

KUBIG 24-S
여름방학 BASIC STUDY SESSION

NLP SESSION

WEEK1

Session 중에는 모두 캠을 켜주시기 바랍니다 :)

01 NLP SESSION 소개

02 자기 소개

03 Deep Learning Reminder

04 과제 설명

01 NLP SESSION 소개

방학동안 무엇을 하나요?

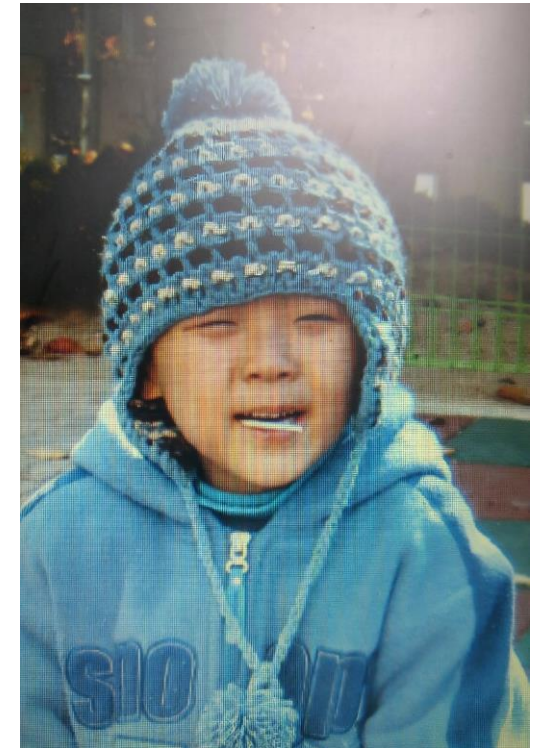
한 학기동안 잘 부탁드립니다!!



분반장 18기 정해원

- 분반장 18기 정해원
- 전 홍보팀장 출신
- DL 초심자에서 분반장까지 순항 중
- 최근 우리말 겨루기를 시청하며 언어 능력 습득 중

분반장 17기 김연규
관심 분야: 데이터 분석 -> 인공지능
프로그래밍과 더 친해지려고 노력 중
스터디와 친목 둘 다 환영



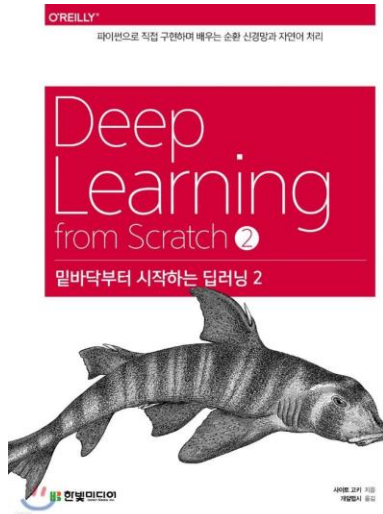
분반장 17기 김연규

주차	복습과제	학습내용	예습과제
1주차	<ul style="list-style-type: none"> - 해당 주차에 배운 내용에 대한 코드 실습 과제 부여 - ex) week1 복습과제: deep learning reminder(pytorch basic) - session 시작 초반부에 우수 코드 선정자가 5분 가량 코드 구현 과정 발표(별도 발표자료 없이 코드를 화면공유하여 발표) - 마감기한: 수요일 오후 6시 	OT, DL Reminder	<ul style="list-style-type: none"> - 다음 주차에 배울 내용에 대한 코드 실습 과제 부여 - ex) week1 예습과제: 텍스트 전처리 pipeline 구현 코드 - week1 예습과제라 함은, week1 session이 끝나고 부여되는, week2 내용에 대한 예습과제를 의미합니다! - 마감기한: 수요일 오후 6시
2주차		텍스트 전처리, 워드 임베딩 (Word2Vec, GloVe)	
3주차		RNN, LSTM, GRU, ELMo	
4주차		Attention, Transformer	
5주차		BERT	
6주차		GPT	
7주차		LLM 기초, HuggingFace	



매주 목 19:00~21:00 총 2시간 진행

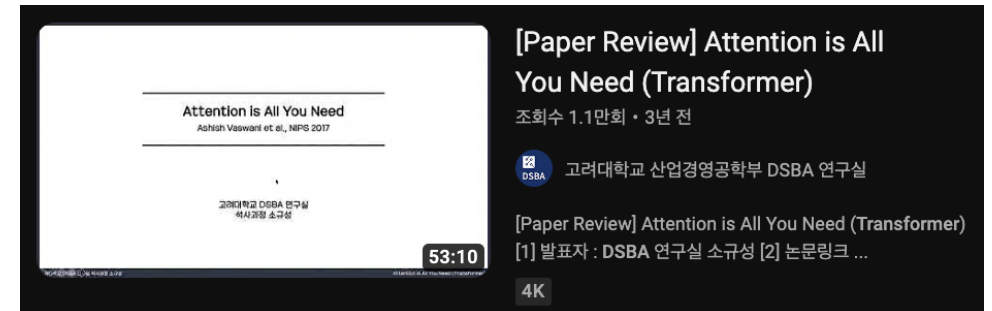




밑바닥부터 시작하는 딥러닝2



딥러닝을 이용한 자연어 처리 입문
위키독스에서 무료 이용 가능(부분 유료)



고려대학교 DSBA 연구실
유튜브 채널에서 paper review 영상 참고

02 자기 소개

친해지길 바라!

2. 자기 소개

한 학기 동안 함께 하실 분들 !

