

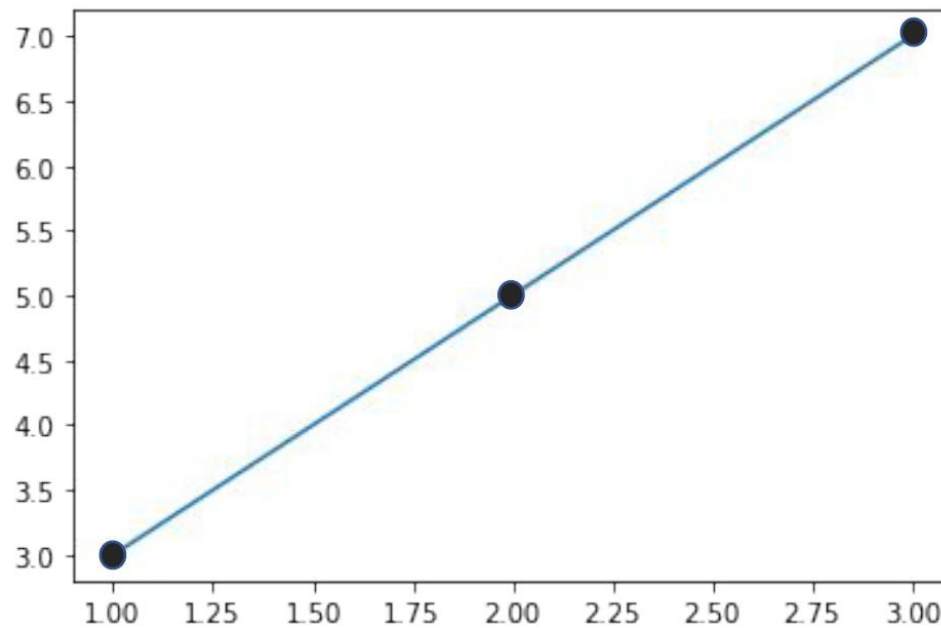
# 10강. 가설함수, 비용, 손실함수

# 가설함수

X	Y
1	3
2	5
3	7

# 가설함수

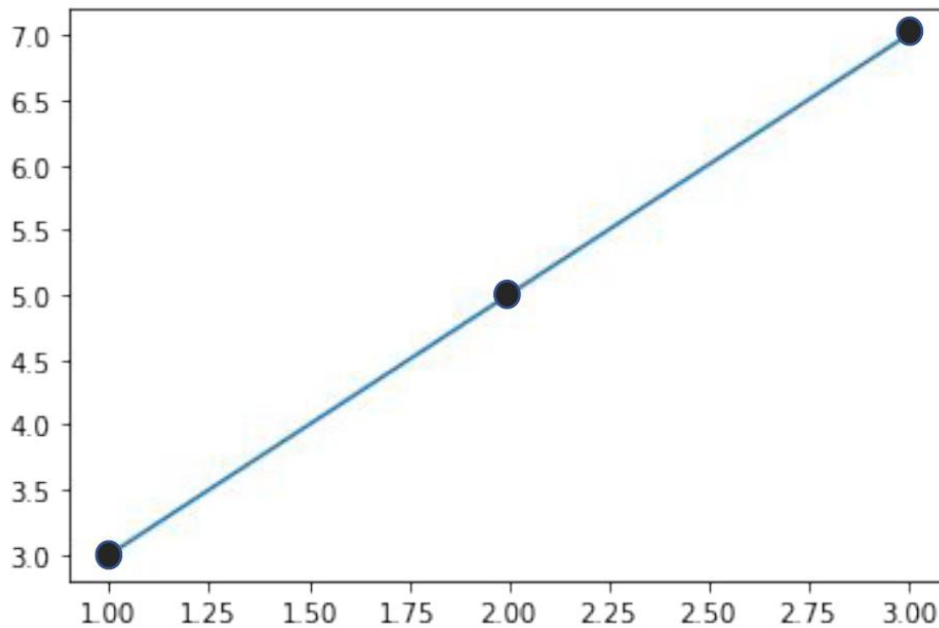
X	Y
1	3
2	5
3	7



# 가설함수

X	Y
1	3
2	5
3	7

$$H(x) = W * X + b$$



# 가설함수

X	Trial	Y Predict	Y
1	w=0.5, b=0.5	1	3
2		1.5	5
3		2	7
1	w=1, b=1	2	3
2		3	5
3		4	7
1	w=2, b=1	3	3
2		5	5
3		7	7

$$H(x) = W * X + b$$

$$H(x) = Y \text{ Predict}$$

X	Trial	Y Predict	Y
1	w=0.5, b=0.5	1	3
2		1.5	5
3		2	7
1	w=1, b=1	2	3
2		3	5
3		4	7
1	w=2, b=1	3	3
2		5	5
3		7	7

$$H(x) = W * X + b$$

$$H(x) = Y \text{ Predict}$$

