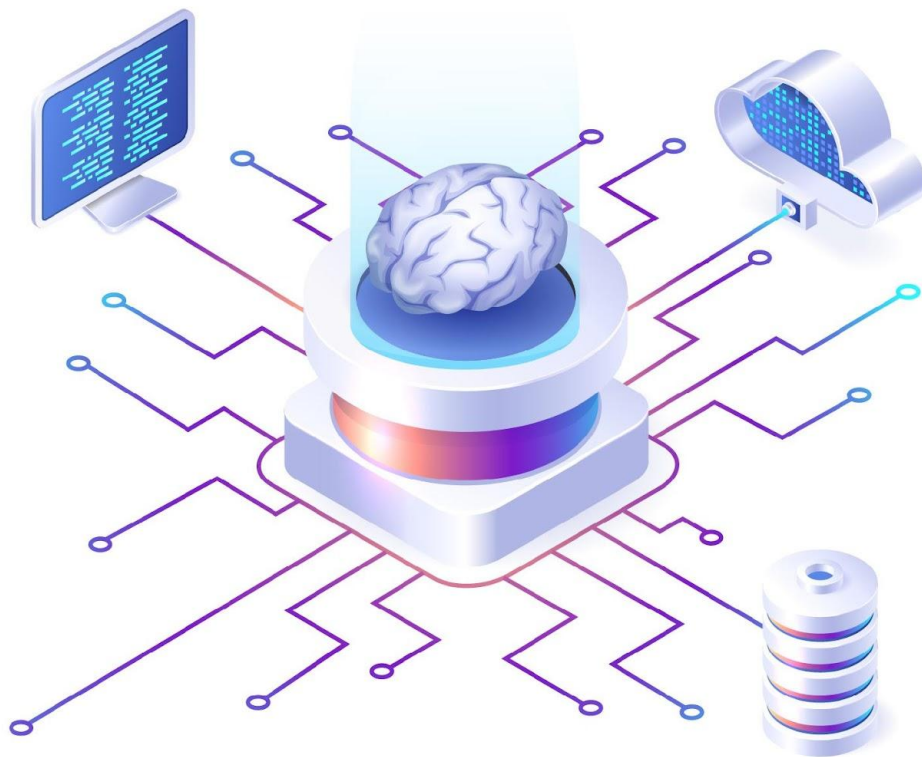


임베딩 (Embedding)

실무형 인공지능 자연어처리



임베딩 (Embedding)