강동헌

강지윤

심승현

이동주

이소희

이승준

정종락

최지우

하진우

강민정

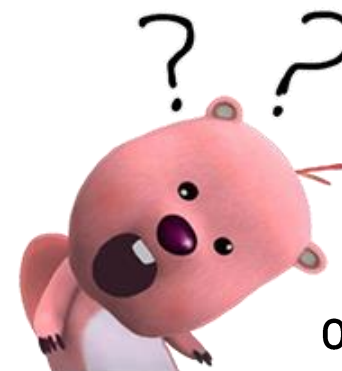
김채원

박준희

윤시호

이세은

이유진



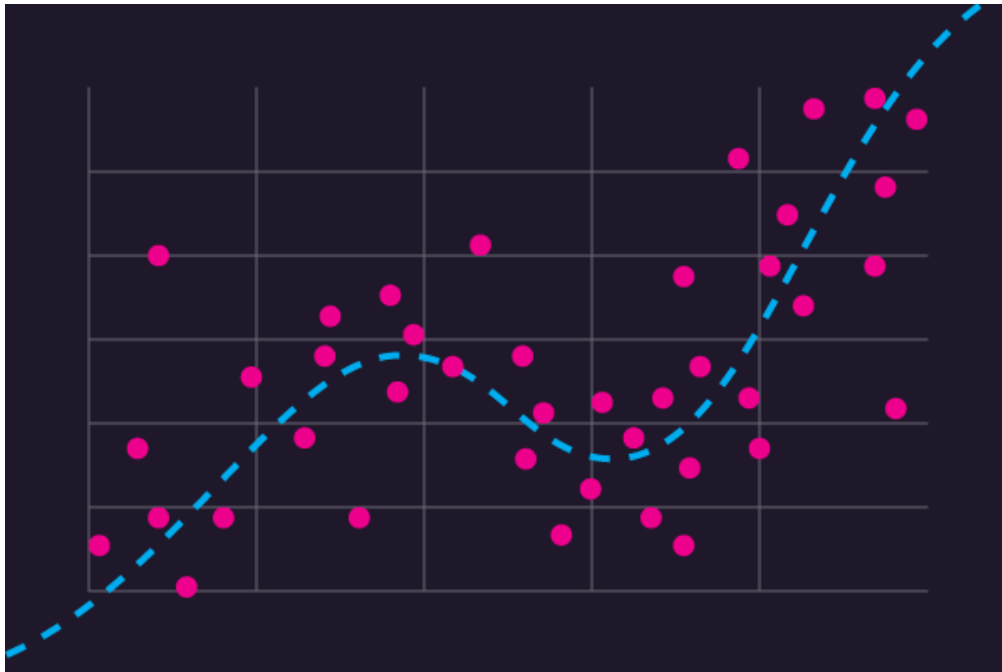
여러분을 소개해주세요!

기수, 이름, 학과, 나이, 사는 곳, 취미

NLP 경험, NLP 선택 이유, 원하는 방향성

03 Theoretical Background of Deep Learning

3-0. Estimate the True Function



What is the True Function?

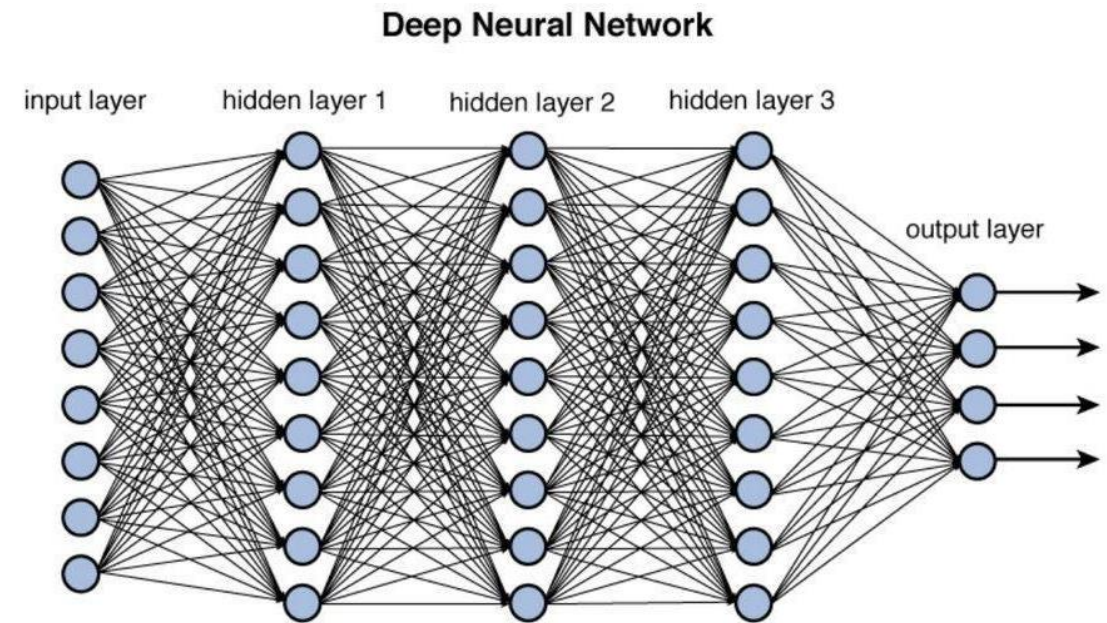
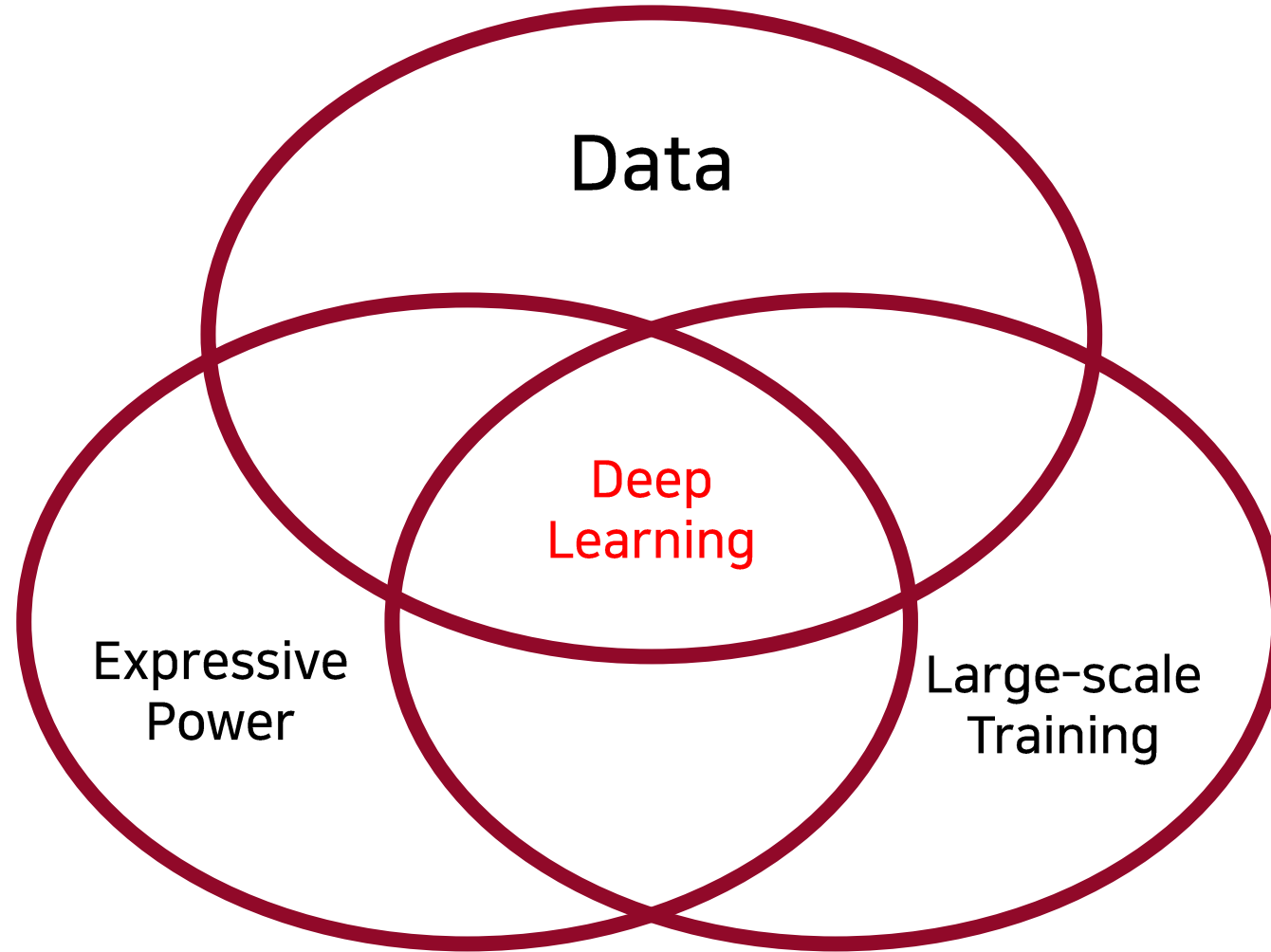
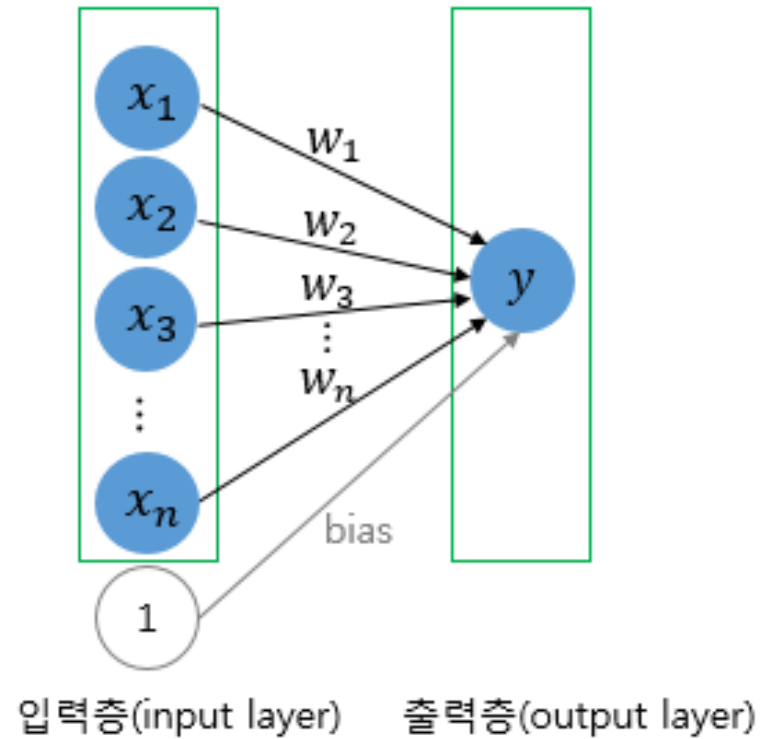


