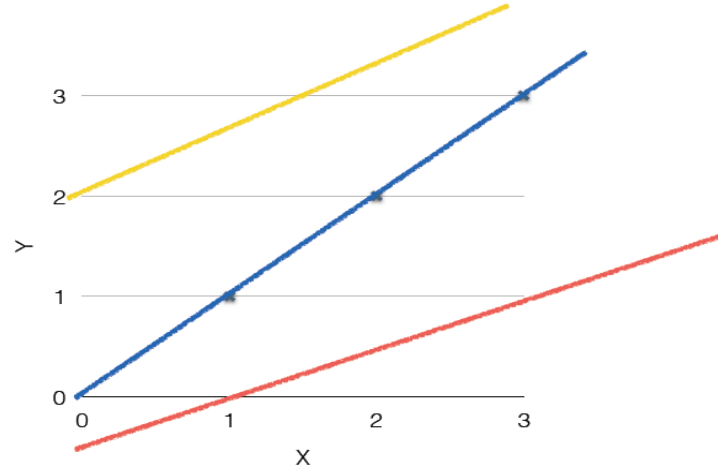


Machine Learning 1주차

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Lec2. Linear regression

- ▶ 어떤 선이 Data에 맞는 선일까?
-> 학습을 통해 찾는 것



- ▶ Step1. $H(x) = Wx + b$ 로 Hypothesis
 ↑ ↑
 b W
- ▶ Step2. cost function으로 최적 Linear찾기

Cost Function

- ▶ Cost function basis form $\Rightarrow H(x) - y$
- ▶ $H(x)$ 는 원하는 가설, y 는 실제 값
- ▶ $cost(W, b) = \frac{1}{m} \sum (H(x^i) - y^i)^2$ 로 정리
- ▶ 목표 : minimize $cost(W, b)$

Full code (less than 20 lines)

```
import tensorflow as tf
```

```
# X and Y data
```

```
x_train = [1, 2, 3]
```

```
y_train = [1, 2, 3]
```

```
W = tf.Variable(tf.random_normal([1]), name='weight')
```

```
b = tf.Variable(tf.random_normal([1]), name='bias')
```

```
# Our hypothesis  $XW+b$ 
```

```
hypothesis = x_train * W + b
```

```
# cost/loss function
```

```
cost = tf.reduce_mean(tf.square(hypothesis - y_train))
```

```
# Minimize
```

```
optimizer = tf.train.GradientDescentOptimizer(learning_rate=0.01)
```

```
train = optimizer.minimize(cost)
```

```
# Launch the graph in a session.
```

```
sess = tf.Session()
```

```
# Initializes global variables in the graph.
```

```
sess.run(tf.global_variables_initializer())
```

```
# Fit the line
```

```
for step in range(2001):
```

```
    sess.run(train)
```

```
    if step % 20 == 0:
```

```
        print(step, sess.run(cost), sess.run(W), sess.run(b))
```

중복?

```
...
```

```
0 2.82329 [ 2.12867713] [-0.85235667]
```

```
20 0.190351 [ 1.53392804] [-1.05059612]
```

```
40 0.151357 [ 1.45725465] [-1.02391243]
```

```
...
```

```
1920 1.77484e-05 [ 1.00489295] [-0.01112291]
```

```
1940 1.61197e-05 [ 1.00466311] [-0.01060018]
```

```
1960 1.46397e-05 [ 1.004444] [-0.01010205]
```

```
1980 1.32962e-05 [ 1.00423515] [-0.00962736]
```

```
2000 1.20761e-05 [ 1.00403607] [-0.00917497]
```

