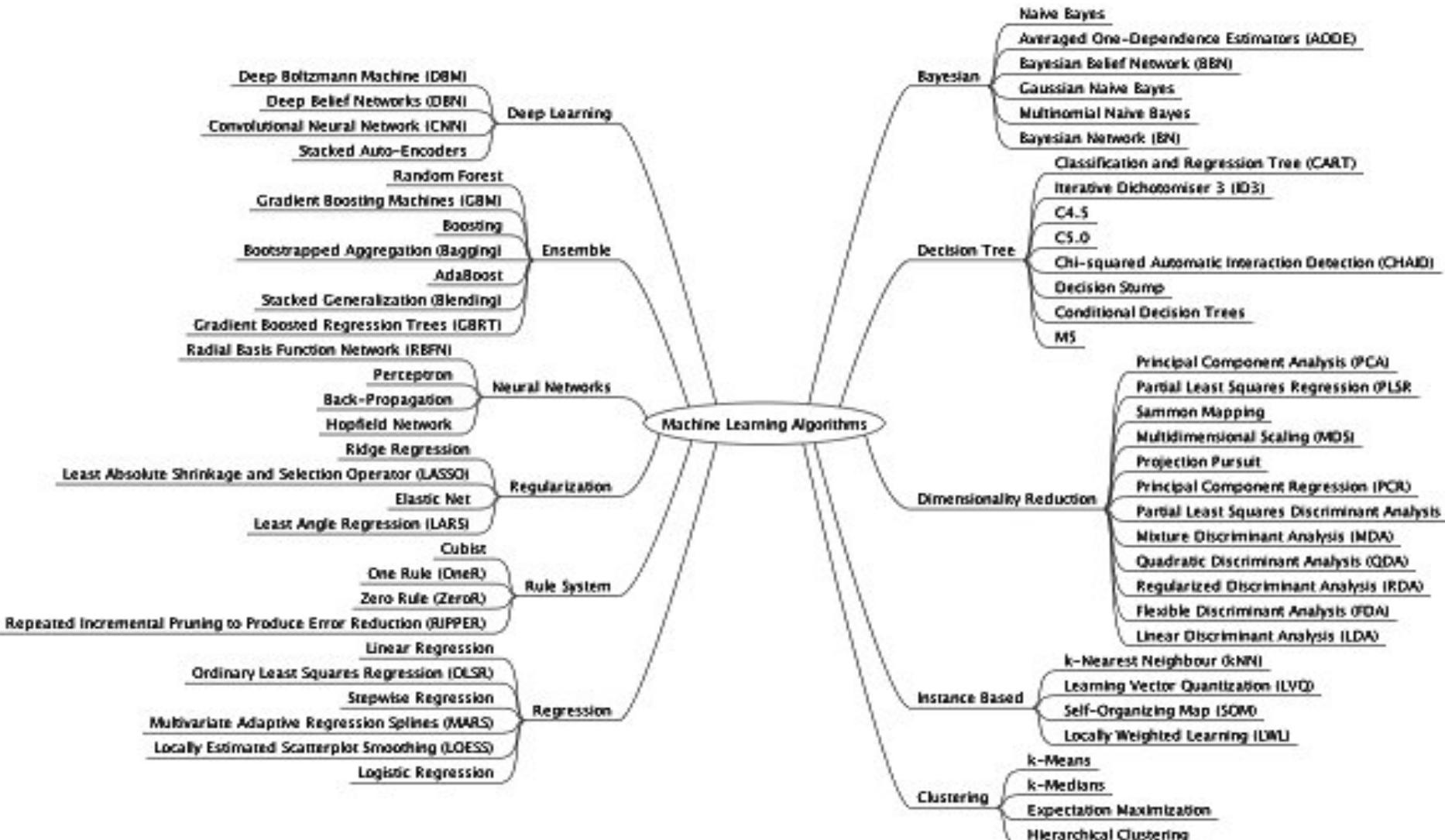
6 faces detected

JSON:

```
[  
  {  
    "faceRectangle": {  
      "top": 159,  
      "left": 117,  
      "width": 95,  
      "height": 95  
    },  
    "scores": {  
      "anger": 2.80264025E-08,  
      "contempt": 6.347948E-08,  
      "disgust": 9.236748E-06,  
      "fear": 2.52110336E-15,  
      "happiness": 0.999983549,  
      "neutral": 7.132026E-06,  
      "sadness": 2.71582626E-12,  
      "surprise": 5.42829448E-10  
    }  
  },  
  {  
    "faceRectangle": {  
      "top": 314,  
      "left": 117,  
      "width": 95,  
      "height": 95  
    },  
    "scores": {  
      "anger": 2.80264025E-08,  
      "contempt": 6.347948E-08,  
      "disgust": 9.236748E-06,  
      "fear": 2.52110336E-15,  
      "happiness": 0.999983549,  
      "neutral": 7.132026E-06,  
      "sadness": 2.71582626E-12,  
      "surprise": 5.42829448E-10  
    }  
  },  
  {  
    "faceRectangle": {  
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      "left": 117,  
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      "height": 95  
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      "contempt": 6.347948E-08,  
      "disgust": 9.236748E-06,  
      "fear": 2.52110336E-15,  
      "happiness": 0.999983549,  
      "neutral": 7.132026E-06,  
      "sadness": 2.71582626E-12,  
      "surprise": 5.42829448E-10  
    }  
  },  
  {  
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      "width": 95,  
      "height": 95  
    },  
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      "disgust": 9.236748E-06,  
      "fear": 2.52110336E-15,  
      "happiness": 0.999983549,  
      "neutral": 7.132026E-06,  
      "sadness": 2.71582626E-12,  
      "surprise": 5.42829448E-10  
    }  
  },  
  {  
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      "height": 95  
    },  
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      "disgust": 9.236748E-06,  
      "fear": 2.52110336E-15,  
      "happiness": 0.999983549,  
      "neutral": 7.132026E-06,  
      "sadness": 2.71582626E-12,  
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    }  
  },  
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      "height": 95  
    },  
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      "contempt": 6.347948E-08,  
      "disgust": 9.236748E-06,  
      "fear": 2.52110336E-15,  
      "happiness": 0.999983549,  
      "neutral": 7.132026E-06,  
      "sadness": 2.71582626E-12,  
      "surprise": 5.42829448E-10  
    }  
  }]
```

# Machine Learning

## ✓ Algorithms



- ✓ Linear Regression
- ✓ Logistic Regression
- ✓ Neural Network
- ✓ ...

# Linear Regression

## Linear Regression : What is Linear Regression?

### ✓ Linear Regression 이란?

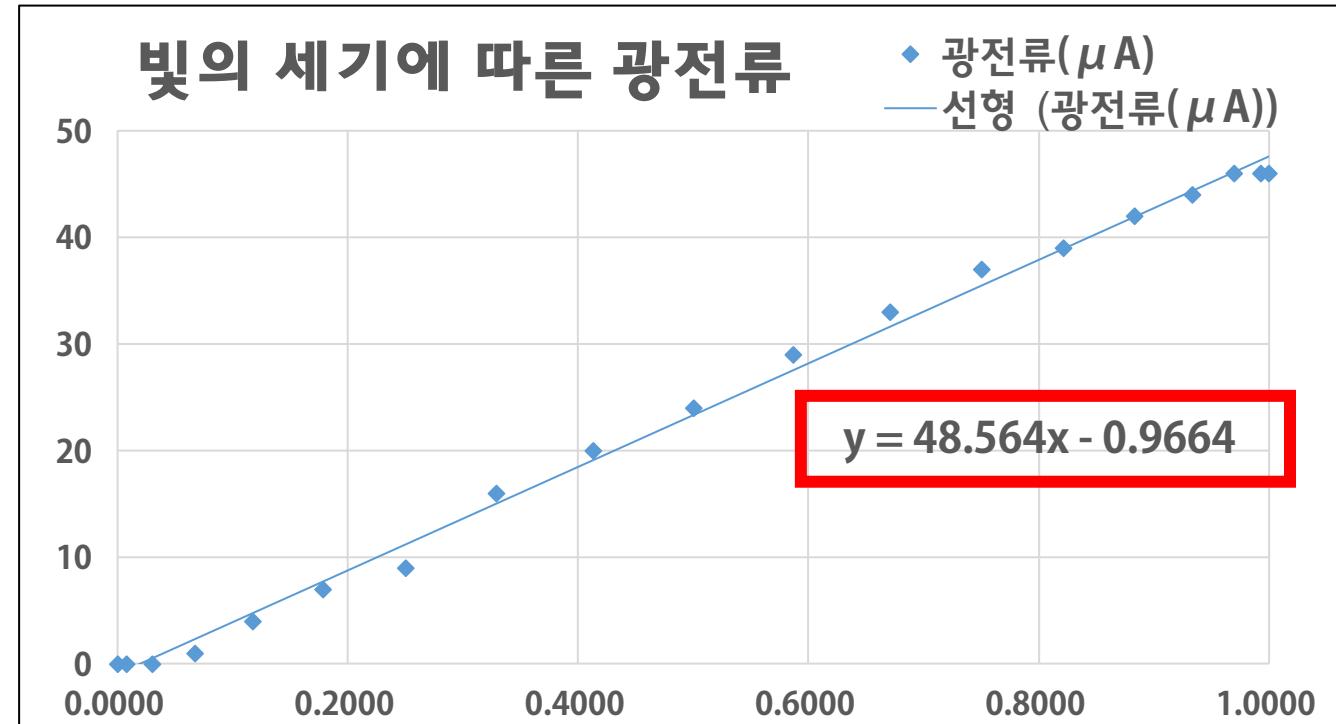
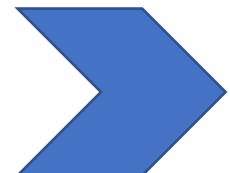
- 종속 변수( $y$ )와 독립 변수( $x$ )간의 관계가 선형적이라는 가정 하에, 이 둘의 상관 관계를 선형적으로 모델링
- 선형적(linear) : 그래프로 그렸을 때 직선 형태로 나타나는 관계 (Ex> 힘의 크기와 가속도의 크기)