딥러닝기반 자연어 처리

2

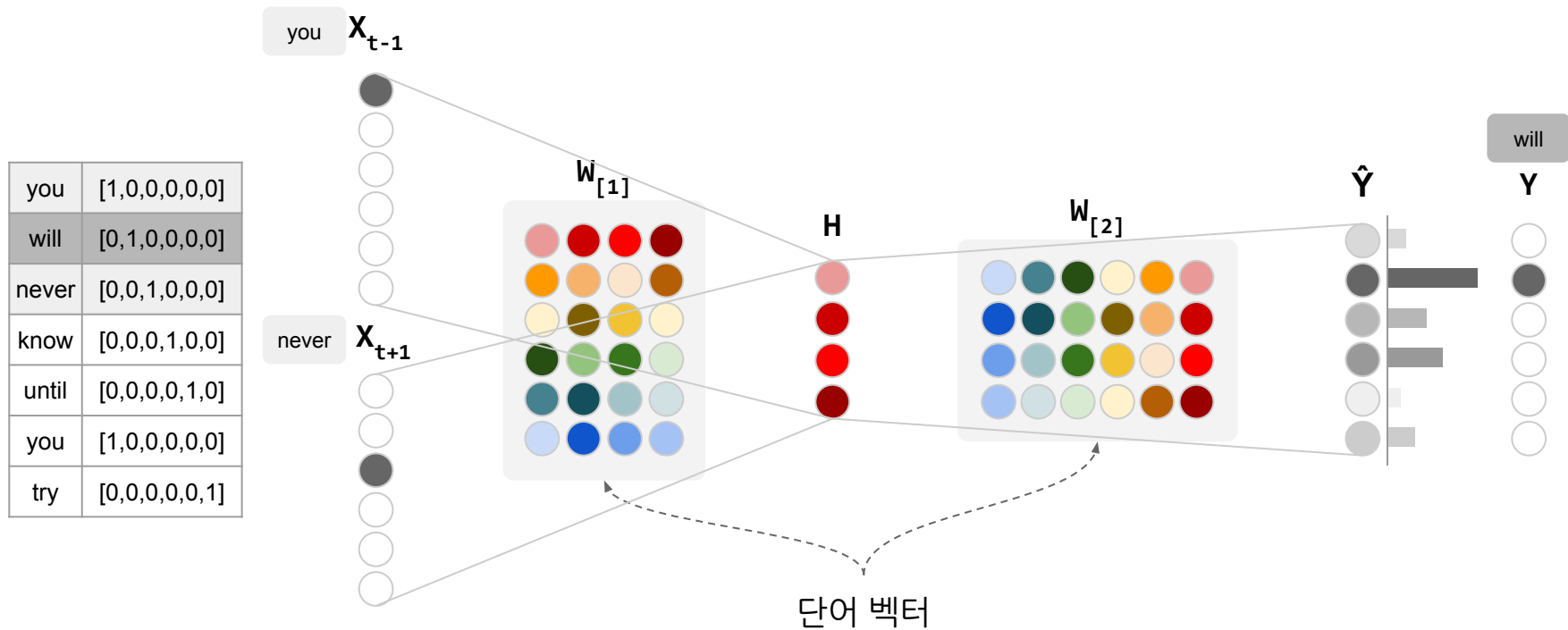
Word2Vec 직접 구현



Word2Vec(CBoW) 학습 - 구조

- 입력 문장 토큰화. (사용하지 않는 품사 제거)
- 가중치(= 파라미터 = 단어 벡터) 초기화
- 원핫 인코딩(one hot encoding)
- epoch 만큼 반복
 - 중심단어, 문맥단어 추출
 - Feed Forward
 - Loss / Gradient 계산
 - Weight (Parameter) 갱신

