딥러닝 몸풀기

누구나 이해할 수 있는 딥러닝

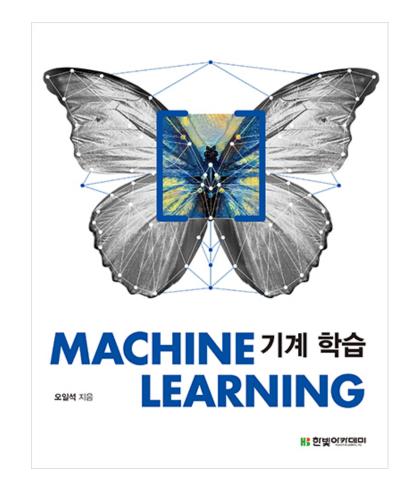
Local Laboratory

딥러닝 몸풀기

Local Laboratory

- 1. 인공지능과 기계학습 그리고 딥러닝
- 2. 다층 퍼셉트론
- 3. 딥러닝의 기초

Reference





기계학습, 오일석

Self-Driving Car Nano degree, Udacity

ARTIFICIAL INTELLIGENCE

A program that can sense, reason, act, and adapt

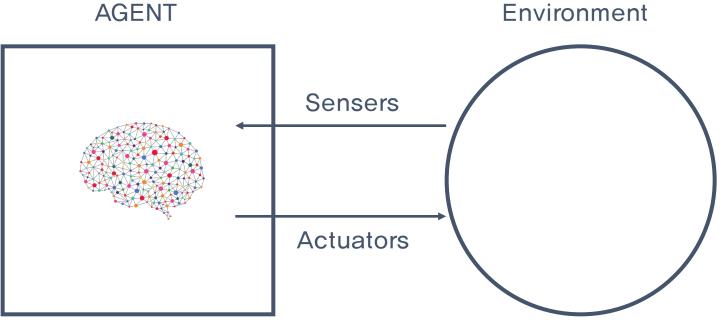
MACHINE LEARNING

Algorithms whose performance improve as they are exposed to more data over time

DEEP LEARNING

Subset of machine learning in which multilayered neural networks learn from vast amounts of data



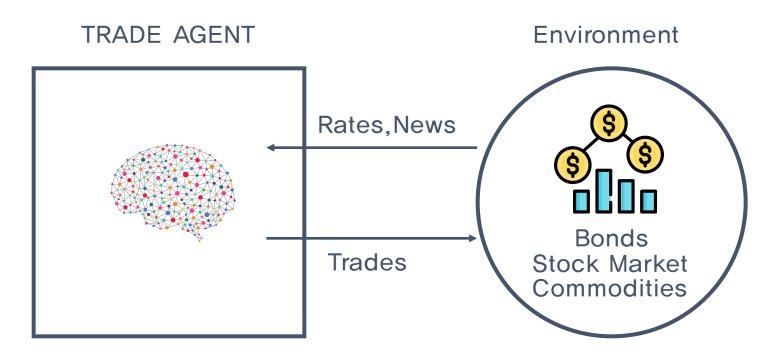


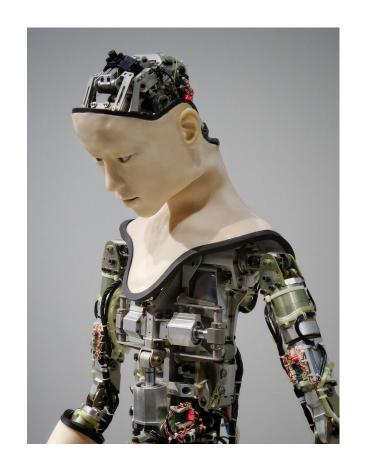
- Finance
- Robotics
- Games

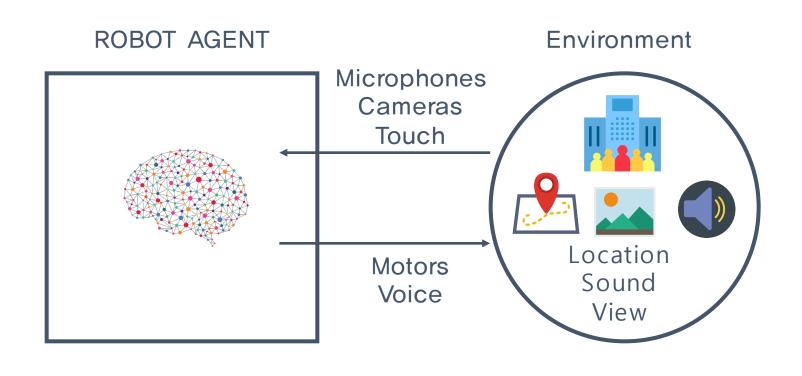
- Medicine
- Web
- Etc.