### ✓ 용어

- Independent variable : input value( $x$ )
- Dependent variable : predicted output( $y$ ) for the set of input values
- Parameter(or Weight) : scale factor( $\theta$ ) for input value

# Linear Regression : What is Linear Regression?

빛의 세기(상대값)	광전류( $\mu A$ )
1.0000	46
0.9924	46
0.9698	46
0.9330	44
0.8830	42
0.8214	39
0.7500	37
0.6710	33
0.5868	29
0.5000	24
0.4132	20
0.3290	16
0.2500	9
0.1786	7
0.1170	4
0.0670	1
0.0302	0
0.0076	0
0.0000	0



- Dependent variable : 빛의 세기(상대값)
- Independent variable : 광전류( $\mu A$ )
- Weight:  $\theta_1 = 48.564, \theta_0 = -0.9664$
- model :  $y = 48.564x - 0.9664$

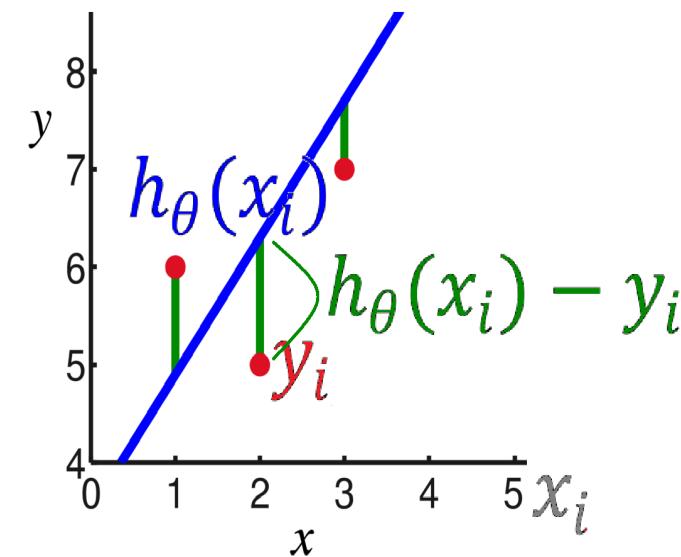
## Linear Regression : Simple Linear Regression

### ✓ Simple Linear Regression : 하나의 독립 변수

- n개의 training set이 있다고 할 때,  $h_{\theta}(x_i) = \hat{y}_i = \theta_0 + \theta_1 x_i$  ( $i = 1, 2, \dots, n$ ) 꼴의 관계임을 가정할 수 있다.

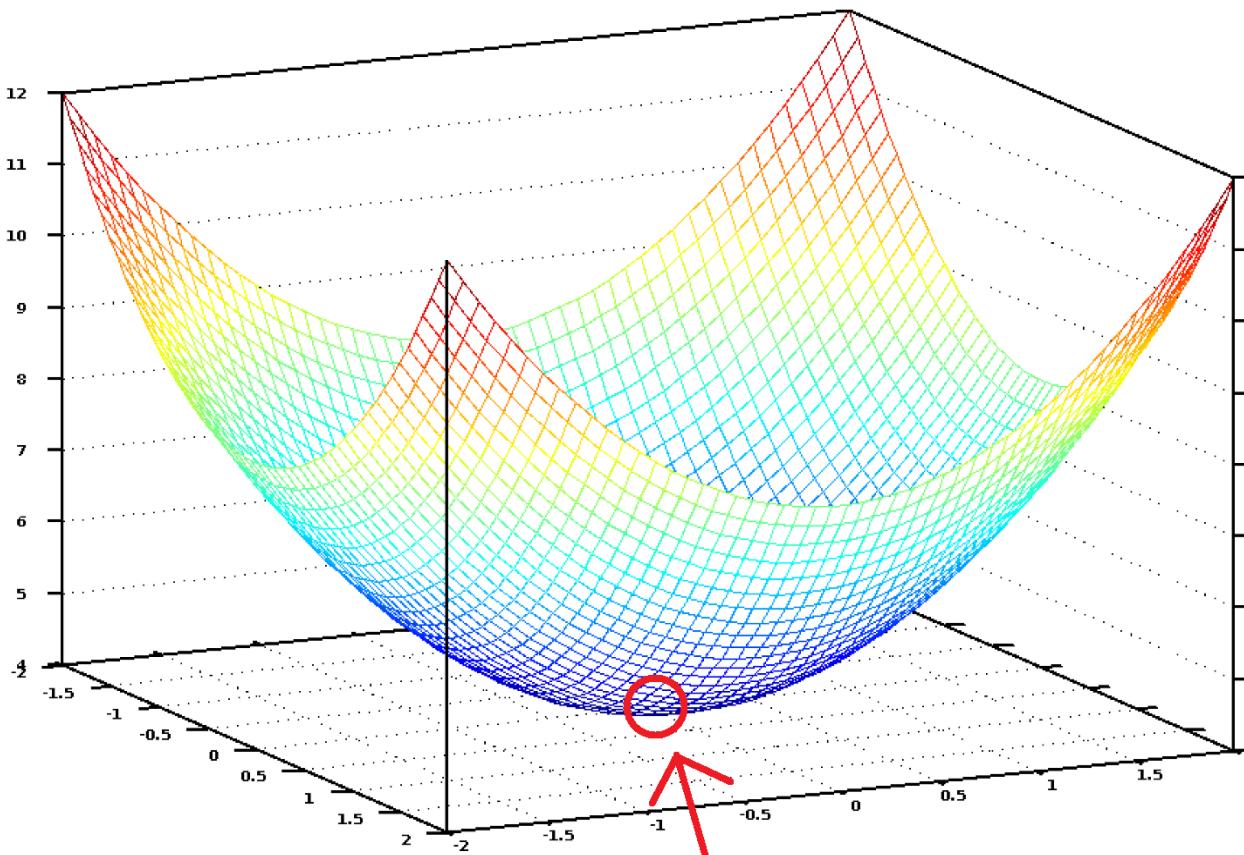
### ✓ Cost function (loss function) : 회귀식의 효율성 검증

- $\text{cost} = \frac{1}{m} \sum_{i=1}^m (\hat{y}_i - y_i)^2$   
 $= \frac{1}{m} \sum_{i=1}^m (h_{\theta}(x_i) - y_i)^2$
- ***Goal: minimize the cost function***