Figure 12.2 Deep network architecture with multiple layers.



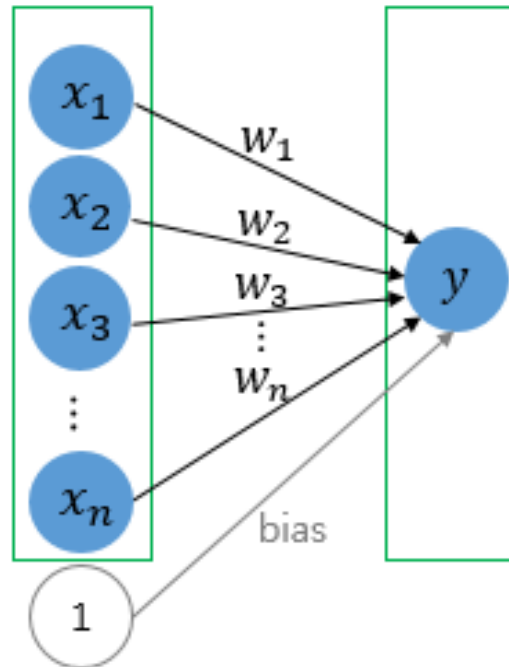
3-1. Single-Layer Perceptron

single-layer perceptron



3-1. Single-Layer Perceptron

single-layer perceptron

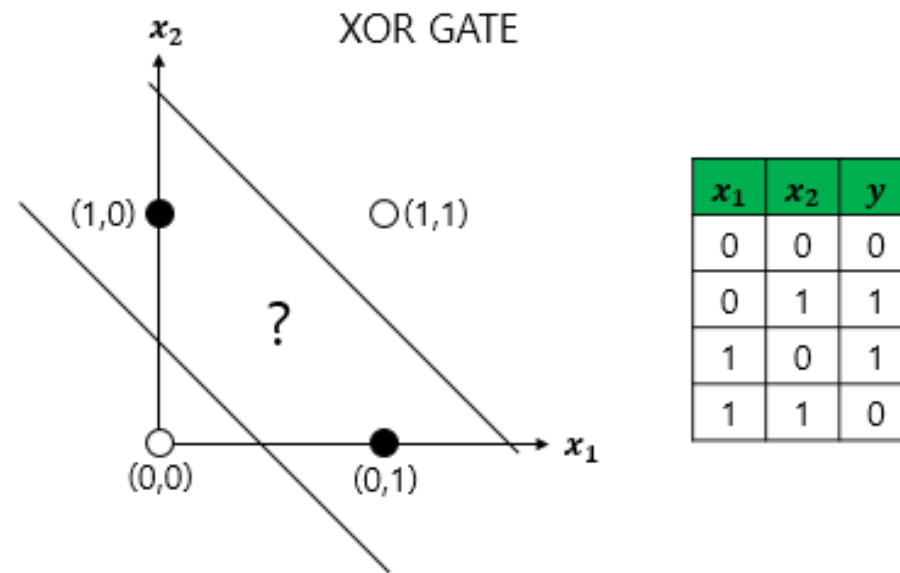


입력층(input layer) 출력층(output layer)

Q) How many parameters?
Q) What Dimension?

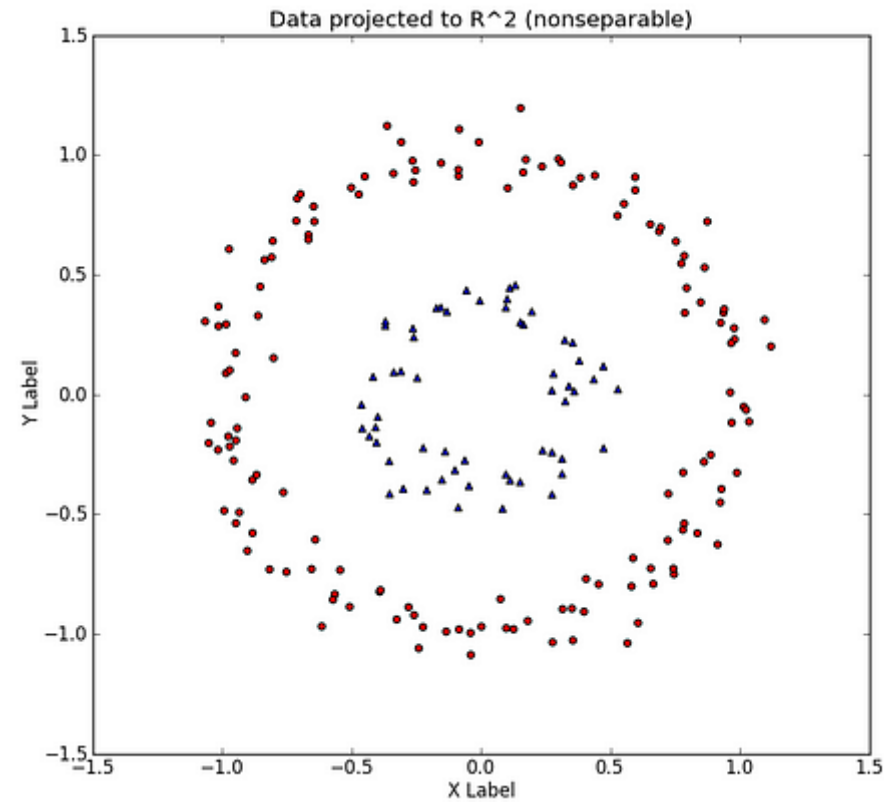
XOR problem

Can you separate B/W by a single line (in 2-dimension space)?