$$\text{손실} = Y \text{ Predict} - Y$$

# 손실함수

X	Trial	Y Predict	Y
1	w=0.5, b=0.5	1	3
2		1.5	5
3		2	7
1	w=1, b=1	2	3
2		3	5
3		4	7
1	w=2, b=1	3	3
2		5	5
3		7	7

$$H(x) = W * X + b$$

$$H(x) = Y \text{ Predict}$$

$$\text{손실} = Y \text{ Predict} - Y$$

$$\text{손실함수} = W * X + b - Y$$

## 손실의 총합

$$\text{손실함수} = W * X + b - Y$$

Y Predict	Y	Loss
1	3	-2
1.5	5	-3.5
2	7	-5
2	3	-1
3	5	-2
4	7	-3
3	3	0
5	5	0
7	7	0

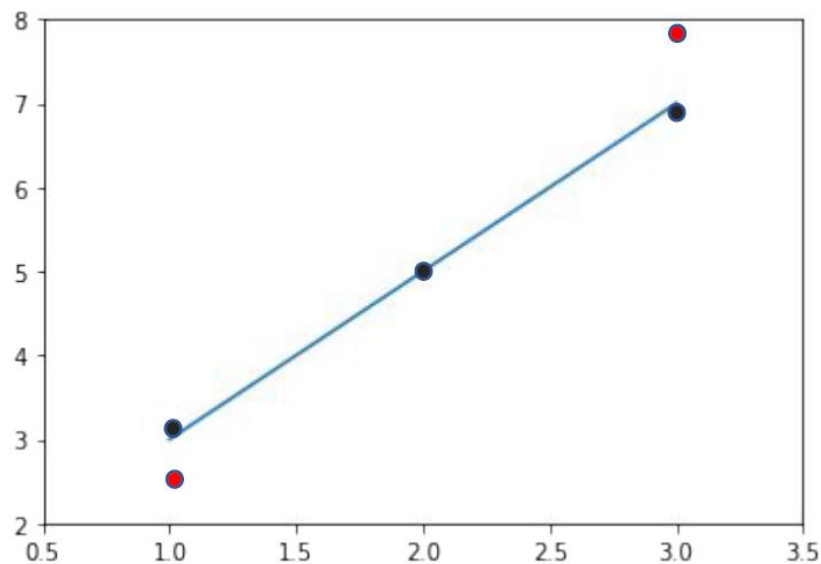
Trial	Y
w=0.5, b=0.5	-2 + -3.5 + -5 = -10.5
w=1, b=1	-5 + -1 + -2 = -8
w=2, b=1	0 + 0 + 0 = 0