# Linear Regression : How to minimize the Cost Function

✓ Graph of Cost Function :  $z = J(\theta_0, \theta_1)$



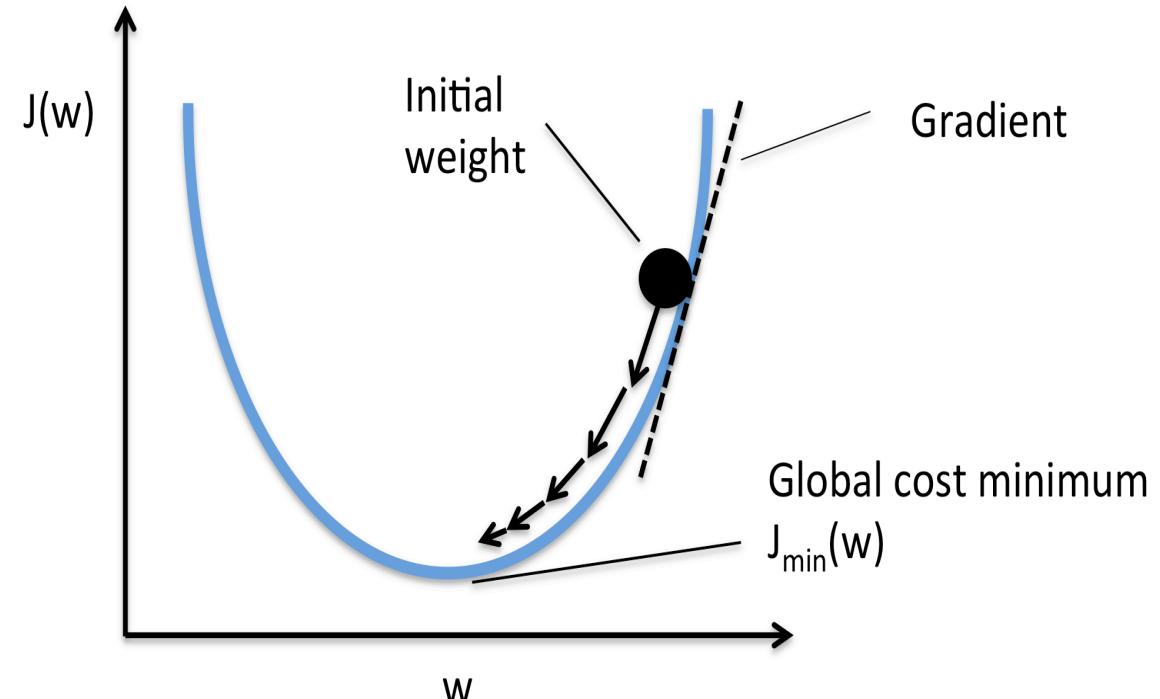
*cost function이 최소화되는 곳  
= 가장 효율적인 회귀식*

# Linear Regression : How to minimize the Cost Function

## ✓ Gradient Descent

$$\theta_i := \theta_i - \alpha \frac{\partial}{\partial \theta_i} J(\theta_0, \theta_1) \text{ for } i = 0, i = 1$$

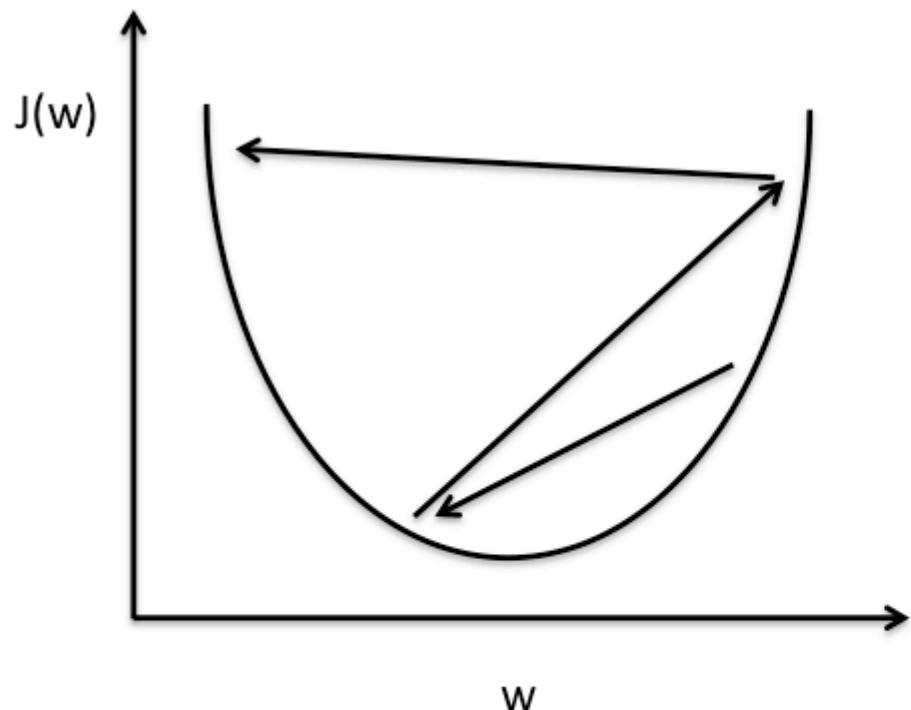
- $\alpha (> 0)$  “learning rate”  
: local minimum을 향해 가는 속도



- $\frac{\partial}{\partial \theta_0} J(\theta_0, \theta_1) = \frac{\partial}{\partial \theta_0} \left[ \frac{1}{2m} \sum_{i=1}^m (h_\theta(x_i) - y_i)^2 \right] = \frac{1}{m} \sum_{i=1}^m (h_\theta(x_i) - y_i)$
- $\frac{\partial}{\partial \theta_1} J(\theta_0, \theta_1) = \frac{\partial}{\partial \theta_1} \left[ \frac{1}{2m} \sum_{i=1}^m (h_\theta(x_i) - y_i)^2 \right] = \frac{1}{m} \sum_{i=1}^m (h_\theta(x_i) - y_i) x_i$   
: 이동할 방향과 크기

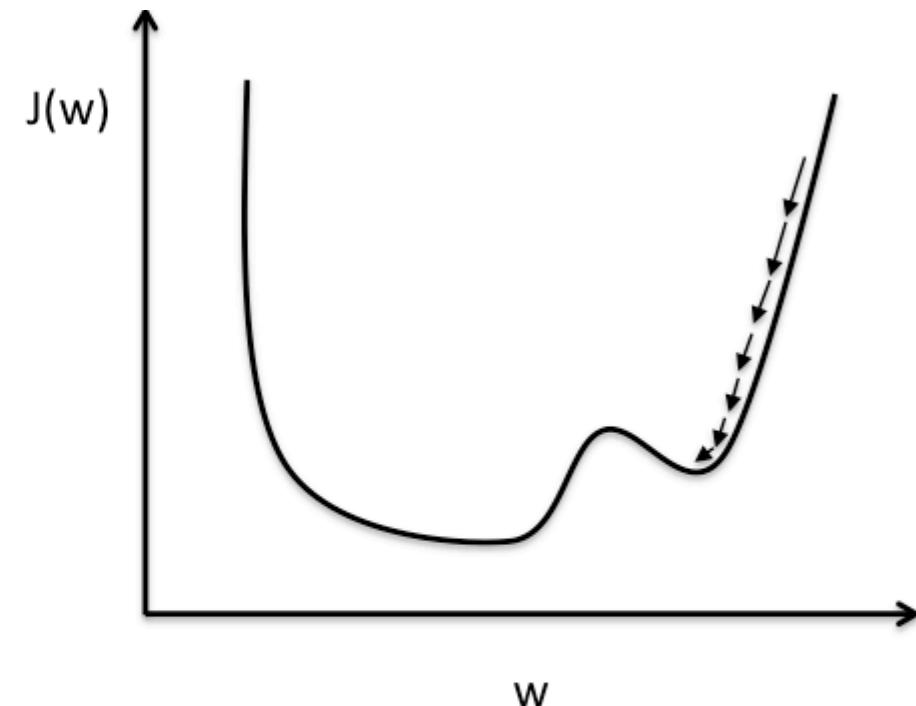
# Linear Regression : How to minimize the Cost Function

## When Learning Rate is Too Big



Large learning rate: Overshooting.