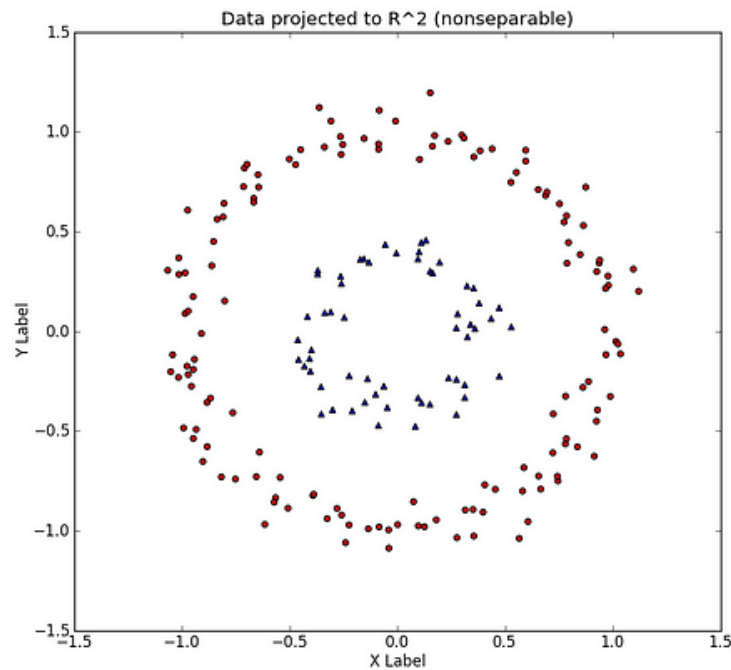
Linearly Non-Separable

(in 2-dimension space)

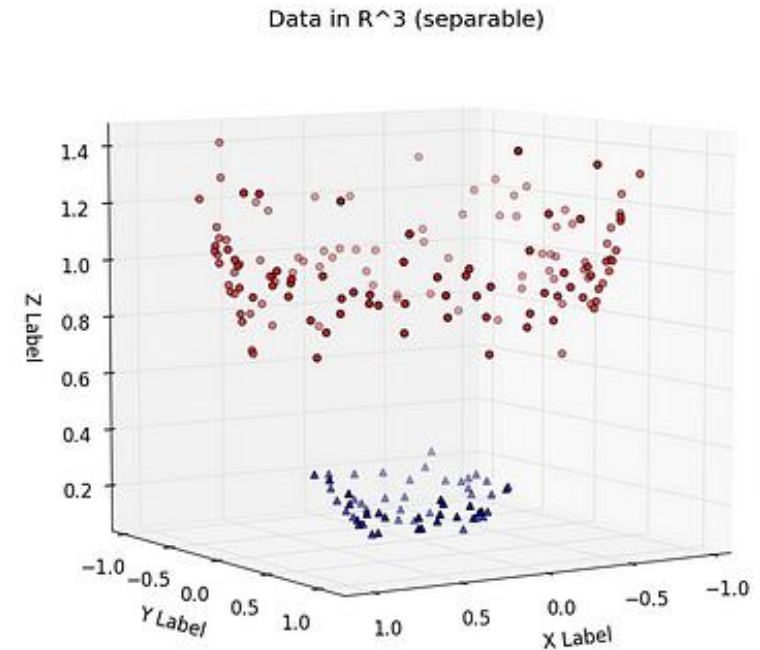


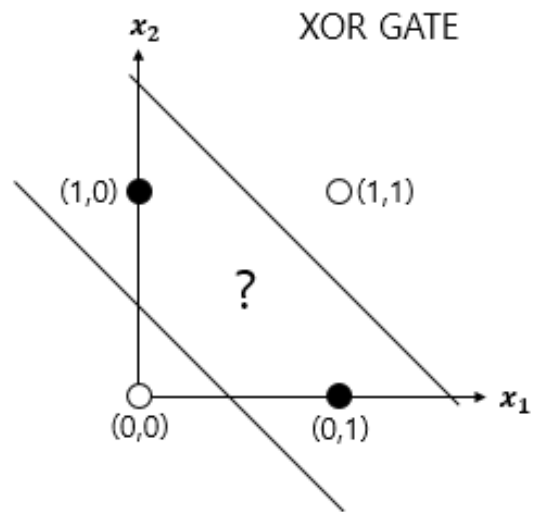
Linearly Non-Separable

(in 2-dimension space)