## 손실의 총합의 오류?

$$\text{손실함수} = W * X + b - Y = 2X + 1 - Y$$

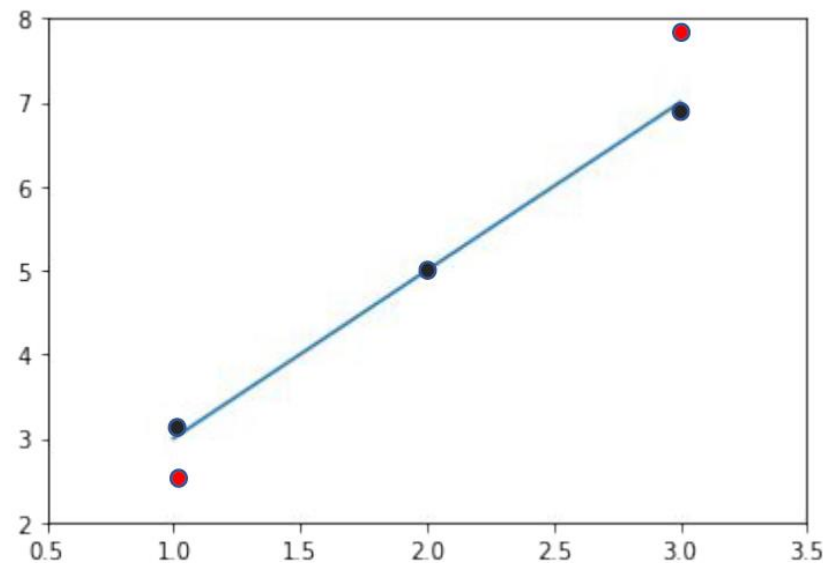
X	Y	Loss	X2	Y2	Loss
1	3	0	1	2	-1
2	5	0	2	5	0
3	7	0	3	8	1



## 손실의 총합의 제공 오차 (Mean Squared Error)

$$\text{손실함수} = W * X + b - Y = (2X + 1 - Y)^2$$

X	Y	Loss	X2	Y2	Loss
1	3	0	1	2	-1
2	5	0	2	5	0
3	7	0	3	8	1



## 제곱 오차 (Mean Squared Error)

---

$$\text{손실함수} = W * X + b - Y$$

$$= (W * X + b - Y)^2 \text{의 전체 합}$$

$$= \sum (W * X + b - Y)^2$$

## 제곱 오차 (Mean Squared Error)

$$\begin{aligned}\text{손실함수} &= W * X + b - Y \\ &= (W * X + b - Y)^2 \text{의 전체 합} \\ &= \sum (W * X + b - Y)^2\end{aligned}$$