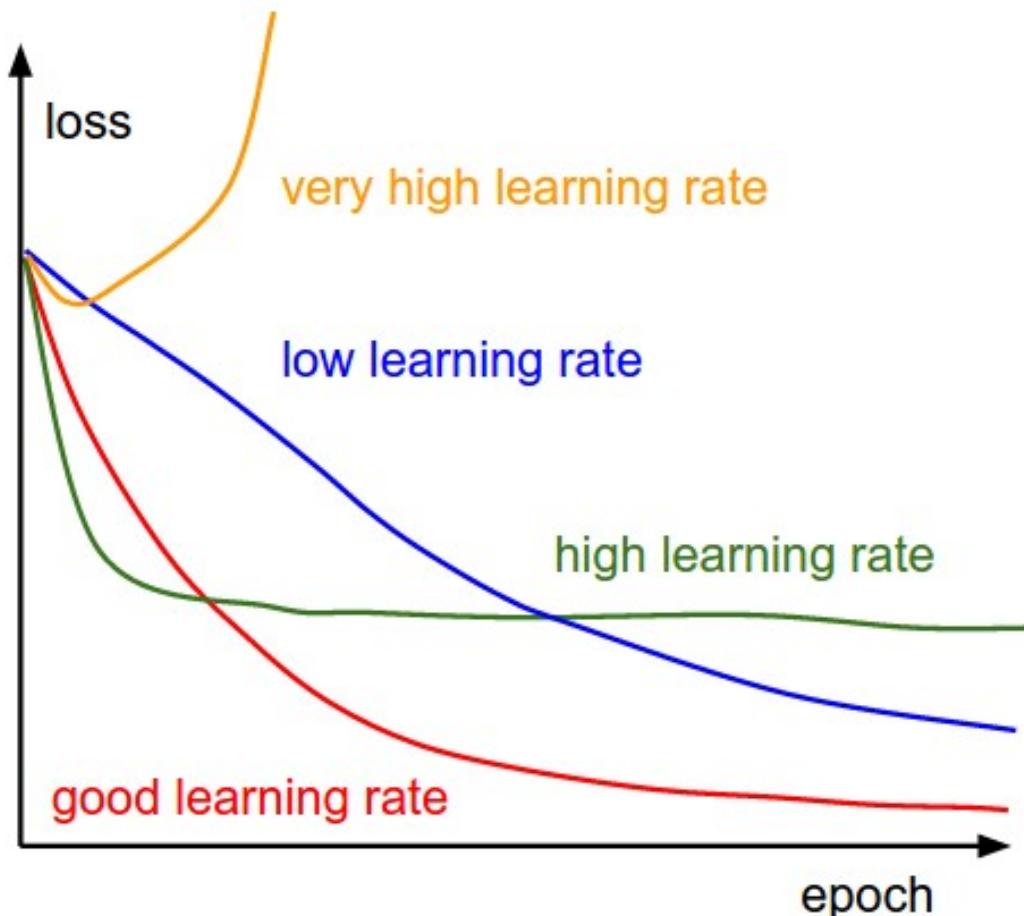
## When Learning Rate is Too Small



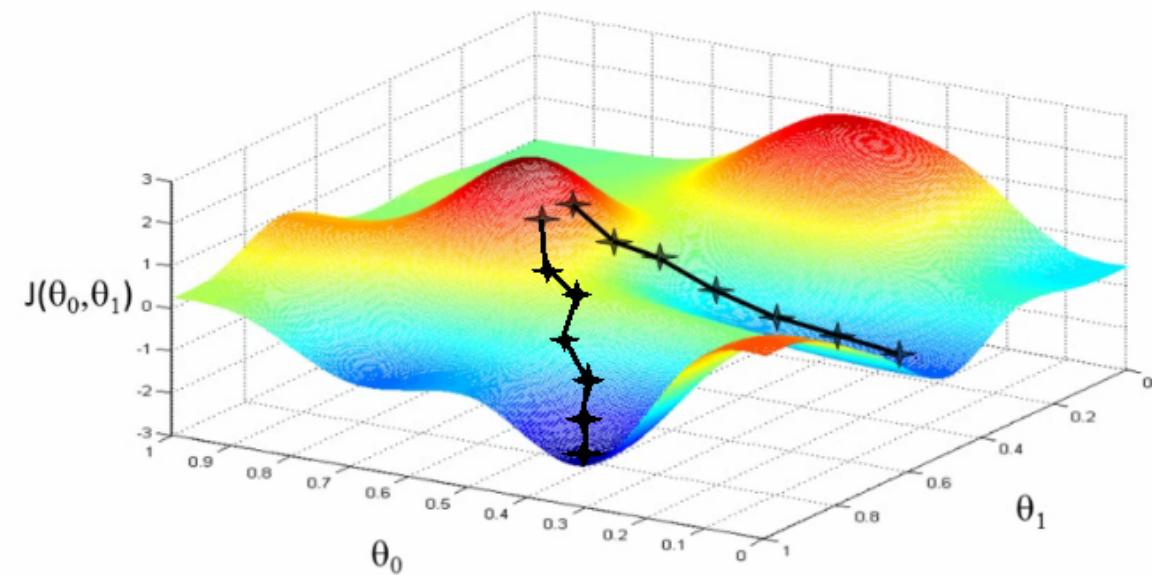
Small learning rate: Many iterations until convergence and trapping in local minima.

# Linear Regression : How to minimize the Cost Function

## Learning Rate



## Local minimum vs Global minimum



- ✓ Linear regression의 cost function은 Convex function!
- Local min. = Global min. = optimal value

## Linear Regression : Multiple Linear Regression

✓ Multiple Linear Regression : 여러 개의 독립 변수

✓ Model

$$h_{\theta}(x) = \hat{y}_i = \theta_0 x_{i0} + \theta_1 x_{i1} + \cdots + \theta_p x_{ip}, \forall x_{i0} = 1 (i = 1, 2, \dots, n)$$

✓ Training set

$$\text{Let } \mathbb{Y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}, \widehat{\mathbb{Y}} = \begin{bmatrix} h_{\theta}(x_1) \\ h_{\theta}(x_2) \\ \vdots \\ h_{\theta}(x_n) \end{bmatrix}, \mathbb{X} = \begin{bmatrix} x_1^T \\ x_2^T \\ \vdots \\ x_n^T \end{bmatrix} = \begin{bmatrix} 1 & x_{11} & \cdots & x_{1p} \\ 1 & x_{21} & \cdots & x_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n1} & \cdots & x_{np} \end{bmatrix}, \theta = \begin{bmatrix} \theta_0 \\ \theta_1 \\ \theta_2 \\ \vdots \\ \theta_p \end{bmatrix}$$

$$\text{Then } h_{\theta}(\mathbb{X}) = \widehat{\mathbb{Y}} = \mathbb{X}\theta$$

✓ Cost Function :  $J(\theta) = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x_i) - y_i)^2$

✓ Gradient Descent :  $\theta := \theta - \alpha \frac{1}{m} \mathbb{X}^T (\mathbb{X}\theta - \mathbb{Y})$

# Logistic Regression

## Logistic Regression : What is logistic regression?

### ✓ Logistic Regression이란?

- 선형 회귀는 종속변수가 다양. 로지스틱 회귀는 종속 변수가 범주형 (이항형 문제)
- 데이터를 통해 Binary로 1 또는 0으로 분류 하는 학습 → binary Classification

### ✓ 용어

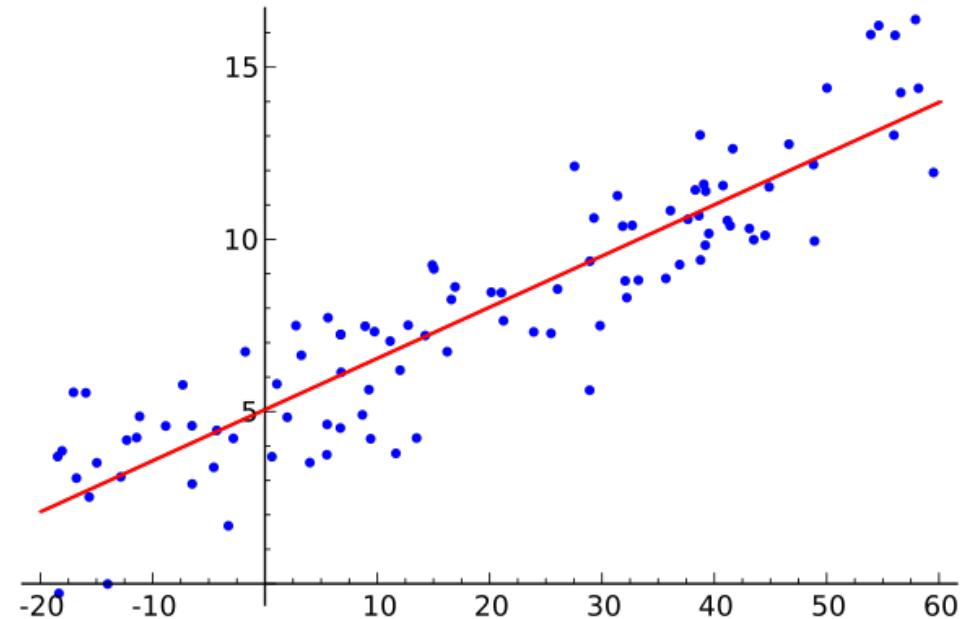
- 독립변수: 입력값이나 원인 (예측변수, 복귀자, 통제변수)
- 종속변수: 결과물이나 효과 (설명변수)

# Logistic Regression: What is logistic regression?

## Linear Regression

### ✓ Linear Regression

Output: multiple



### ✓ Logistic Regression

- Output: 0 or 1  
(binary classification)
- Output: 0 or 1 or 2  
(multi-class classification)

Example of [simple linear regression](#), which has one independent variable

## Logistic regression : What is logistic regression? (binary classification)

- ✓ 독립 변수가  $[-\infty, \infty]$ 의 어느 숫자이든 상관 없이 종속 변수 즉 결과 값이 항상 범위 [0,1] 사이에 존재

X is  $\infty$

Y is 0 or 1

- ✓ **Example**

- **Spam Detection:** spam(1) or Ham(0)
- **Facebook feed:** show(1) or hide(0)
- 부상을 입은 환자들의 사망 예측을 위해 사용되는 **Trauma and Injury Severity Score (TRISS)**
- 이미지를 분석하여 lion dog cat 분류

The image shows a software application window titled "TRISS". It is divided into three main sections: "ISS CALCULATOR", "RTS CALCULATOR", and "TRISS".