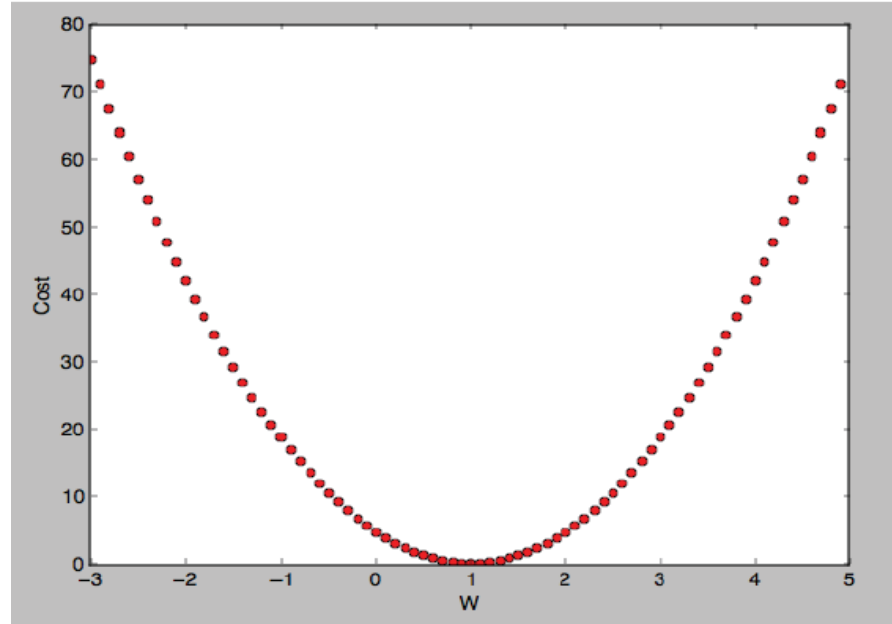
```
...
```

Lec3. How to minimize cost

▶ 우선 Simplified hypothesis

▶ $H(x) = Wx + b$

▶ $cost(W) = \frac{1}{m} \sum (Wx^i - y^i)^2$



Gradient descent algorithm

▶ $cost(W) = \frac{1}{m} \sum (Wx^i - y^i)^2$



▶ $cost(W) = \frac{1}{2m} \sum (Wx^i - y^i)^2$



중간에 사라지는 이유??

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

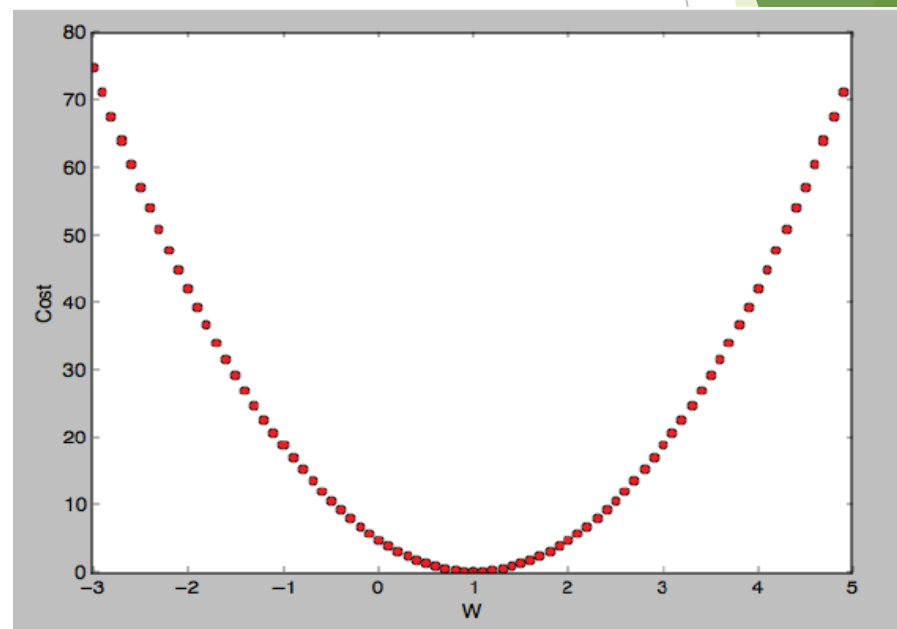
Formal definition

충분히 작지 않으면?