에이전트는 특정 목적에 대해 사용자를 대 신하여 작업을 수행하는 자율적 프로세스 (autonomous process)

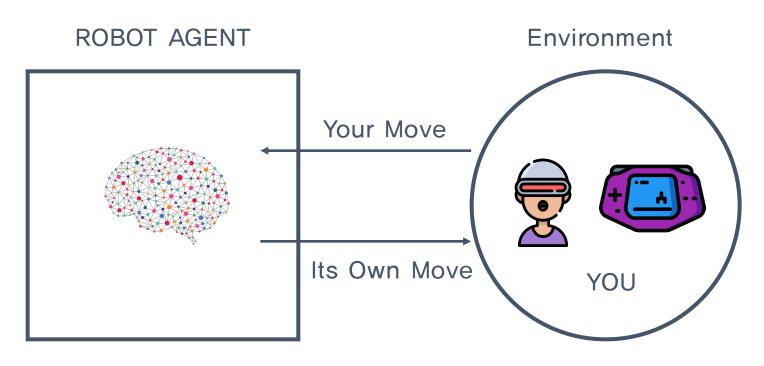


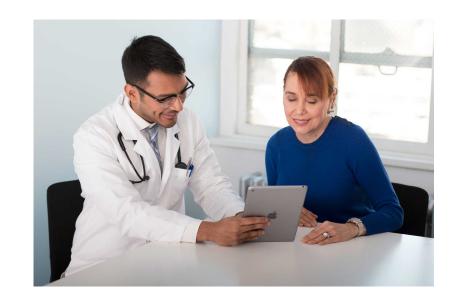


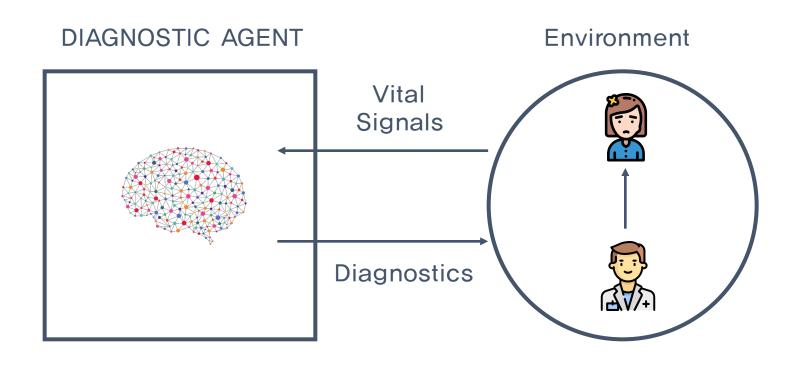




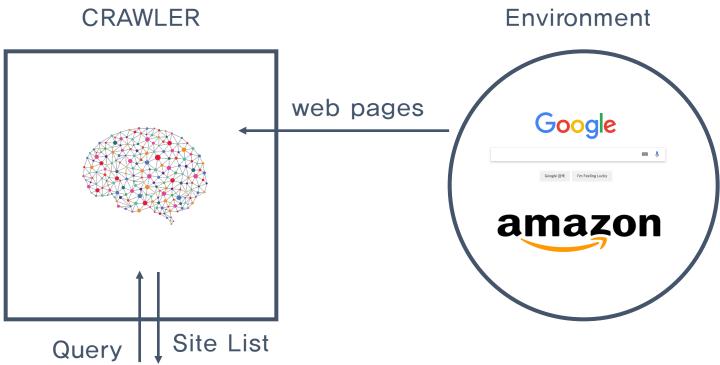