- ISS CALCULATOR:** Contains fields for AIS Scores: Head, Face, Chest, Abdomen, Extremity, and External. Below these is a "Calculate" button and an "ISS:" output field.
- RTS CALCULATOR:** Contains fields for Systolic BP, Resp. Rate, and Coma Score. Below these is a "Calculate" button and an "RTS:" output field.
- TRISS:** Contains a "Age" input field, a "Calculate" button, and a "Probability of Survival" section with fields for Blunt and Penetrating trauma. Below these is a "Clear" button.

The [TRISS Calculator](#) will open as a standalone window on your desktop.

**(Injury Severity Score - Revised Trauma Score  
- Trauma Injury Severity Score)**

# Logistic regression : Sigmoid

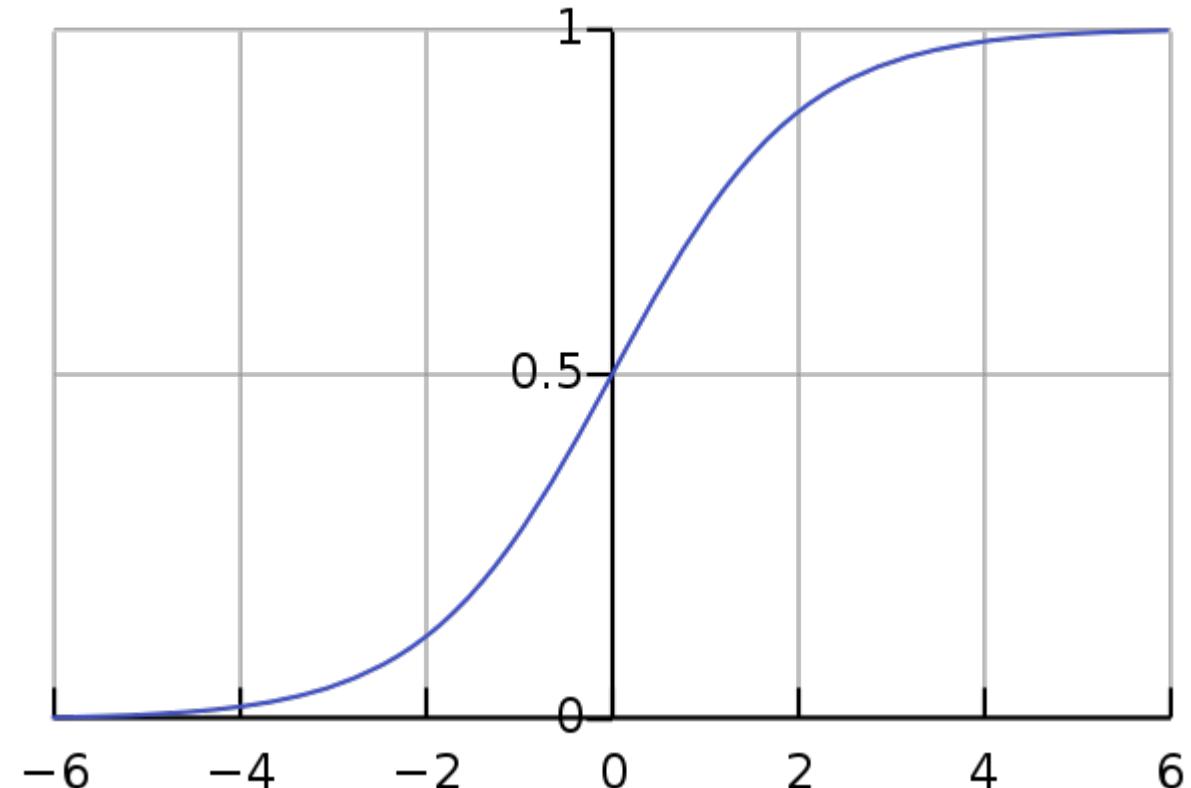
X is  $\infty$   
Y is 0 or 1

**sigmoid:**

Curved in two directions, like the letter "S", or the Greek  $\varsigma$  (sigma).

→ 모든 독립변수에 대해  
종속변수는 0과 1사이의 값이 출력

## Logistic Regression Hypothesis



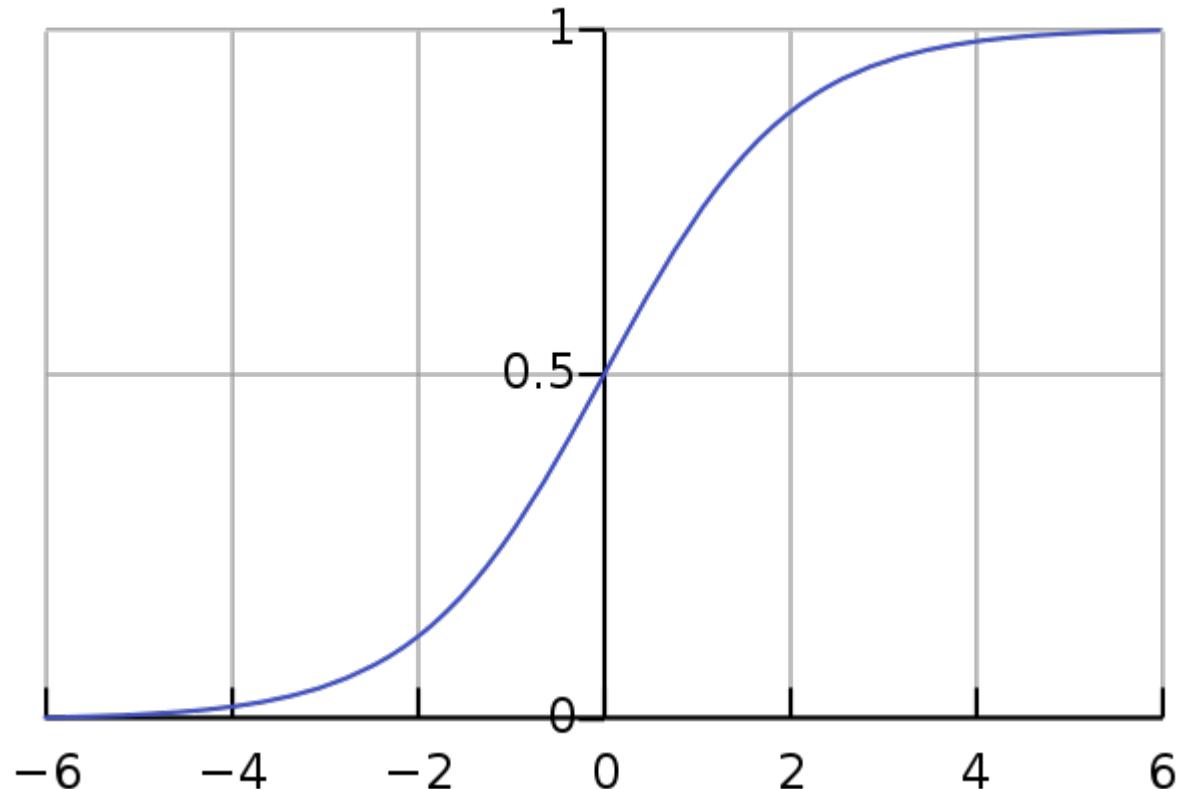
## Logistic regression: Logistic Hypothesis