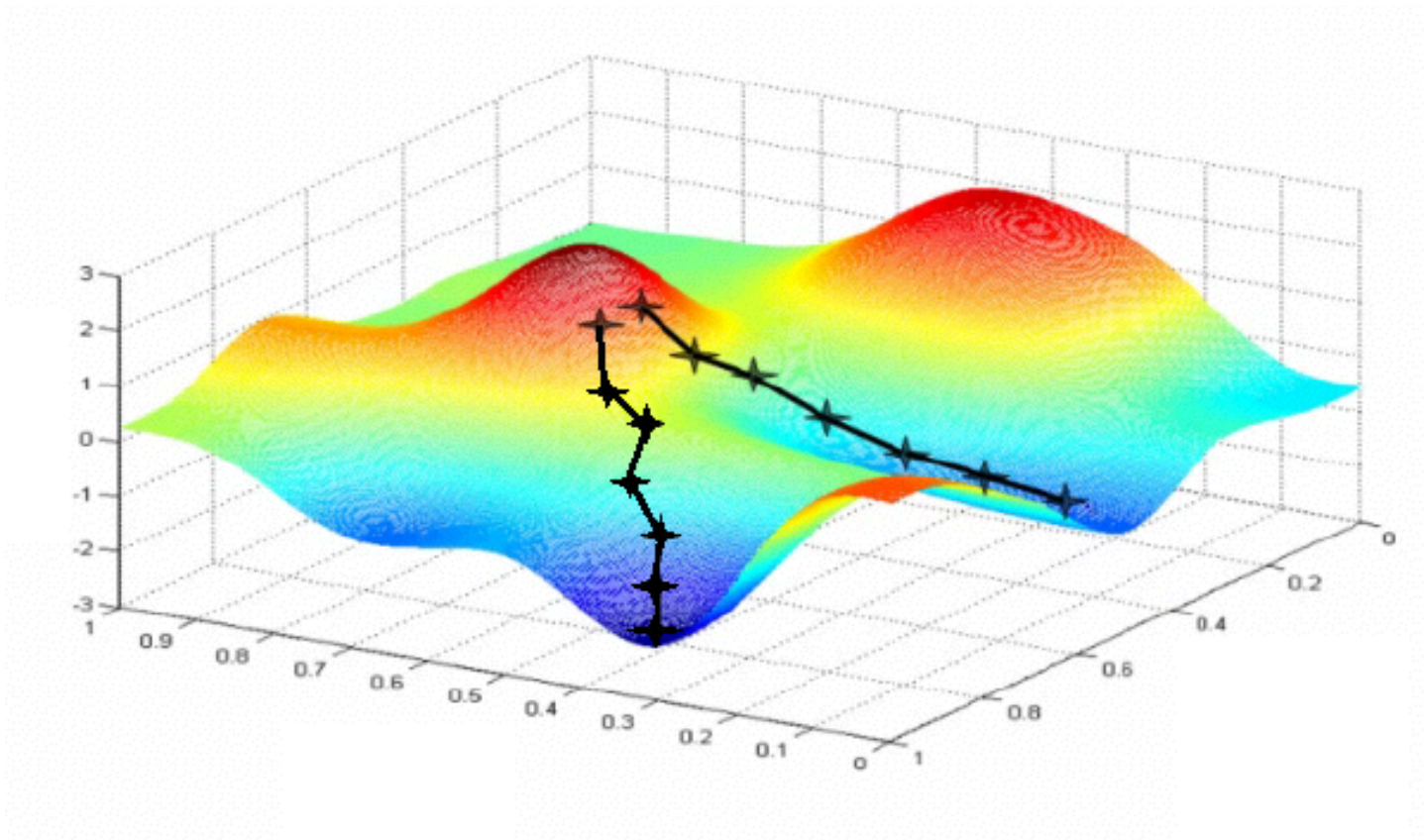
$$W := W - \alpha \frac{\partial}{\partial W} \frac{1}{2m} \sum_{i=1}^m (W x^{(i)} - y^{(i)})^2$$

$$W := W - \alpha \frac{1}{2m} \sum_{i=1}^m 2(W x^{(i)} - y^{(i)}) x^{(i)}$$

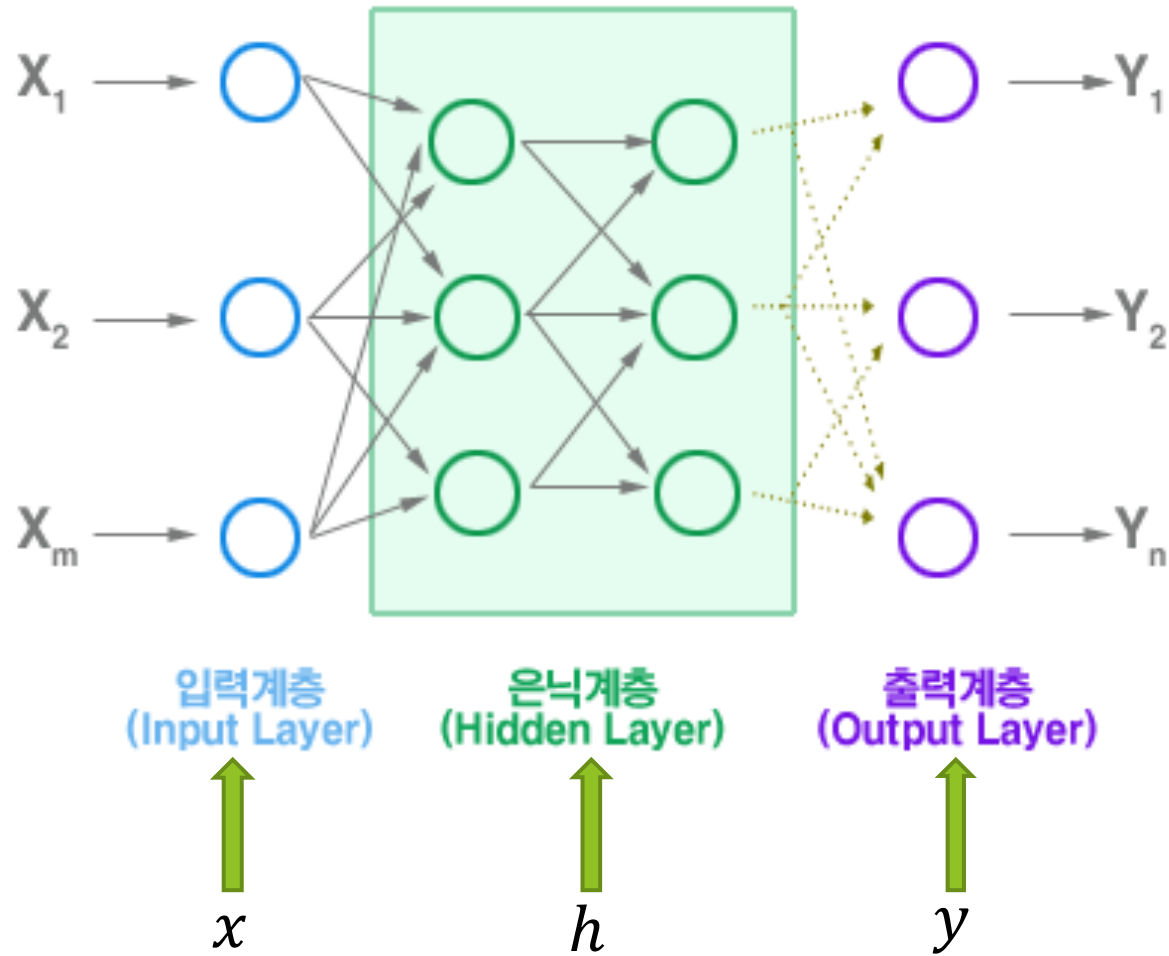
$$W := W - \alpha \frac{1}{m} \sum_{i=1}^m (W x^{(i)} - y^{(i)}) x^{(i)}$$