데이터의 개수가 많아지면 손실이 커지므로  
전체 손실의 평균을 구합니다

$$= \sum (W * X + b - Y)^2 / N (\text{데이터 개수})$$

## 제곱 오차 (Mean Squared Error)

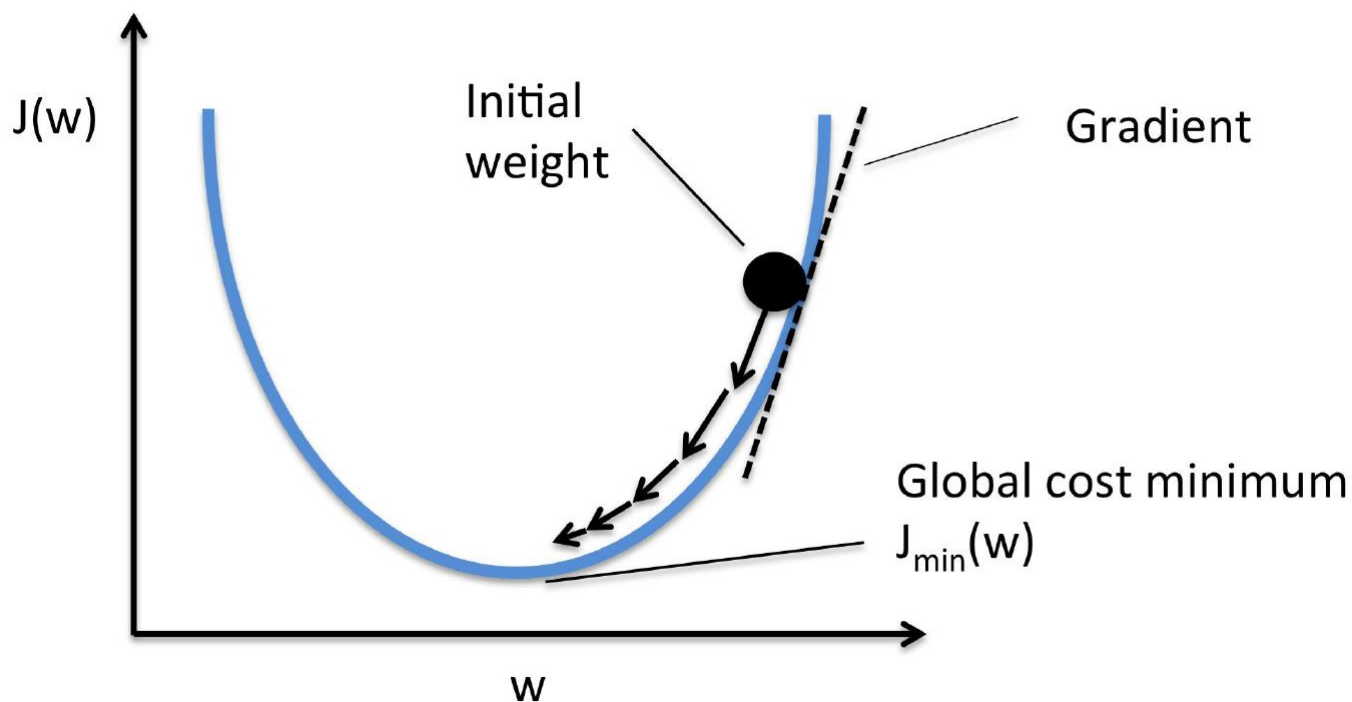
데이터의 개수가 많아지면 손실이 커지므로  
전체 손실의 평균을 구합니다