- ✓ **Logistic Hypothesis(Sigmoid)**

$$H(X) = \frac{1}{1 + e^{-W^T X}}$$

- ✓ Suppose predict if  $H(x) \geq 0.5$  "y = 1"  
predict if  $H(x) < 0.5$  "y = 0"

Logistic Regression Hypothesis

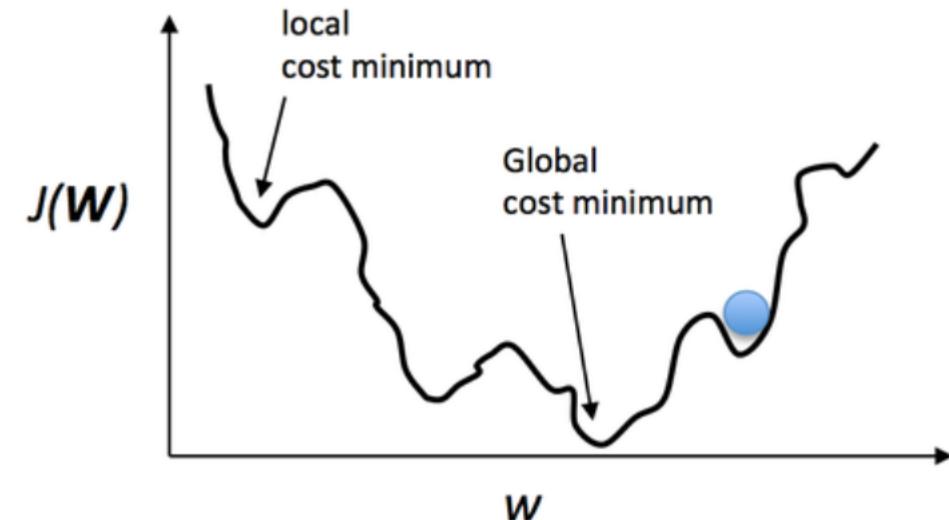


# Logistic regression: Cost function

## ✓ Cost Function (linear Regression)

$$cost(W, b) = \frac{1}{m} \sum_{i=1}^m (H(x^{(i)}) - y^{(i)})^2$$

→ Linear regression의 cost function을 이용하면, 볼록한 형태가 아니라 유통불통한 형태가 나오기 때문에 Gradient descent algorithm 을 통하여 최솟값을 구할 수 없게 된다.



Local cost minimum != Global cost minimum  
Not Convex function!

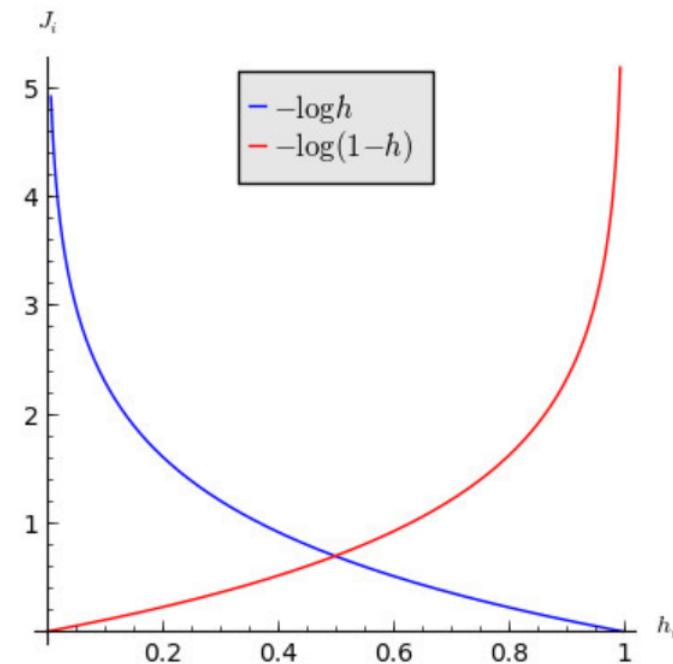
# Logistic regression : Cost function

## ✓ Cost Function (Logistic Regression)

$$cost(W) = \frac{1}{m} \sum c(H(x), y)$$

$$c(H(x), y) = \begin{cases} -\log(H(x)) & : y = 1 \\ -\log(1 - H(x)) & : y = 0 \end{cases}$$

$$C(H(x), y) = y\log(H(x)) - (1 - y)\log(1 - H(x))$$



→ log 값을 사용하게 되어 실제 데이터 결과 값인 y가 1일 때는

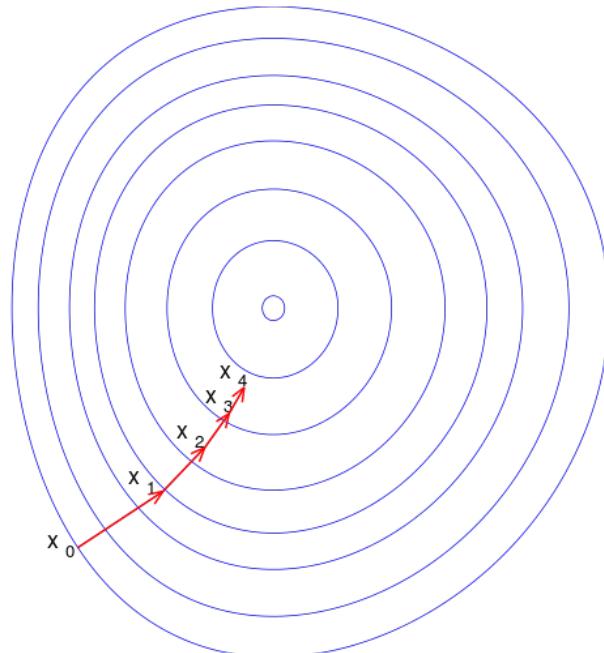
- 예측값( $x$ )이 1에 가깝다면 (예측값 == 실제 값) cost값은 0으로 수렴(cost값이 최저),
- 예측값( $x$ )이 0에 가깝다면 (예측값 != 실제 값) cost값은 cost값이 커짐.

Y가 0일때는 반대로 성립.

## Logistic regression : Minimize cost - Gradient descent

- ✓ Gradient descent algorithm  
 (= linear Regression)

$$W := W - \alpha \frac{\partial}{\partial W} \text{cost}(W)$$

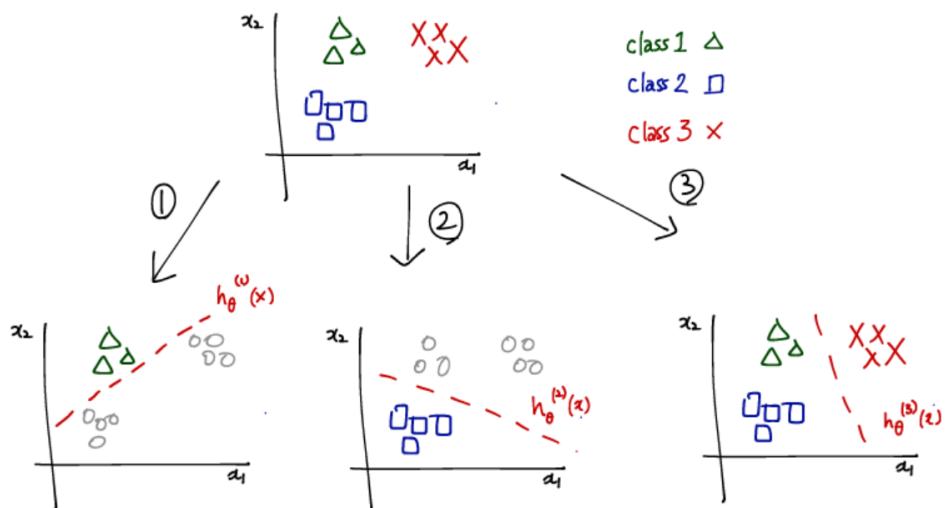


# Logistic regression : Multi-class classification

## ✓ Multi-class Classification

### ✓ Example

- Azure Cognitive service 를 이용한 multi-class classification
- Cat(1) Dog(2) Lion(3) 3가지의 범주로 분류



$$h_{\theta}^{(i)}(x) = P(y = i|x; \theta), \quad i \in \{1, 2, 3\}$$

출처: <https://wikidocs.net/4291#one-vs-all-one-vs-rest>

# Logistic regression : Multi-class classification

## ✓ Multi-class Classification

### ✓ Example

- Azure Cognitive service 를 이용한 multi-class classification
- Cat(1) Dog(2) Lion(3) 3가지의 범주로 분류