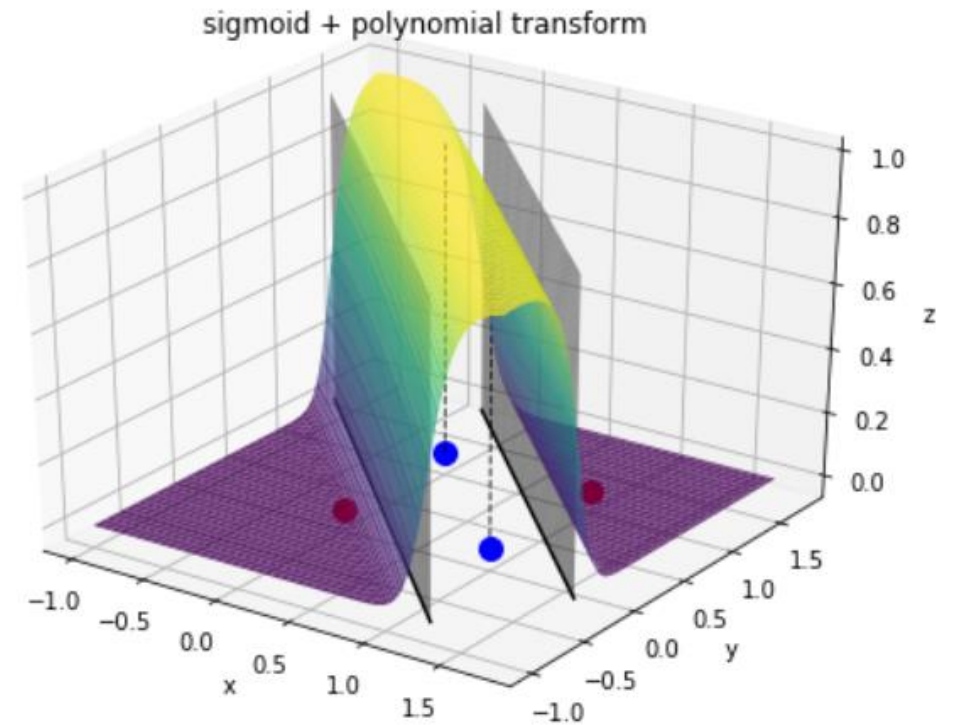
High-dimensional
Space



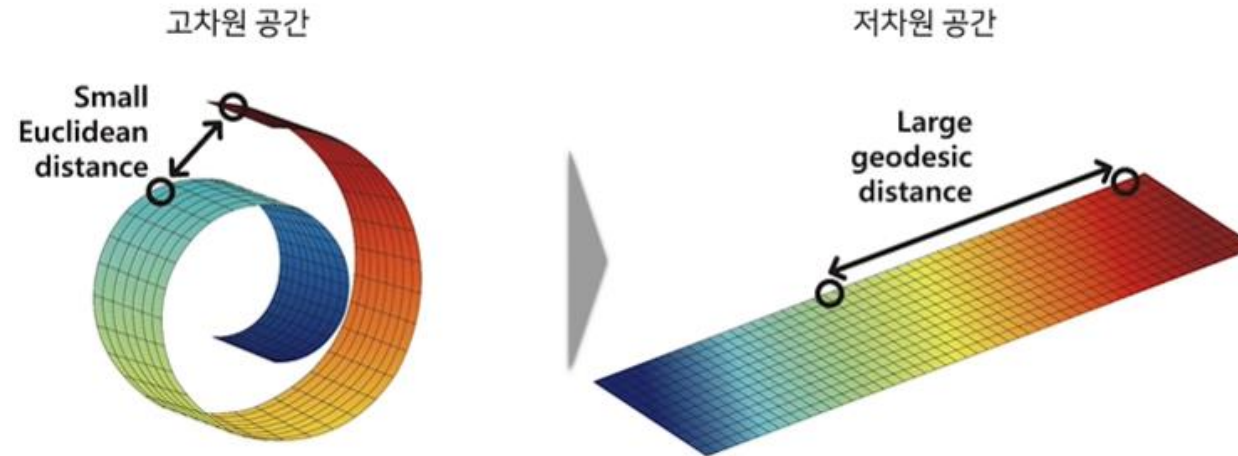


x_1	x_2	y
0	0	0
0	1	1
1	0	1
1	1	0

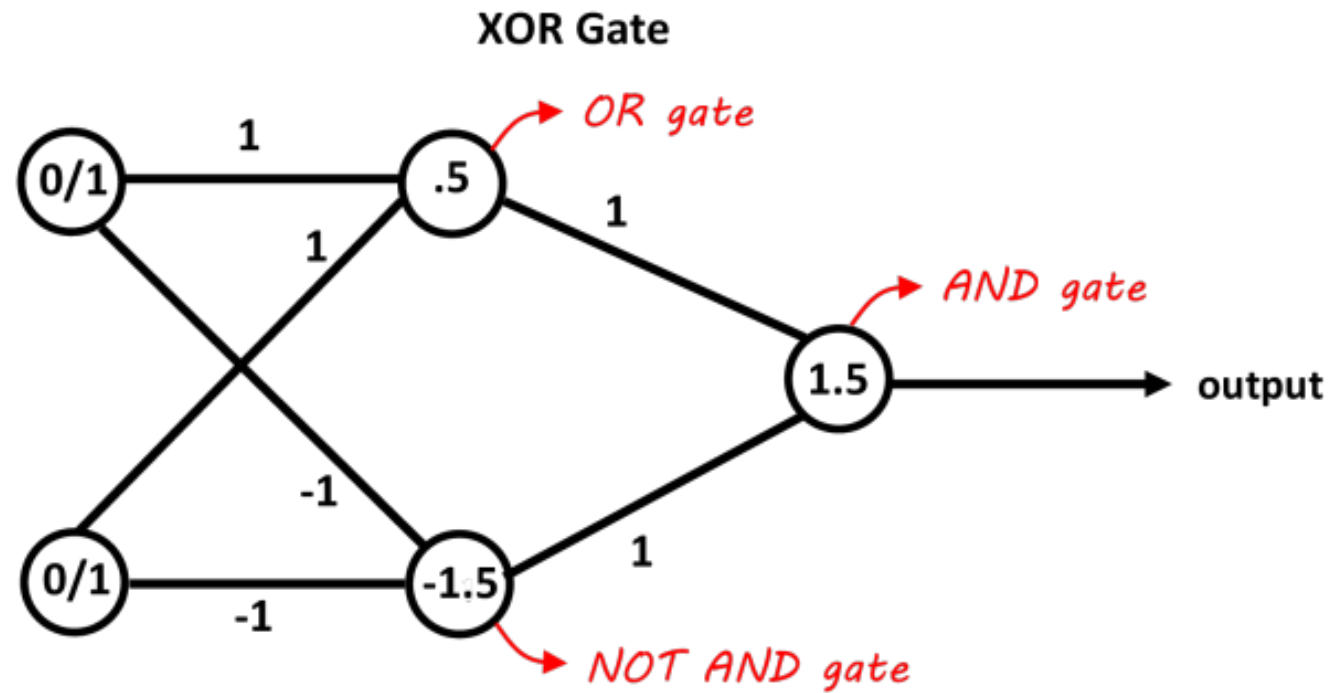
High-dimensional
Space



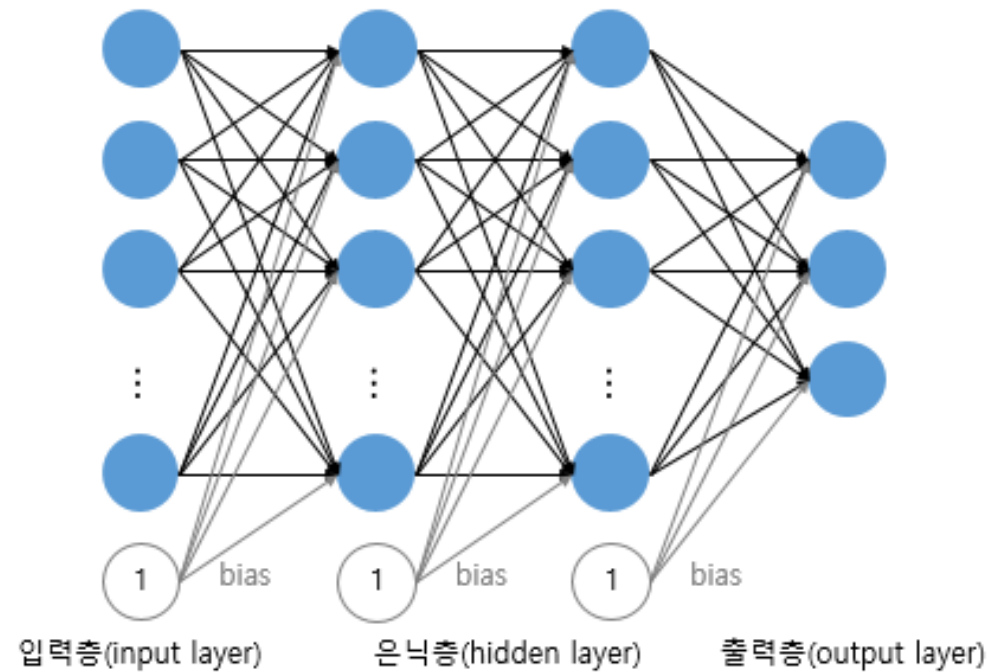
Manifold Hypothesis



- Hypothesis: High-dimensional data **tend to lie in the vicinity of a low-dimensional manifold**
- We can map data to a high-dimensional space through a smooth representation if the manifold hypothesis holds



Multi-Layer Perceptron (More Layers, much more parameters)



Universal Approximation Theorem

Why deep learning is powerful

Universal approximation theorem — Let $C(X, \mathbb{R}^m)$ denote the set of **continuous functions** from a subset X of a Euclidean \mathbb{R}^n space to a Euclidean space \mathbb{R}^m . Let $\sigma \in C(\mathbb{R}, \mathbb{R})$. Note that $(\sigma \circ x)_i = \sigma(x_i)$, so $\sigma \circ x$ denotes σ applied to each component of x .