$$= \sum (W * X + b - Y)^2 / N (\text{데이터 개수})$$

$$MSE = \frac{1}{n} \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$$

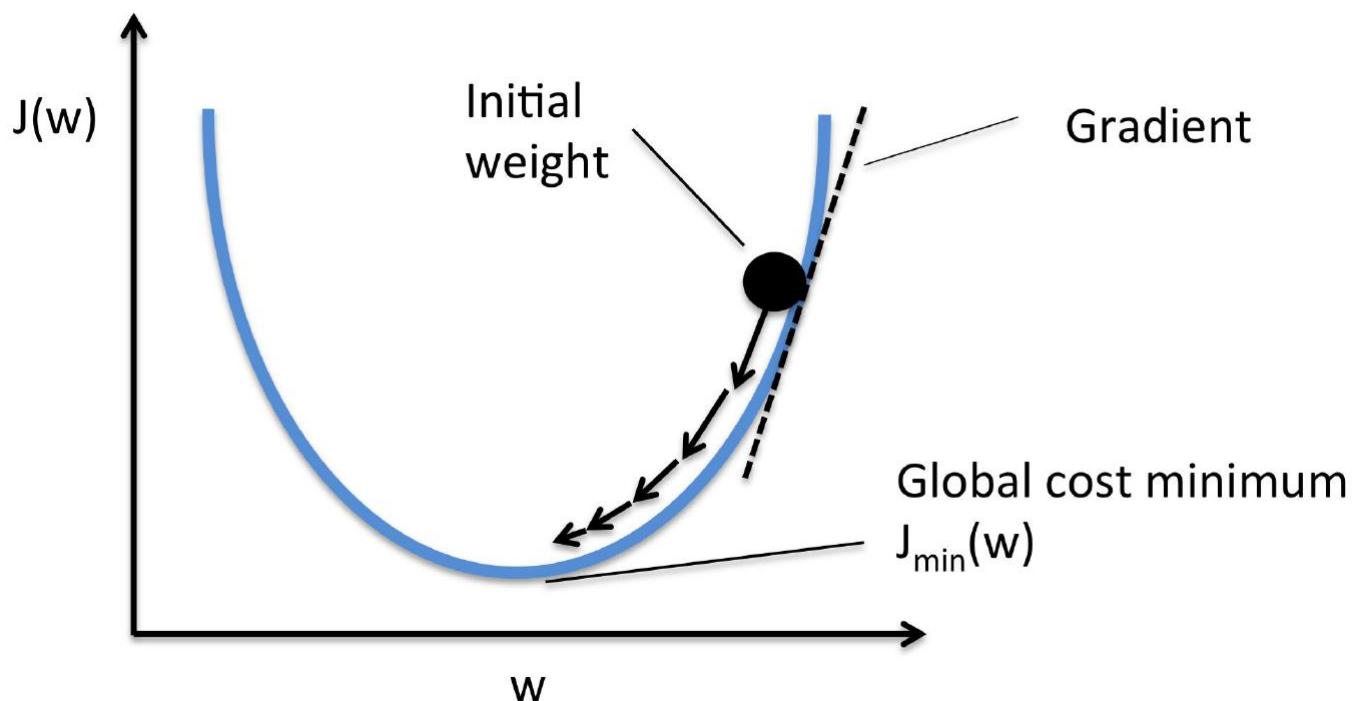
## 최적의 W를 찾는 것 = 최소 오차

$$MSE = \frac{1}{n} \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$$



# 경사하강법 (Gradient Descent)

$$MSE = \frac{1}{n} \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$$



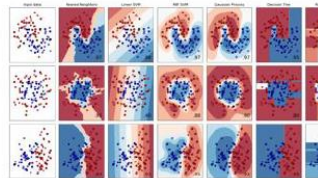


## Classification

Identifying which category an object belongs to.

**Applications:** Spam detection, image recognition.

**Algorithms:** SVM, nearest neighbors, random forest, and more...



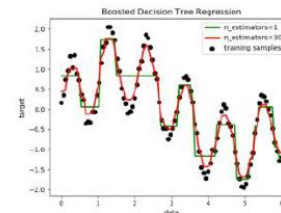
Examples

## Regression

Predicting a continuous-valued attribute associated with an object.

**Applications:** Drug response, Stock prices.

**Algorithms:** SVR, nearest neighbors, random forest, and more...



Examples

## Clustering

Automatic grouping of similar objects into sets.

**Applications:** Customer segmentation, Grouping experiment outcomes

**Algorithms:** k-Means, spectral clustering, mean-shift, and more...



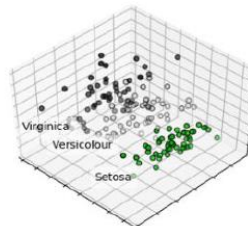
Examples

## Dimensionality reduction

Reducing the number of random variables to consider.

**Applications:** Visualization, Increased efficiency

**Algorithms:** k-Means, feature selection, non-negative matrix factorization, and more...



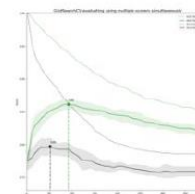
Examples

## Model selection

Comparing, validating and choosing parameters and models.

**Applications:** Improved accuracy via parameter tuning

**Algorithms:** grid search, cross validation, metrics, and more...



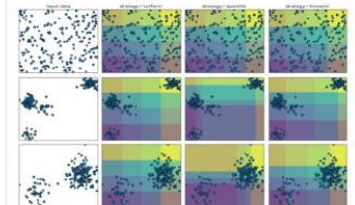
Examples

## Preprocessing

Feature extraction and normalization.

**Applications:** Transforming input data such as text for use with machine learning algorithms.

**Algorithms:** preprocessing, feature extraction, and more...



Examples