⟨Training set⟩

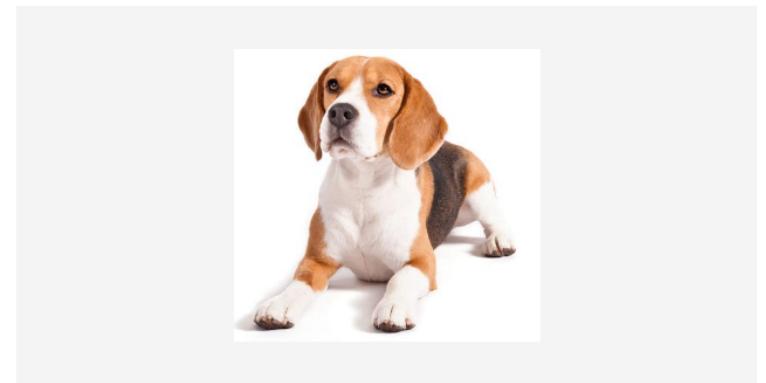


lion

dog

cat

⟨Result⟩



Results

Tag	Probability
dog	100%
cat	0%
lion	0%

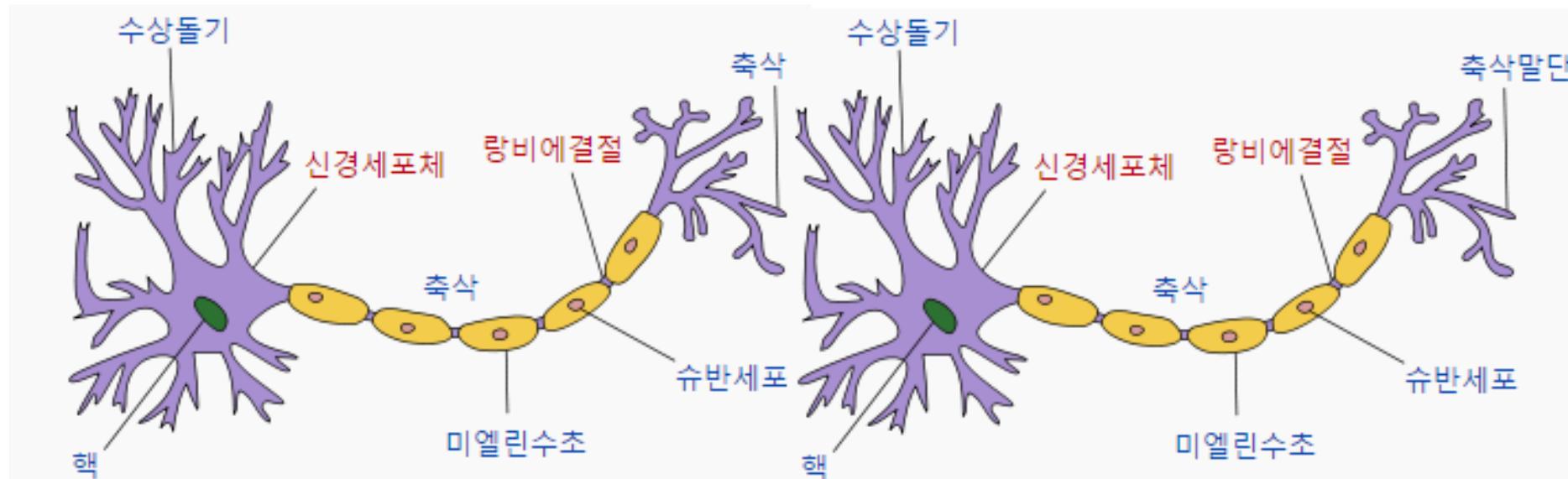
# Neural Network

# Neural Network : What is Neural Network?

## ✓ Neural Network 란?

: 생물학의 신경망에서 영감을 얻은 알고리즘  
-> 인공신경망

## ✓ 뉴런의 구조



# Neural Network : What is Neural Network?

✓ 뉴런 -> 퍼셉트론



✓ 퍼셉트론(perceptron)

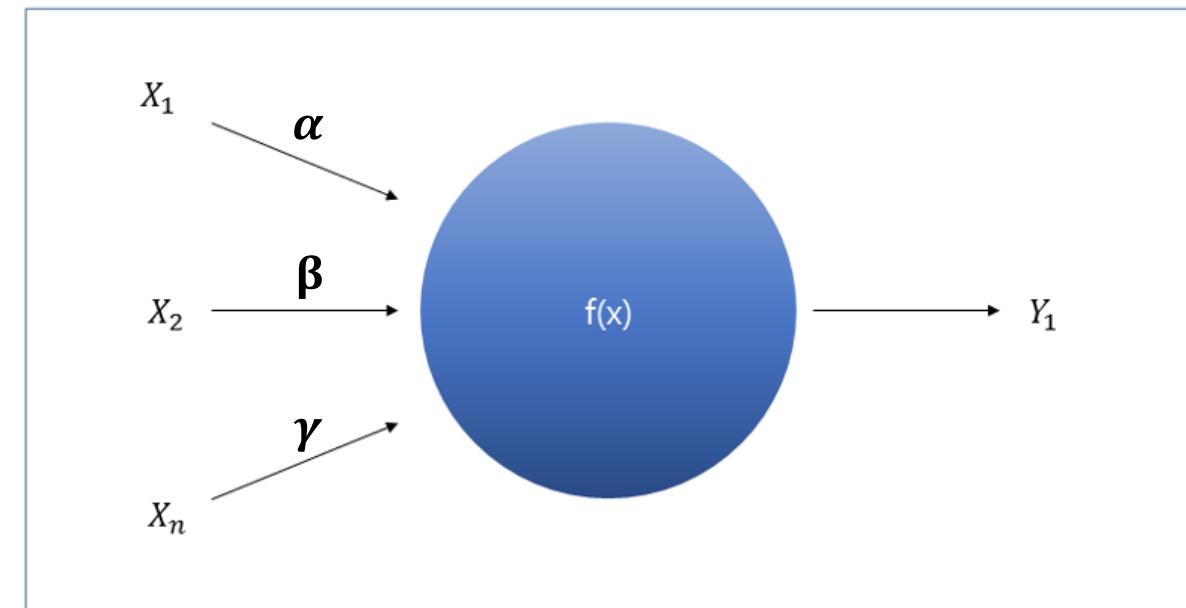
: 인간 신경 세포 하나를 흉내 낸 것

- 하나의 퍼셉트론에 대해 학습 가능

$y = \text{목표치}$

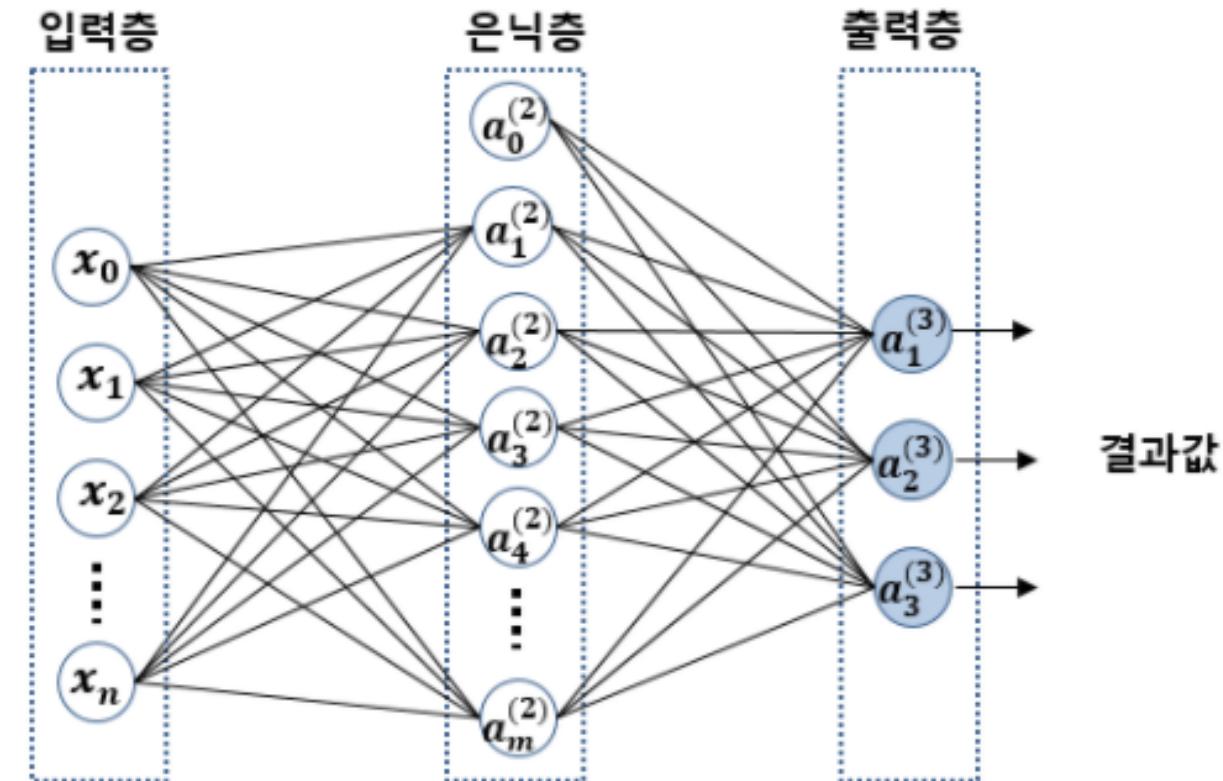
$$x_1 \times \alpha + x_2 \times \beta + x_3 \times \gamma = y_0$$