Non-convex



Lec4. 종합 정리



$$h_1 = f(w_1x_1 + w_2x_2 + w_3x_3 + b) ; (b \text{는 bias, 평행이동})$$

OSError: data-01-test-score.csv not found.

```
In [14]: xy = np.loadtxt('c:\\Users\\Jee\\Desktop\\data-01-test-score.csv', delimiter=',', dtype=np.float32)
```

```
In [15]: x_data = xy[:, 0:-1]
```

```
In [16]: y_data = xy[:, [-1]]
```

```
In [17]: print(x_data.shape, x_data, len(x_data))
```

```
(6, 3) [[ 73.  80.  75.]
 [ 93.  88.  93.]
 [ 89.  91.  90.]
 [ 96.  98. 100.]
 [ 73.  66.  70.]
 [ 53.  46.  55.]] 6
```

```
In [18]: print(y_data.shape, y_data)
```

```
(6, 1) [[152.]
 [185.]
 [180.]
 [196.]
 [142.]
 [101.]]
```

```
In [19]: X = tf.placeholder(tf.float32, shape=[None, 3])
```

```
In [20]: Y = tf.placeholder(tf.float32, shape=[None, 1])
```

```
In [21]: W = tf.Variable(tf.random_normal([3, 1]), name='weight')
```

```
In [22]: b = tf.Variable(tf.random_normal([1]), name='bias')
```

```
In [23]: hypothesis = tf.matmul(X, W) + b
```

```
In [24]: cost = tf.reduce_mean(tf.square(hypothesis - Y))
```

```
In [25]: optimizer = tf.train.GradientDescentOptimizer(learning_rate=1e-5)
```

```
In [26]: train = optimizer.minimize(cost)
```

```
In [27]: sess = tf.Session()
```

2018-03-14 17:50:52.023087: I C:\tf_jenkins\workspace\rel-win\windows\WPY\36\tensorflow\core\platform\cpu_feature_guard.cc:140] Your CPU supports instructions that this TensorFlow binary was not compiled to use: AVX2

```
In [28]: sess.run(tf.global_variables_initializer())
```

```
In [29]: for step in range(2001):
```

```
    ...:     cost_val, hy_val, _ = sess.run([cost, hypothesis, train], feed_dict={X: x_data, Y: y_data})
```

```
    ...:     if step % 10 == 0:
```

```
    ...:         print(step, "Cost: ", cost_val, "\nPrediction:\n", hy_val)
```

```
    ...:
```

```
2000 Cost: 2.7586195
```

```
Prediction:
```