Word2Vec(CBoW) 학습 - 절차



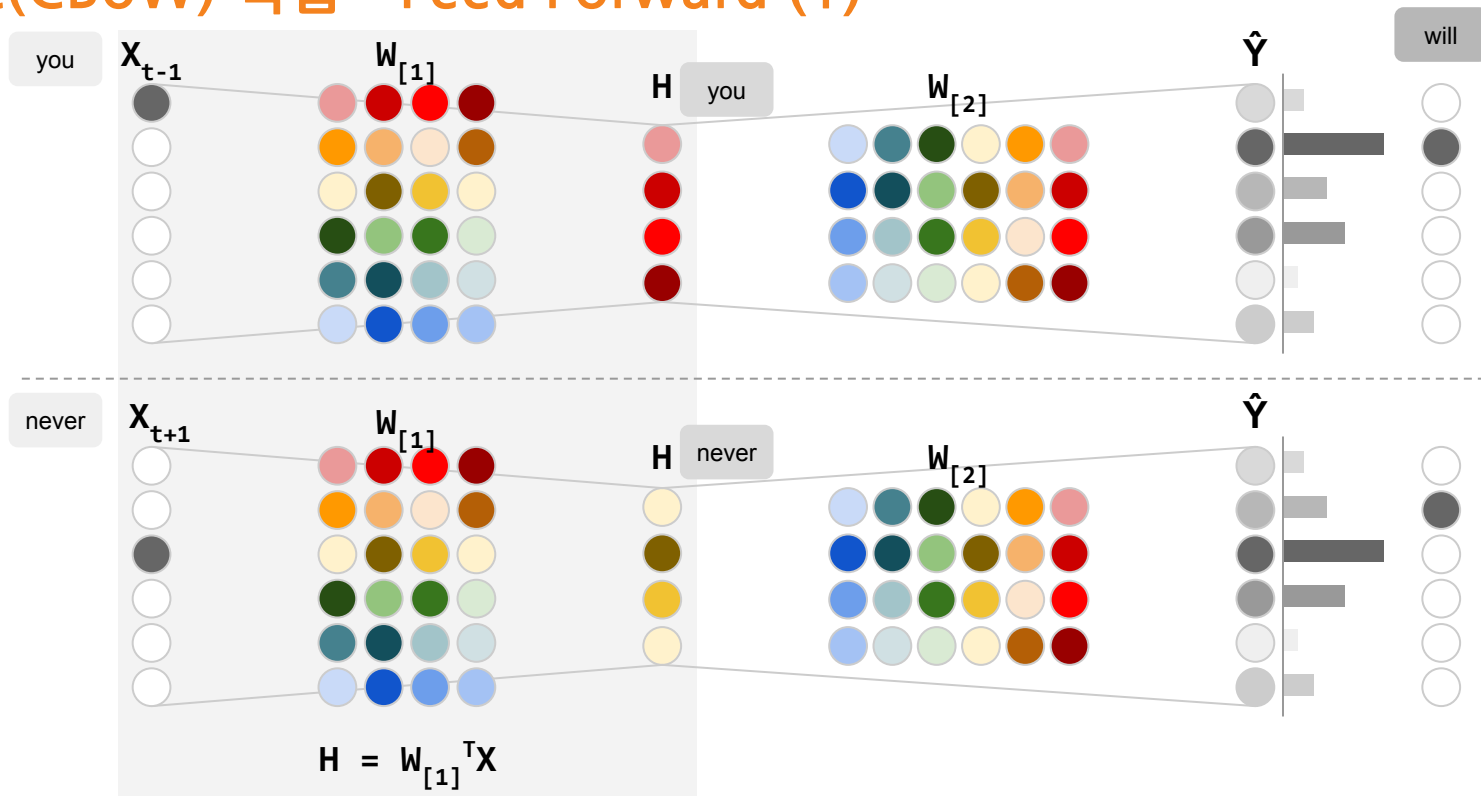
Word2Vec(CBoW) 학습 - 단어벡터



학습하는 가중치(파라미터)가 단어 벡터

Word2Vec(CBoW) 학습 - Feed Forward (1)