$y$ 와  $y_0$ 를 비교 -> 가중치  $\alpha, \beta, \gamma$  조정



# Neural Network : What is Neural Network?

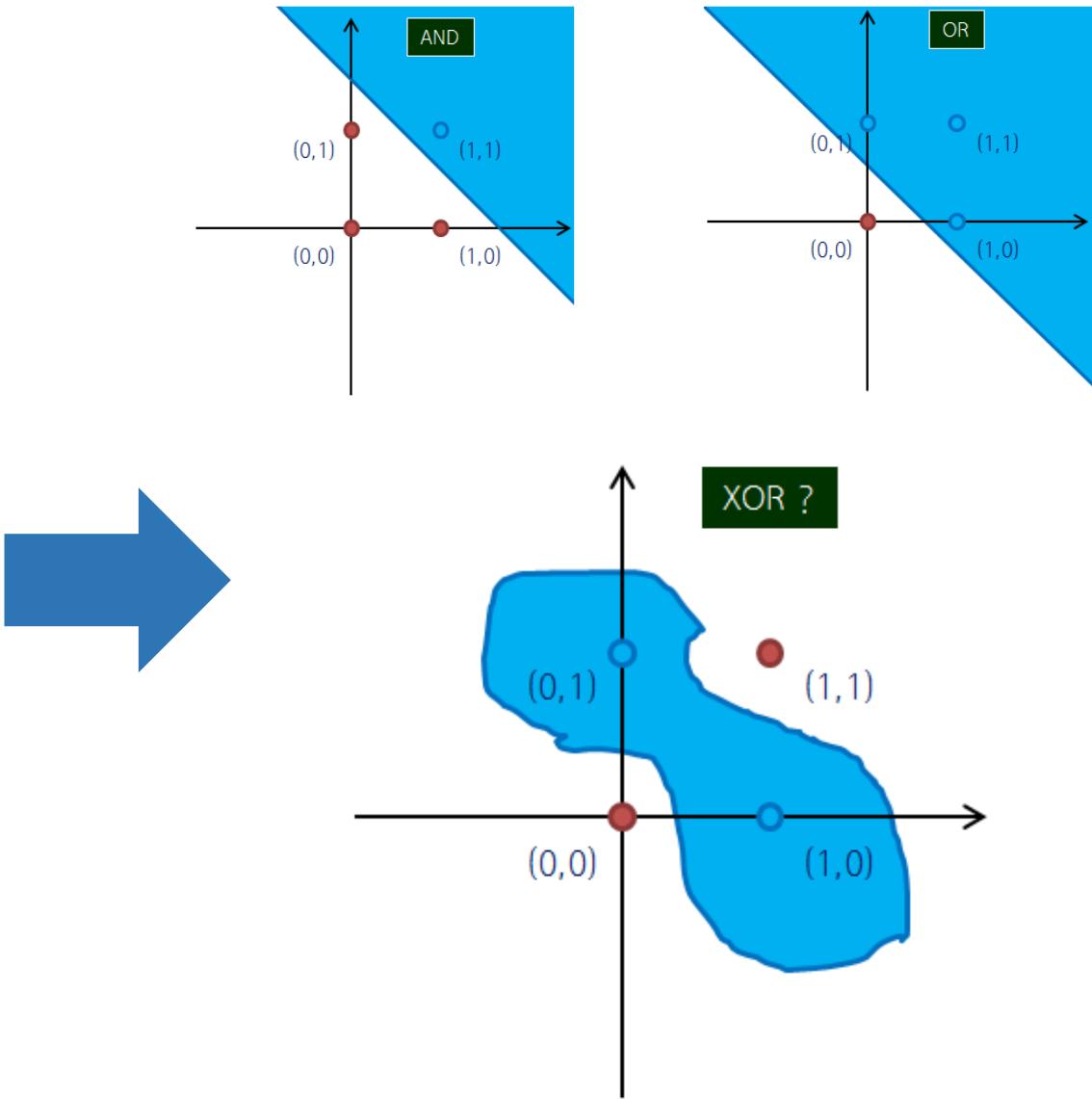
✓ Neural Network  
: 하나의 퍼셉트론을 병렬로 배치하고  
이 하나의 layer를 여러 개로 배치한  
다층 퍼셉트론



# Neural Network : XOR Example

✓ XOR example

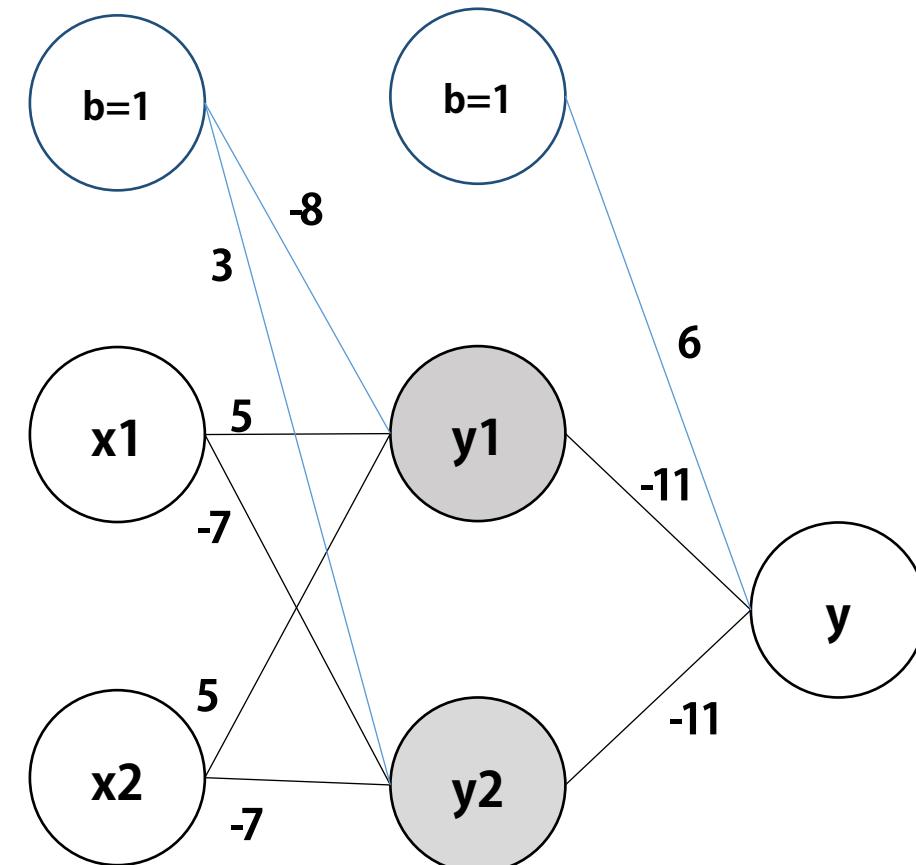
A	B	A <b>XOR</b> B
0	0	0
0	1	1
1	0	1
1	1	0



# Neural Network : XOR Example

## ✓ XOR example

x1	x2	y1	y2	y
0	0	0	1	0
0	1	0	0	1
1	0	0	0	1
1	1	1	0	0



$$y_1: 0*5 + 0*(-7) - 8 = \text{sigmoid}(-8) = 0$$

$$y_2: 0*5 + 0*(-7) + 3 = \text{sigmoid}(3) = 1$$

$$y: 0*-11 + 1*-11 + 6 = \text{sigmoid}(-5) = 0$$

감사합니다