Then σ is not **polynomial if and only if** for every $n \in \mathbb{N}$, $m \in \mathbb{N}$, **compact** $K \subseteq \mathbb{R}^n$, $f \in C(K, \mathbb{R}^m)$, $\varepsilon > 0$ there exist $k \in \mathbb{N}$, $A \in \mathbb{R}^{k \times n}$, $b \in \mathbb{R}^k$, $C \in \mathbb{R}^{m \times k}$ such that

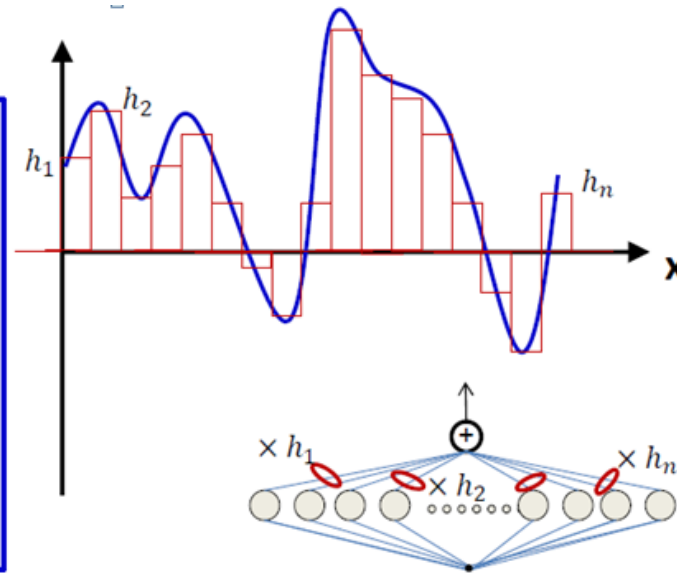
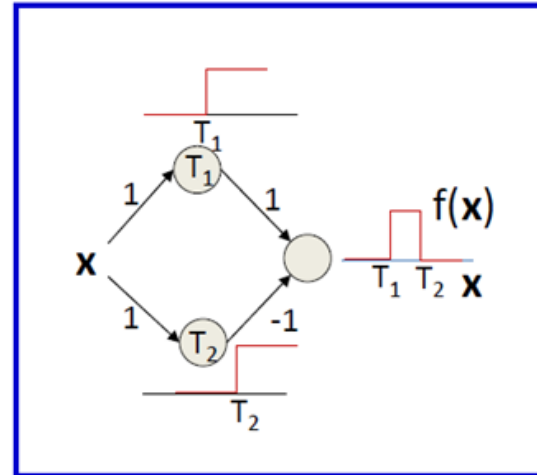
$$\sup_{x \in K} \|f(x) - g(x)\| < \varepsilon$$

where $g(x) = C \cdot (\sigma \circ (A \cdot x + b))$

(Universal Approximation Theorem) For a given arbitrary continuous function on a bounded domain and an error bound, there **always exists a one-hidden-layer neural network** that can approximate the given continuous function within the error bound.