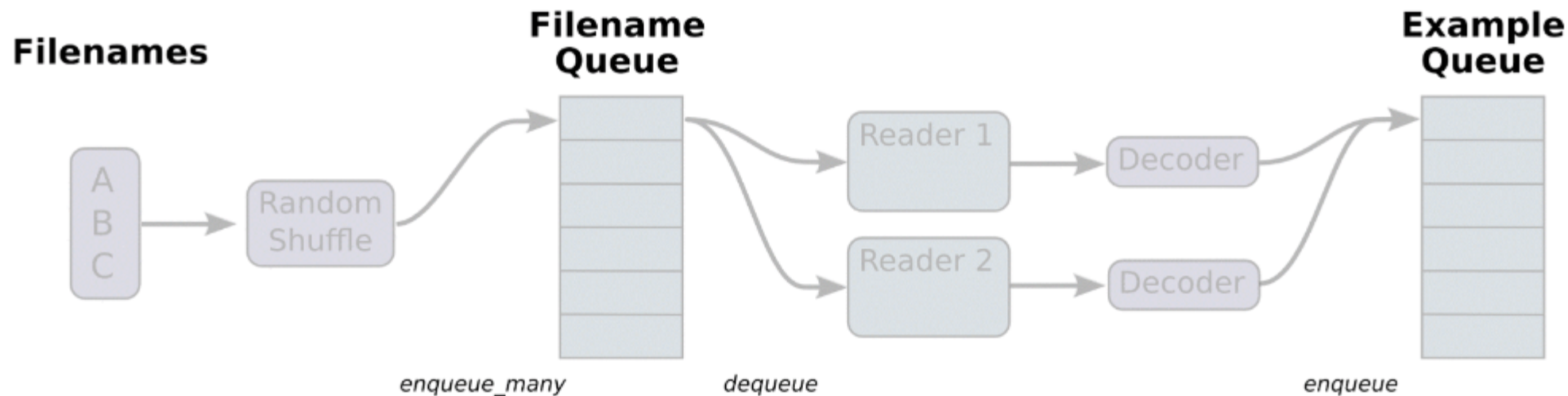
```
[[154.27278 ]
 [183.32866 ]
 [181.96426 ]
 [194.5939  ]
 [142.35019 ]
 [ 99.376816]]
```

```
In [30]: print("Your score will be ", sess.run(hypothesis, feed_dict={X: [[100, 70, 101]]}))
```

```
Your score will be [[171.15341]]
```

1 `filename_queue = tf.train.string_input_producer(
 ['data-01-test-score.csv', 'data-02-test-score.csv', ...],
 shuffle=False, name='filename_queue')`

3 `record_defaults = [[0.], [0.], [0.], [0.]]
xy = tf.decode_csv(value, record_defaults=record_defaults)`



2 `reader = tf.TextLineReader()
key, value = reader.read(filename_queue)`

Generalized expression?(irls)

Iteratively reweighted least squares

From Wikipedia, the free encyclopedia

For iterated weighted least squares, see [Feasible generalized least squares](#).

The method of **iteratively reweighted least squares** (**IRLS**) is used to solve certain optimization problems with [objective functions](#) of the form:

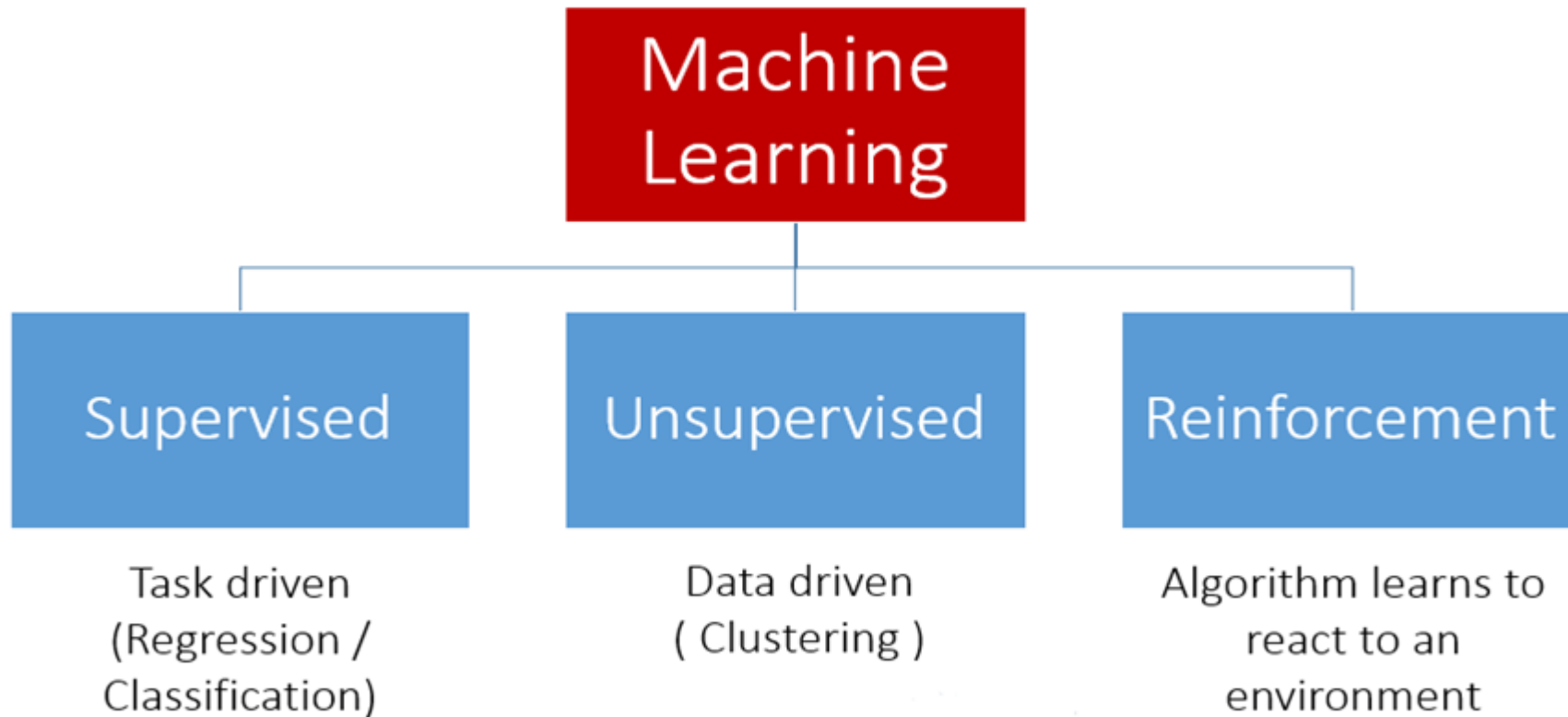
$$\arg \min_{\boldsymbol{\beta}} \sum_{i=1}^n |y_i - f_i(\boldsymbol{\beta})|^p,$$