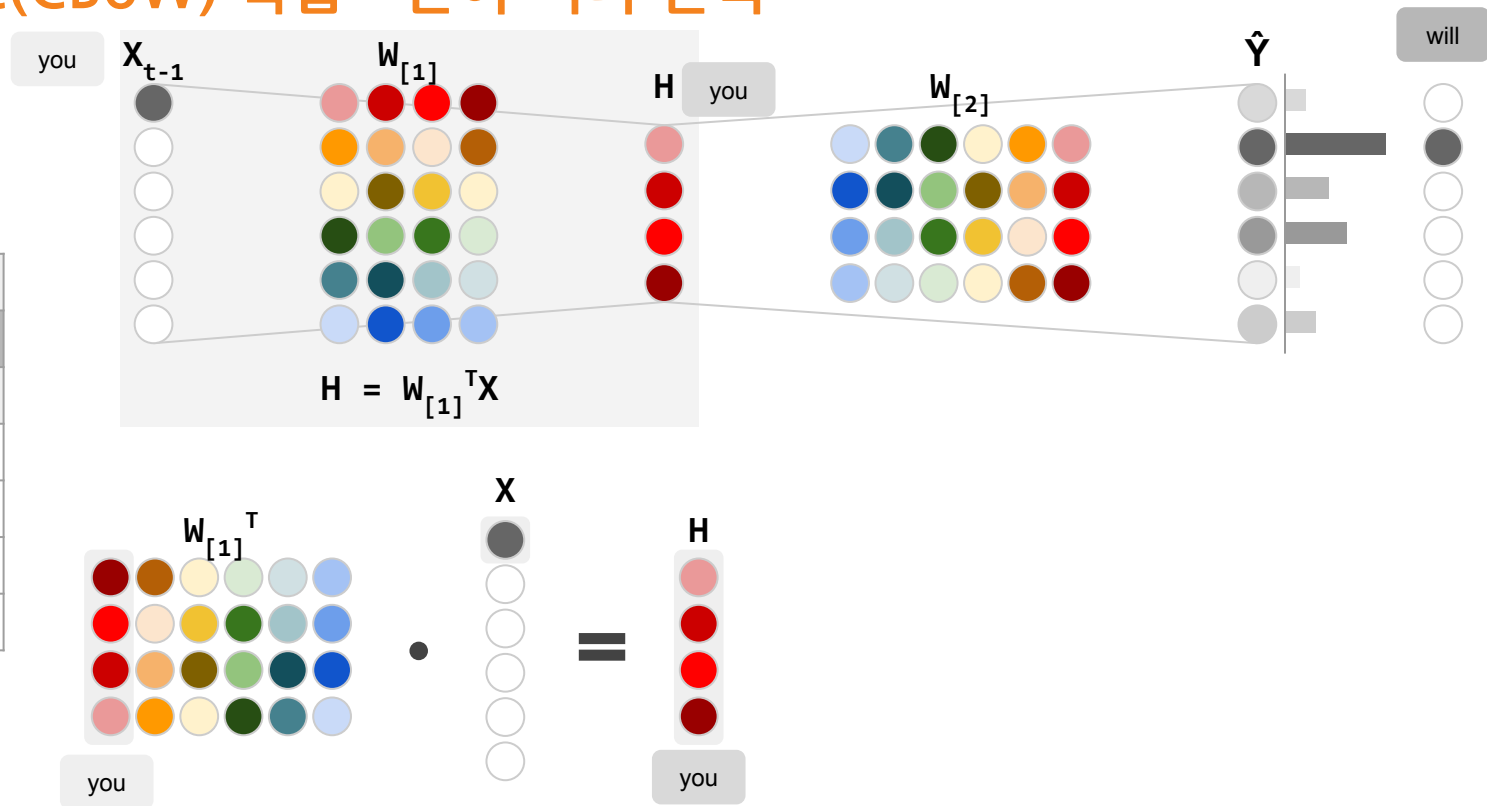
you	[1,0,0,0,0,0]
will	[0,1,0,0,0,0]
never	[0,0,1,0,0,0]
know	[0,0,0,1,0,0]
until	[0,0,0,0,1,0]
you	[1,0,0,0,0,0]
try	[0,0,0,0,0,1]