Universal Approximation Theorem

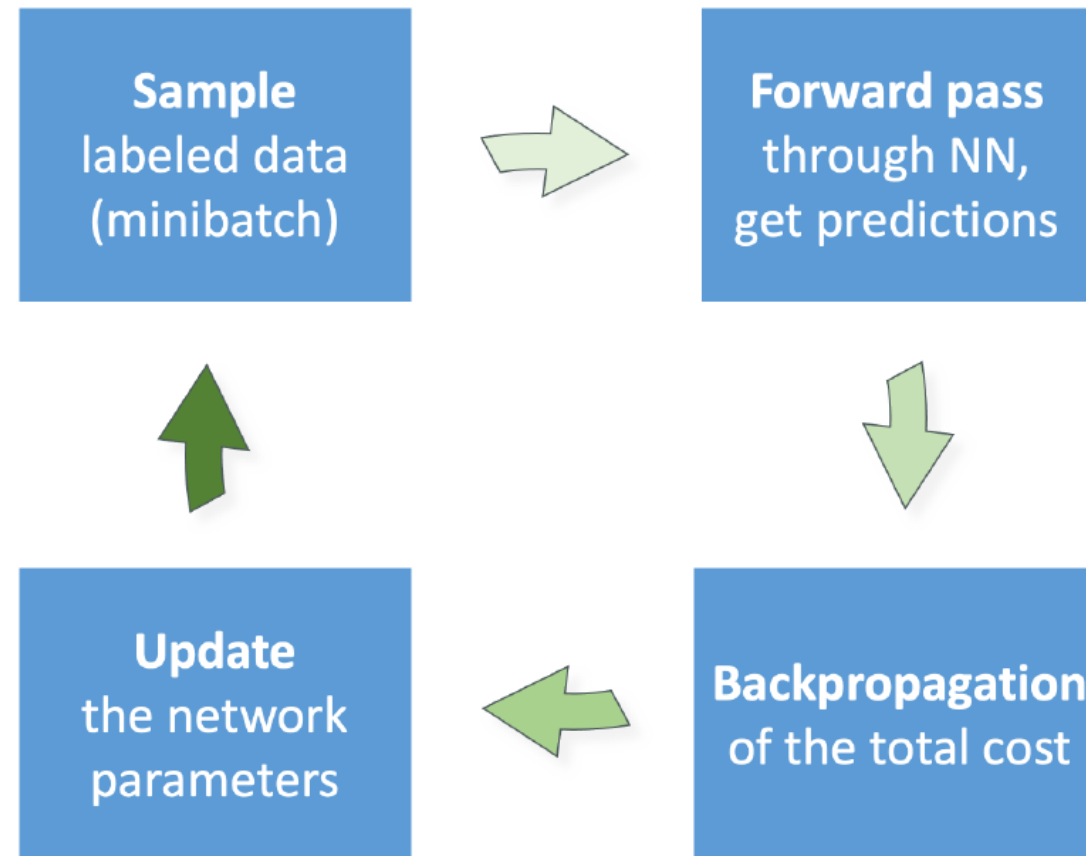
Why deep learning is powerful



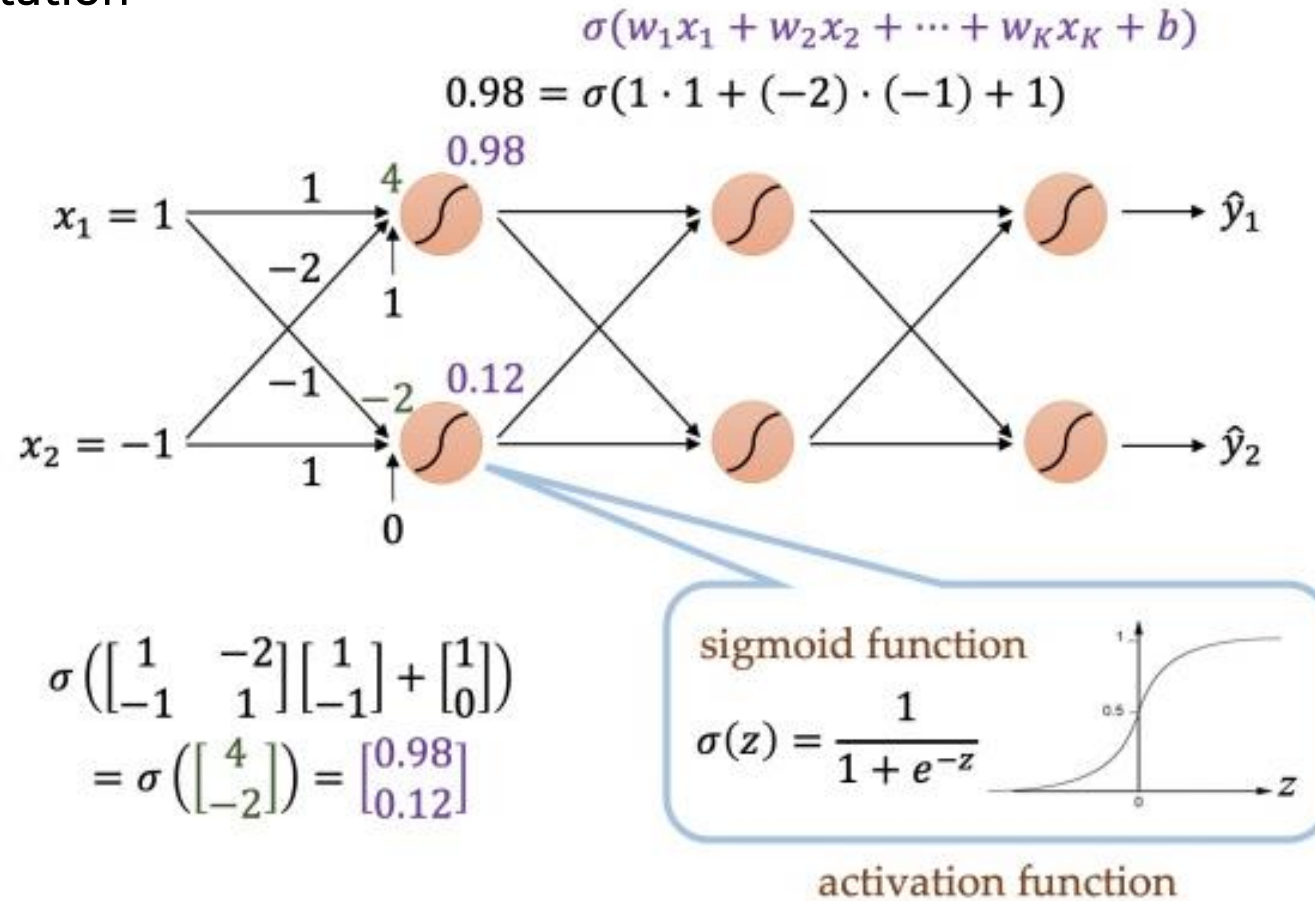
Increase in # of **hidden units** -> increase in # of **parameters**
-> increase in # of **regions** -> increase in # of **patterns a function can represent**

04 Factors of Deep Learning

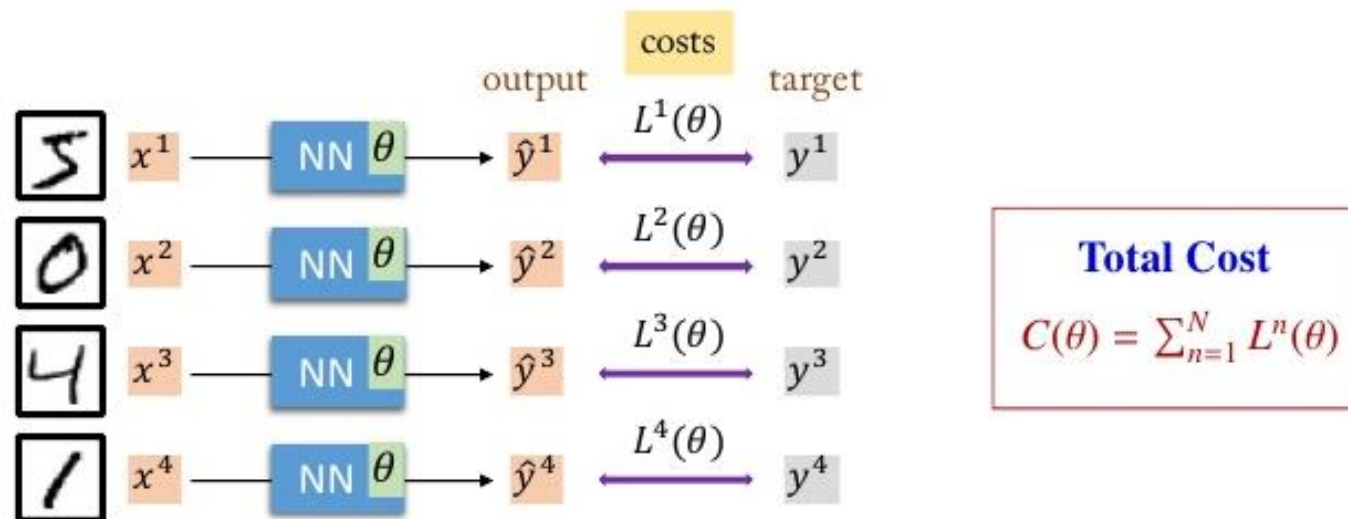
4-1. Deep Learning Training Cycle



Forward Pass computation

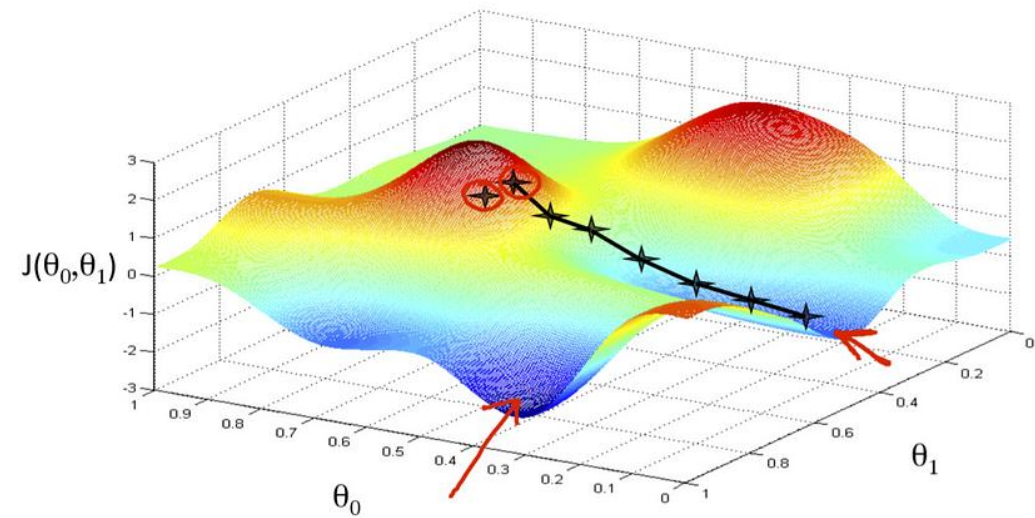
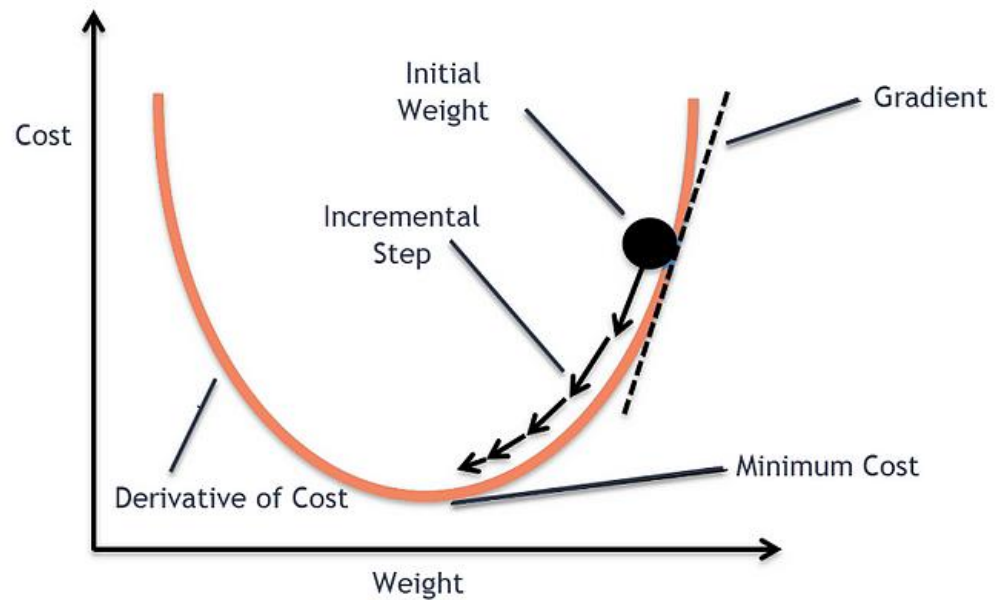


Total Cost



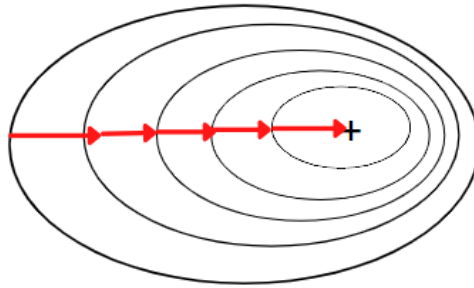
Total cost를 minimize하는 network parameter 찾기

Gradient Descent

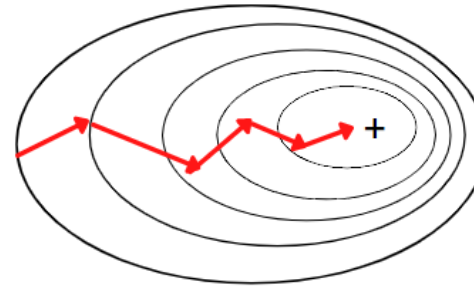


Gradient Descent

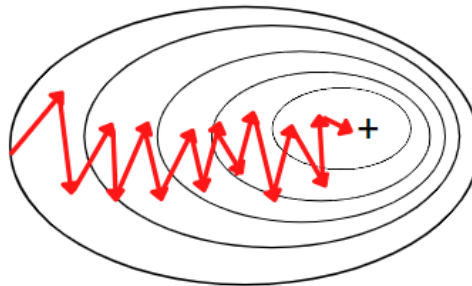
Batch Gradient Descent



Mini-Batch Gradient Descent

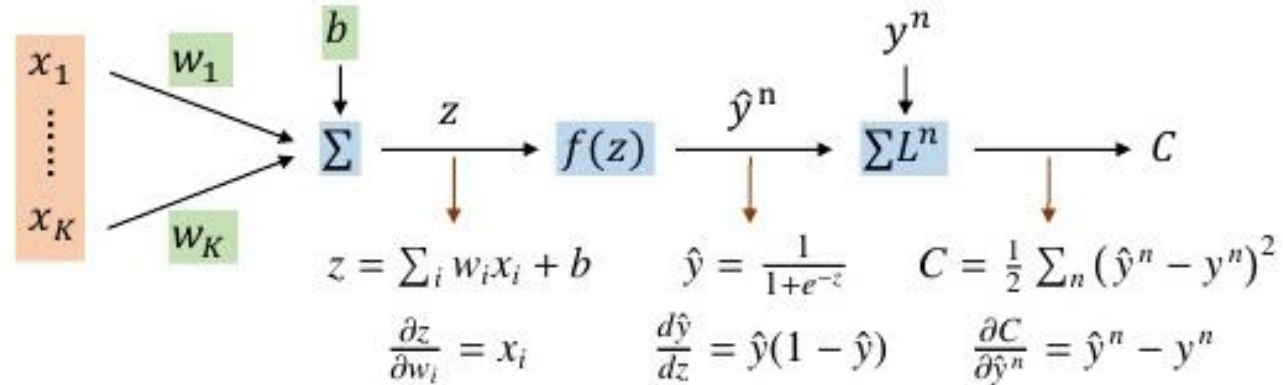


Stochastic Gradient Descent



4-5. Backpropagation

$$w_i(t+1) = w_i(t) - \eta \frac{\partial C}{\partial w_i}$$



$$\frac{\partial C}{\partial w_i} = \sum_{n=1}^N \frac{\partial z^n}{\partial w_i} \frac{d\hat{y}^n}{dz^n} \frac{\partial C}{\partial \hat{y}^n} = \sum_{n=1}^N x_i^n \hat{y}^n (1 - \hat{y}^n) (\hat{y}^n - y^n)$$

Chain rule

05 과제 설명

1주차 복습과제, 예습과제

4-1. 예습과제, 복습과제

1주차
복습과제1

Pytorch 기본 익히기 code

1주차
복습과제2

내용 복습 / 추론

1주차
예습과제1

텍스트 전처리 +
워드클라우드 code

1주차
예습과제2

Word2vec code



코드과제의 파일형식은 ipynb로, KUBIG 24-2 Github repo에 업로드 될 예정입니다!
Colab 환경에서 제작된 과제들이므로 **google colab**에서 실행하시는 것을 권장드립니다.

수고하셨습니다