by an [iterative method](#) in which each step involves solving a [weighted least squares](#) problem of the form:^[1]

$$\boldsymbol{\beta}^{(t+1)} = \arg \min_{\boldsymbol{\beta}} \sum_{i=1}^n w_i(\boldsymbol{\beta}^{(t)}) |y_i - f_i(\boldsymbol{\beta})|^2.$$

Intro(수업x)

Types of Machine Learning



수업 정리

- ▶ 보통은 fully connected : 이전 단계 모든 뉴런과 연결
- ▶ CNN은 fully connected 하지 않는다.
- ▶ DNN은 더 복잡한 함수 표현 가능하다.
- ▶ (chain rule??)
- ▶ (nonlinear 함수를 복잡하게)

수업 정리

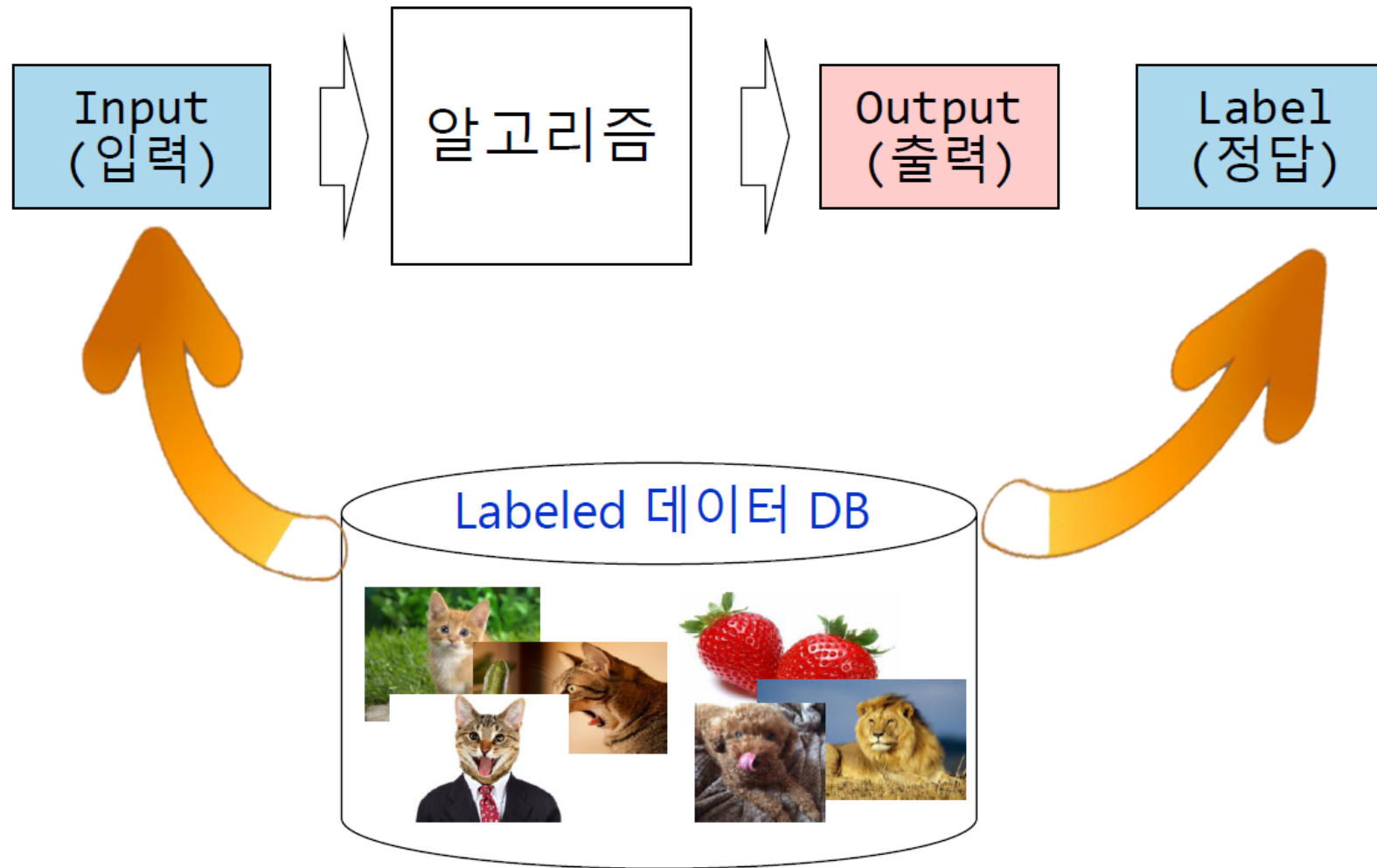
▶ Supervised learning

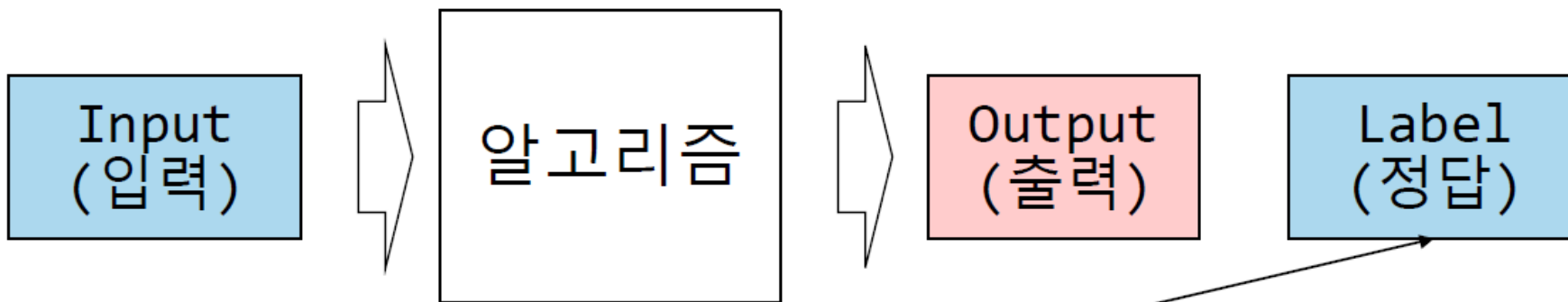
- ▶ 데이터가 충분히 필요하다.(labeling또한 필요)
- ▶ 정답도 같이 수집(high cost)

▶ Unsupervised learning

- ▶ 데이터 label이 없는 것(feature 입력 x)
- ▶ 데이터 축약(feature)->데이터 복원
- ▶ data driven approach->딥러닝이 feature expression (특징추출)

Supervised Learning(지도 학습)