입력 단어(원핫 벡터)와 단어 벡터의 곱으로 Hidden layer 계산

Word2Vec(CBoW) 학습 - 단어 벡터 선택

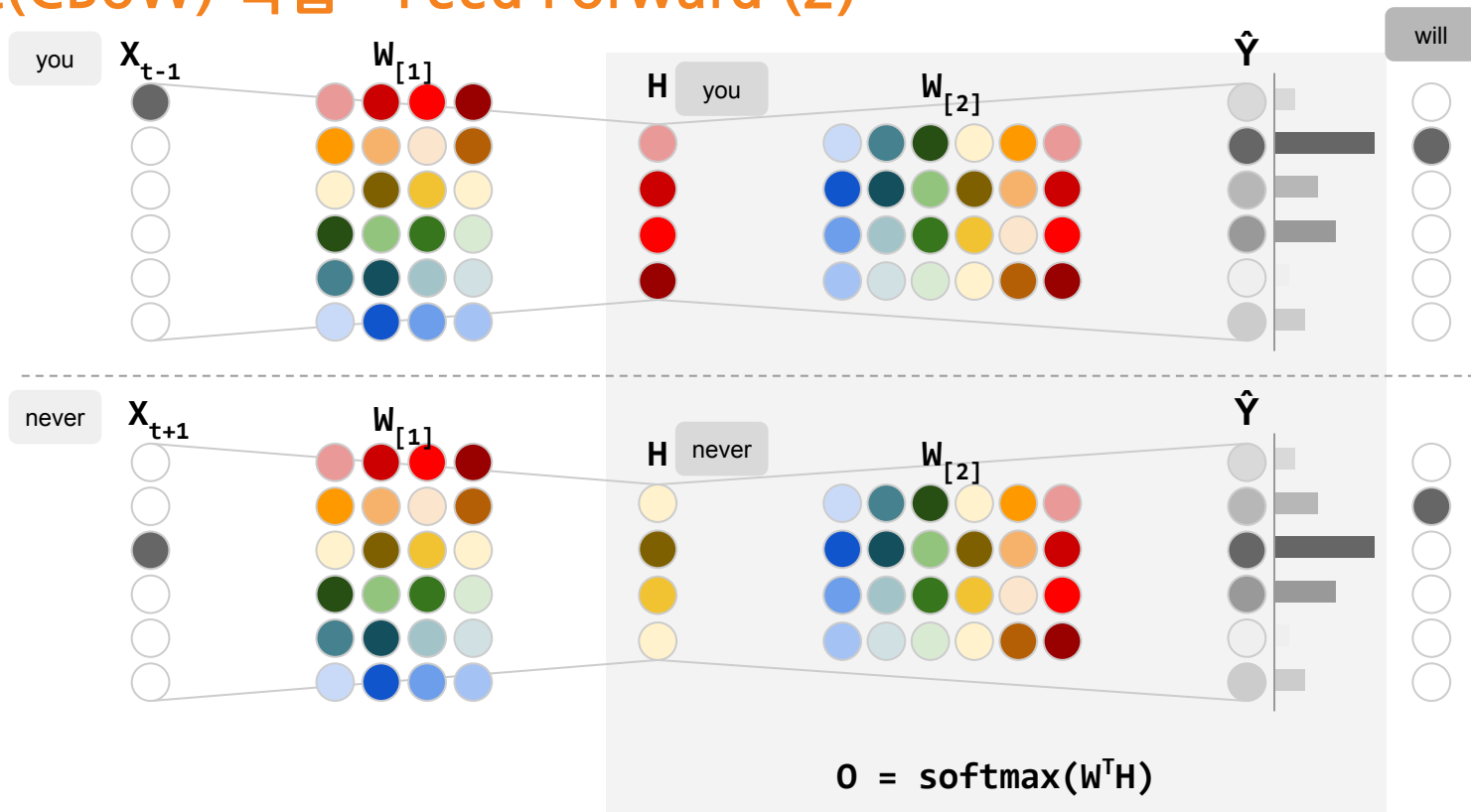
you	[1,0,0,0,0,0]
will	[0,1,0,0,0,0]
never	[0,0,1,0,0,0]
know	[0,0,0,1,0,0]
until	[0,0,0,0,1,0]
you	[1,0,0,0,0,0]
try	[0,0,0,0,0,1]



Hidden layer는 선택된 하나의 단어 벡터

Word2Vec(CBoW) 학습 - Feed Forward (2)

you	[1,0,0,0,0,0]
will	[0,1,0,0,0,0]
never	[0,0,1,0,0,0]
know	[0,0,0,1,0,0]
until	[0,0,0,0,1,0]
you	[1,0,0,0,0,0]
try	[0,0,0,0,0,1]



Hidden layer 와 단어벡터의 곱으로 Output layer 계산

Word2Vec(CBoW) 학습 - 단어간 유사도 계산

you	[1,0,0,0,0,0]
will	[0,1,0,0,0,0]
never	[0,0,1,0,0,0]
know	[0,0,0,1,0,0]
until	[0,0,0,0,1,0]
you	[1,0,0,0,0,0]
try	[0,0,0,0,0,1]



$$\text{similarity} = \frac{A \cdot B}{\|A\| \cdot \|B\|}$$

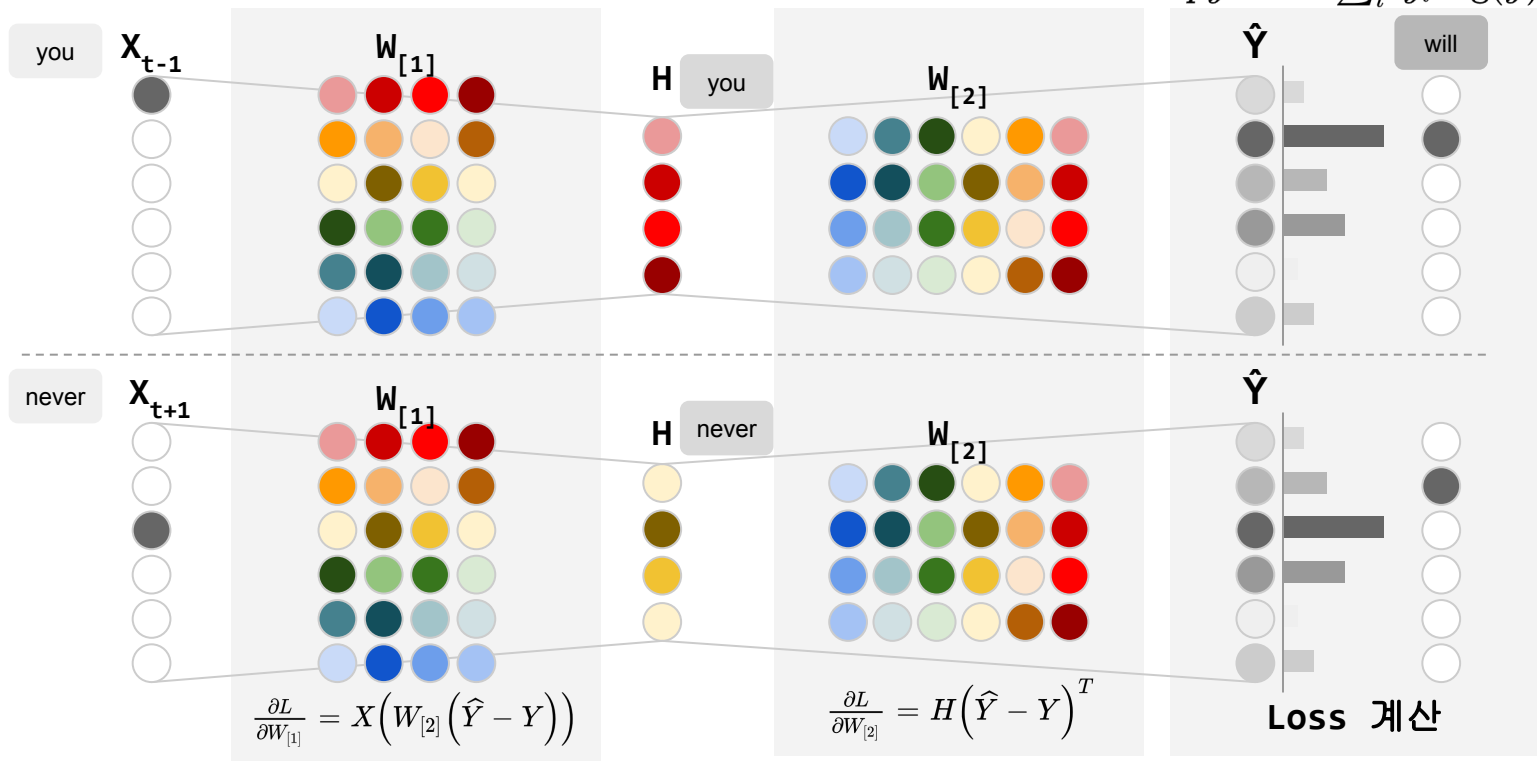
$$\text{softmax}(\text{selected word vector}(H) \cdot \text{all word vectors}(W_{[2]}^T)) =$$

선택된 단어 벡터(H)와 전체 단어 벡터($W_{[2]}^T$)간 유사도

Word2Vec(CBoW) 학습 - Loss / Gradient

$$CrossEntropy = - \sum_i^N y_i \log(\hat{y})$$

you	[1,0,0,0,0,0]
will	[0,1,0,0,0,0]
never	[0,0,1,0,0,0]
know	[0,0,0,1,0,0]
until	[0,0,0,0,1,0]
you	[1,0,0,0,0,0]
try	[0,0,0,0,0,1]



epoch 별 Loss / Gradient 계산

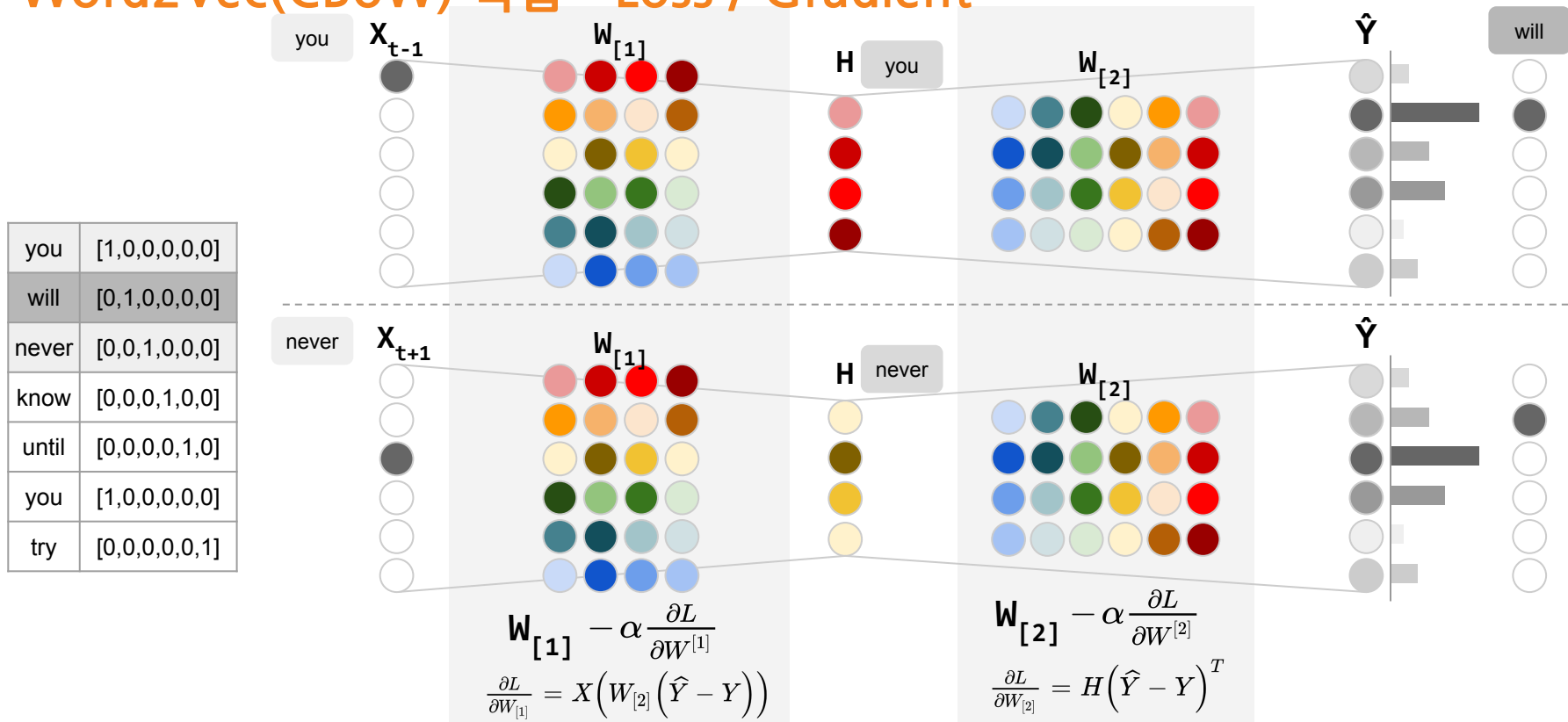
Word2Vec(CBoW) 학습 - Loss / Gradient

$$\textit{CrossEntropy} = - \sum_i^N y_i \log(\hat{y})$$



$$\textit{BinaryCrossEntropy} = - (y_i \log(\hat{y}) + (1 - y_i) \log(1 - \hat{y}))$$

Word2Vec(CBoW) 학습 - Loss / Gradient



Weight(Parameter)를 gradient * learning rate 만큼 갱신

감사합니다.

Insight⁺campus