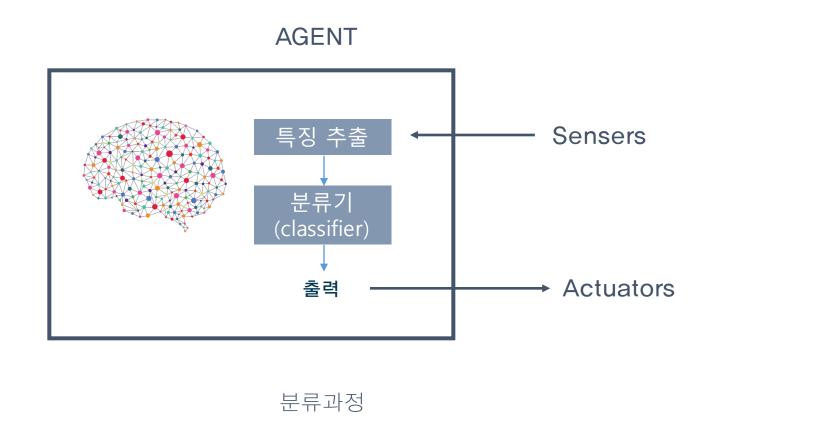


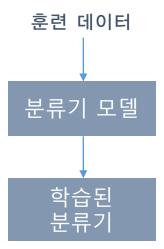




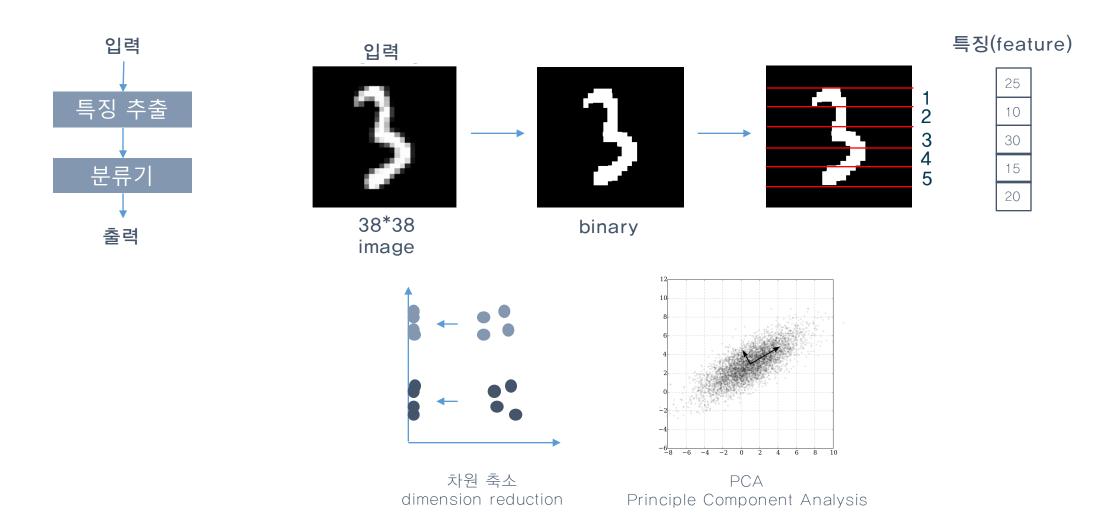








학습과정

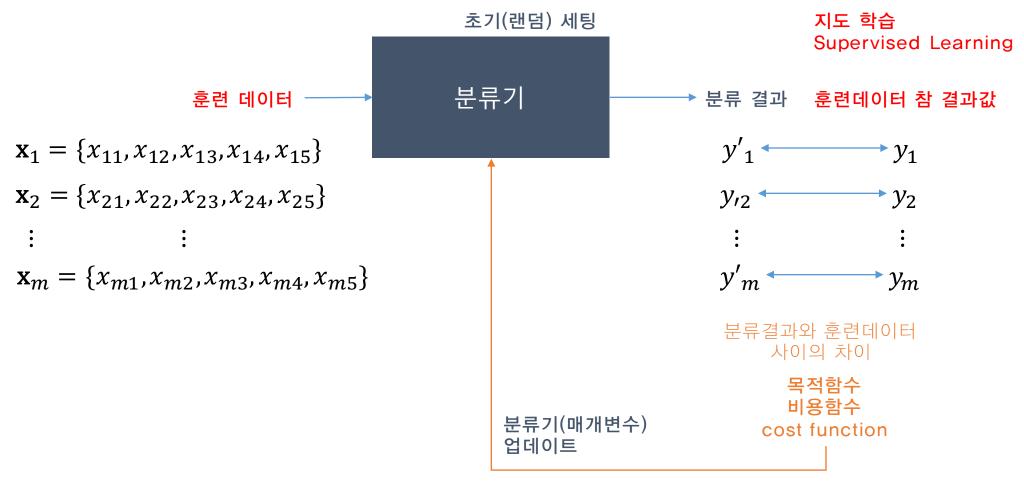


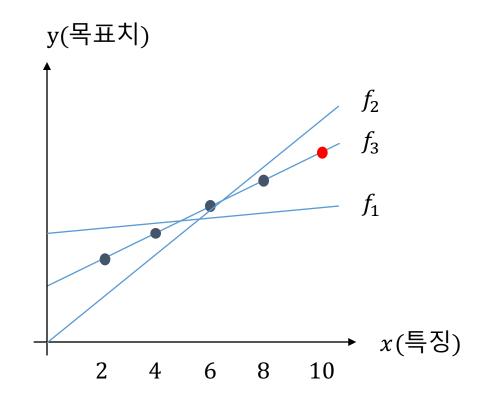
특징(feature)

x_1
x_2
x_3
x_4
x_5

$$\mathbf{x} = (x_1, x_2, x_3, x_4, x_5)^T = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix}$$

Machine Learning = Learn models from data





분류기 모델: 직선 모델

$$f_{\Theta}(x) = wx + b$$

추정해야할 매개변수

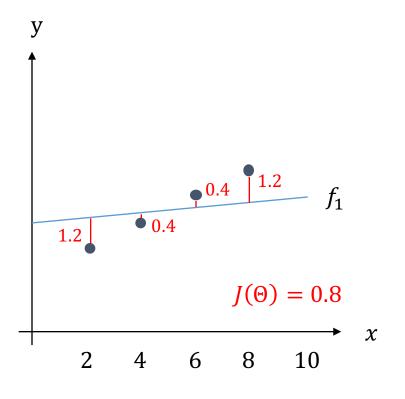
$$\Theta = (w, b)^T$$

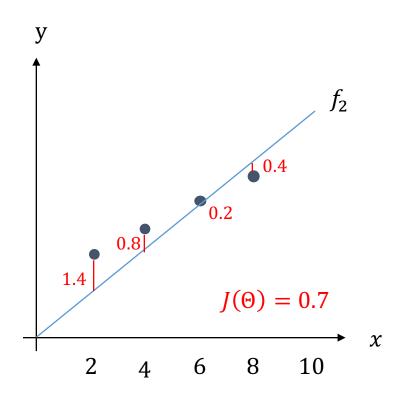
목적 함수: 평균 제곱 오차

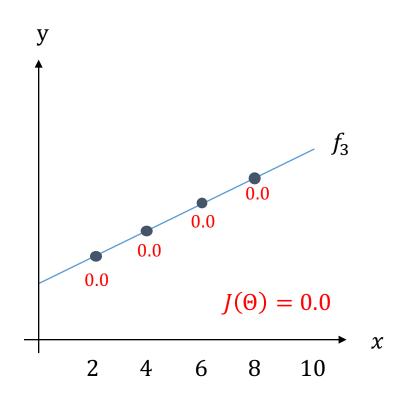
$$J(\Theta) = \frac{1}{n} \sum_{i=1}^{n} (f_{\Theta}(X_i) - y_i)^2$$

파라미터 변화에 따른 목적함수 값 변화 목적함수 $J(\Theta) = \frac{1}{n} \sum_{i=1}^{n} (f_{\Theta}(X_i) - y_i)^2$

$$J(\Theta) = \frac{1}{n} \sum_{i=1}^{n} (f_{\Theta}(X_i) - y_i)^2$$







난수를 생성하여 초기해 Θ_1 설정

$$t = 1$$

while $(J(\Theta_t)$ 가 0.0에 충분히 가깝지 않음)

 $J(\Theta_t)$ 가 작아지는 방향 $\Delta\Theta_t$ 를 구한다.

$$\Theta_{t+1} = \Theta_t + \Delta\Theta_t$$

$$t = t + 1$$

$$\widehat{\Theta} = \Theta_t$$

Price of House



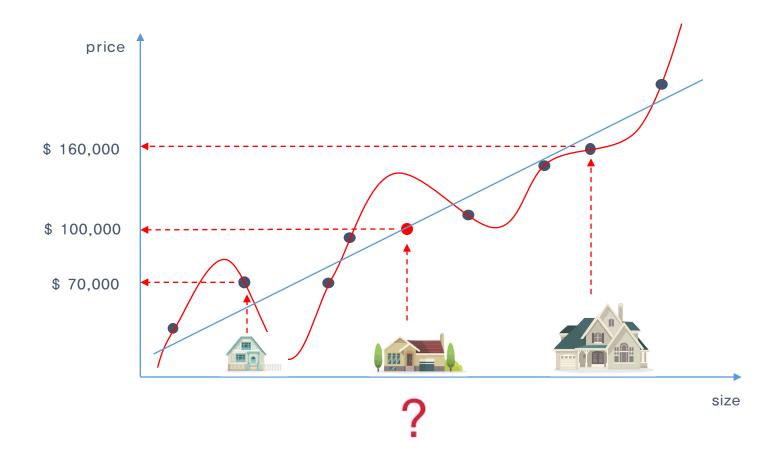
\$70,000

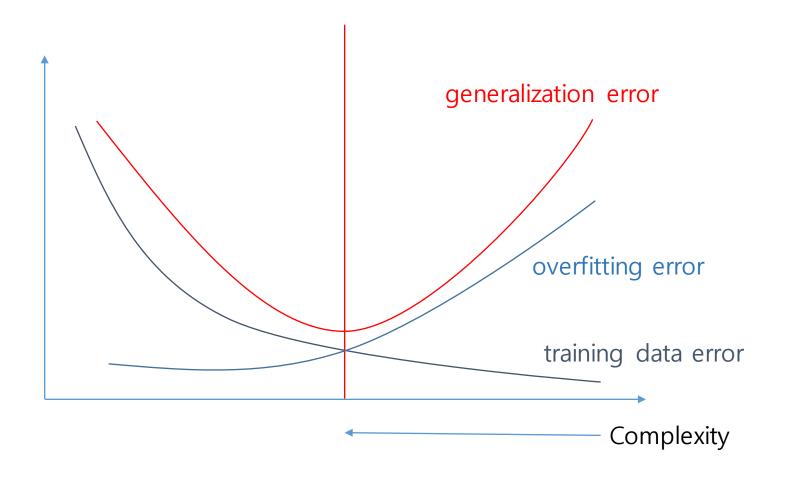


?

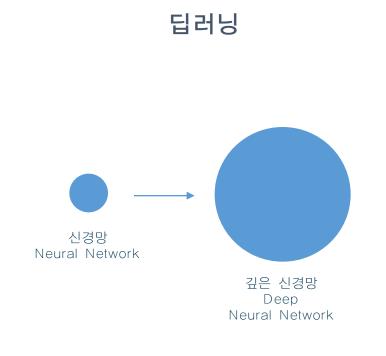


\$160,000





이전의 기계학습 K-Means 베이지안 K-NN 분류기 신경망 SVM 결정트리



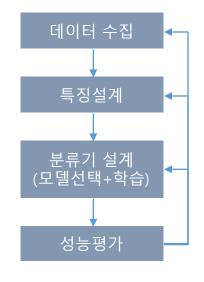
특징공간 변환

이전의 기계학습: 전처리를 통한 특징 공간 변환

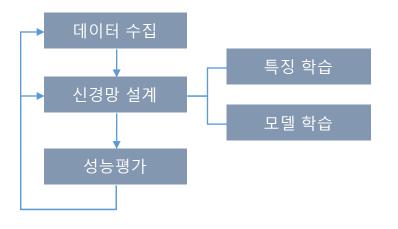
딥러닝: 표현학습 representation learning

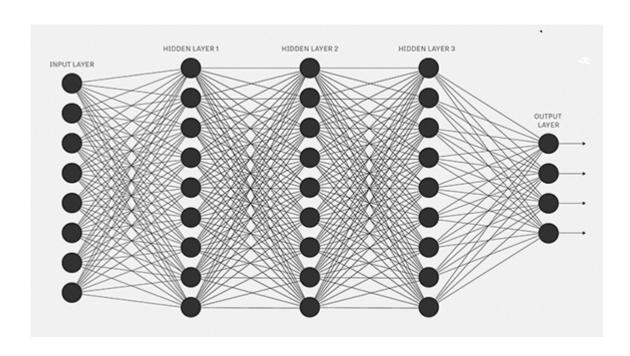
feature learning, representation learning 22

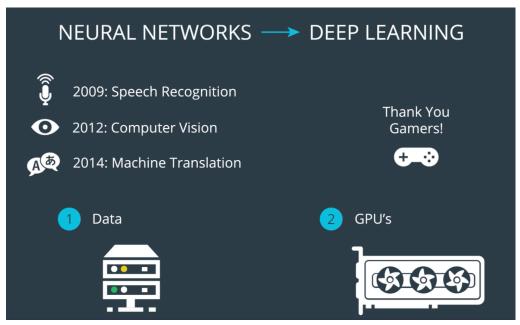
이전의 기계학습



현재의 기계학습(딥러닝)







값싼 GPU / 학습데이터 증가 성능 좋은 활성함수 개발 효과적인 규제 기법 개발