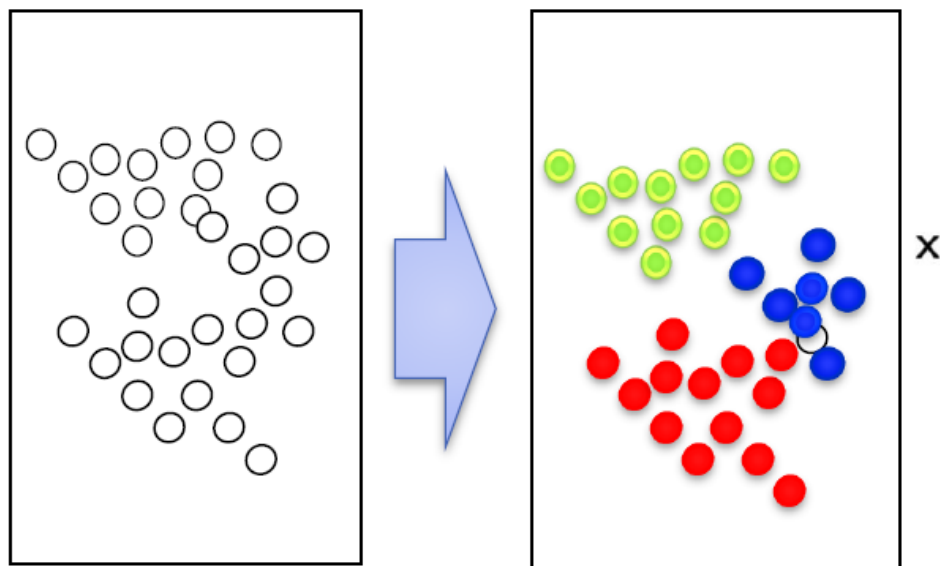




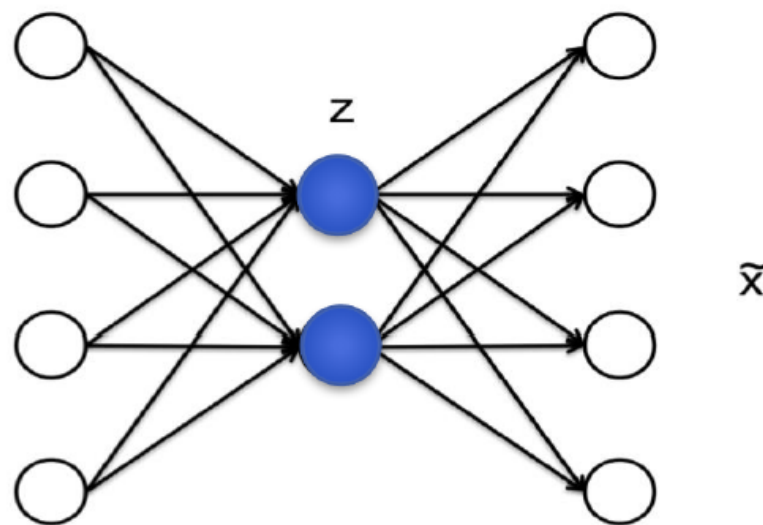
- **Label**은 대개 정확한 해답이나 정확한 분류(class)를 말함
- **Classification** : 정확한 class
 - + 예) 0~9이미지 분류시 label은 0~9
- **Regression** : 정확한 수치
 - + 환율 예측시 환율이 label

Unsupervised Learning (비지도 학습)

- 데이터의 label이 없는 데, 패턴을 알고 싶을 때
- Clustering 등
- 데이터 압축



Clustering



Reconstruction Error

수업 정리

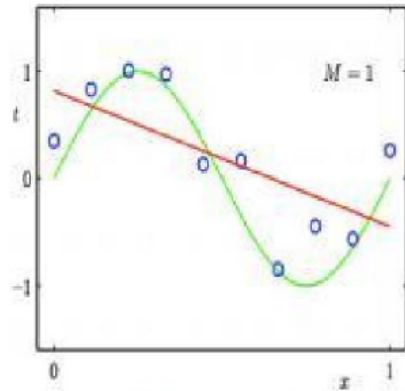
- ▶ 1) overfitting-> 너무 크게 잡는 것,
(이미 뭉든 데이터 샘플을 암기)
▶ training sample작을 때 overfitting일어난다.
- ▶ 2) underfitting-> network가 충분히 깊고 크기 않아서 혹은 데이터가 작아서
- ▶ Deep learning인 경우 충분히 layer가 깊어야 효율UP (bottleneck 제거)

Under- and Over-fitting examples

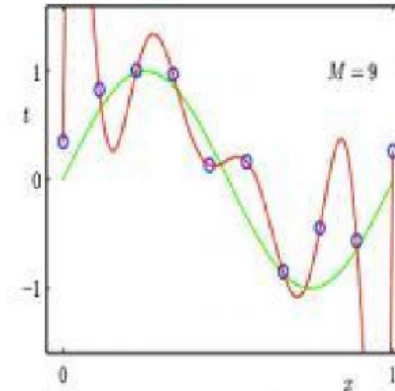
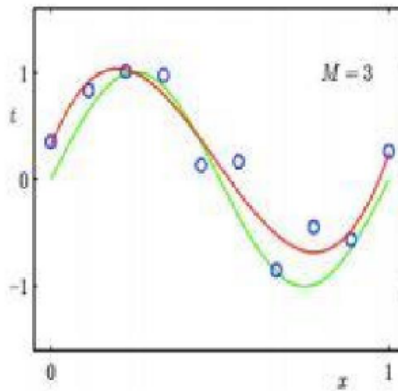
underfitting

overfitting

Regression:



predictor too inflexible:
cannot capture pattern



predictor too flexible:
fits noise in the data

Classification:

