Summary

Ki Hyun Kim

nlp.with.deep.learning@gmail.com



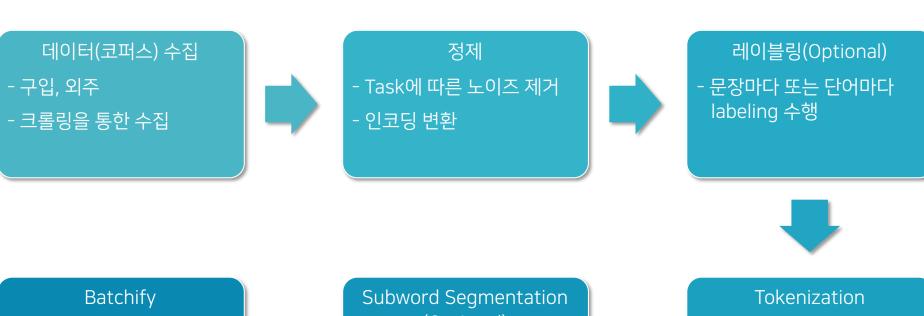
- 1) Introduction to Natural Language Processing
- 2) Preprocessing
- 3) Word Embedding
- 4) Text Classification



- 1) Introduction to Natural Language Processing
- 2) Preprocessing
- 3) Word Embedding
- 4) Text Classification



- 1) Introduction to Natural Language Processing
- 2) Preprocessing
- 3) Word Embedding
- 4) Text Classification



- 사전 생성 및 word2index 맵핑 수행
- 효율화를 위한 전/후처리



(Optional)

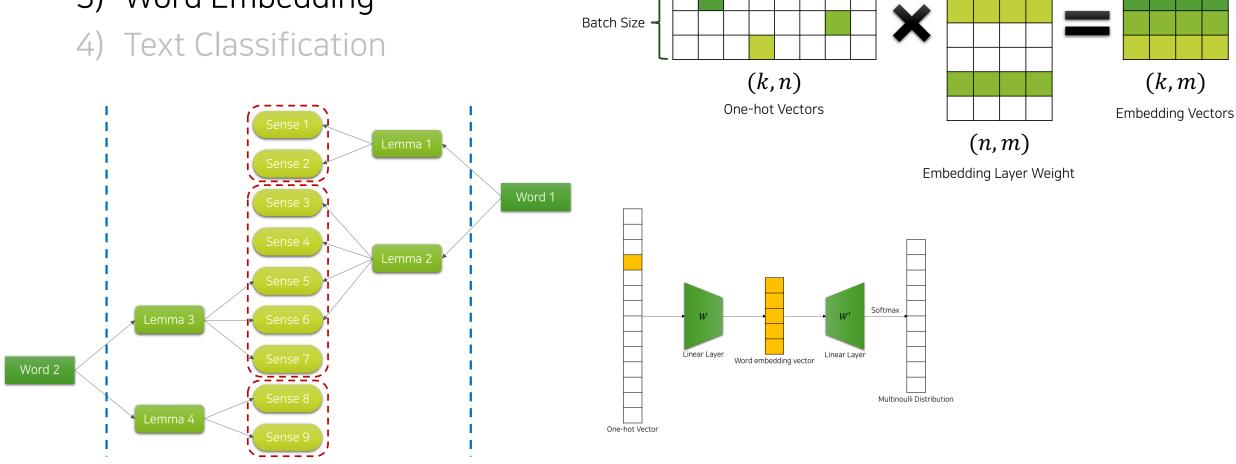
- 단어보다 더 작은 의미 단위 추가 분절 수행



- 형태소 분석기를 활용하여 분절 수행



- 1) Introduction to Natural Language Processing
- 2) Preprocessing
- 3) Word Embedding



Vocabulary Size = |V|



Embedding Vector Size

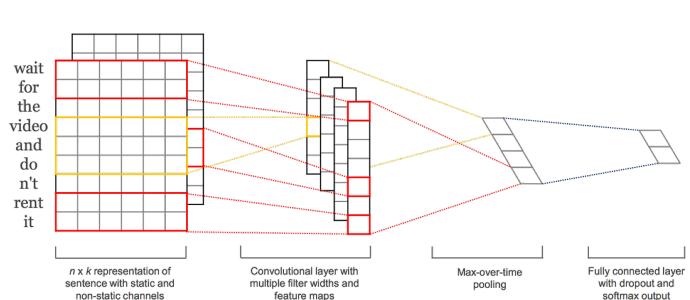
- 1) Introduction to Natural Language Processing
- 2) Preprocessing
- 3) Word Embedding
- 4) Text Classification

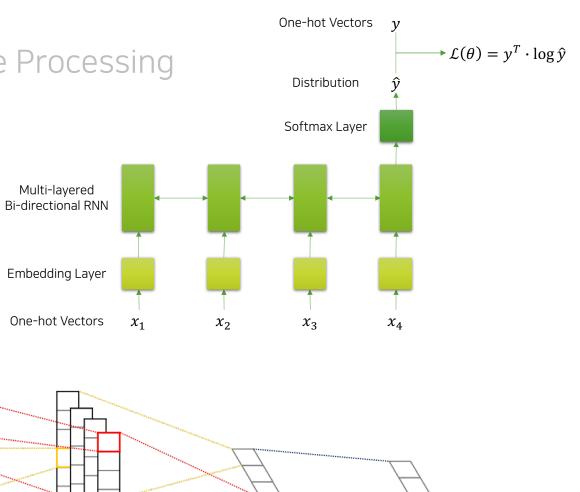
$$\hat{ heta} = rgmax_{ heta \in \Theta} \mathbb{E}_{ ext{x} \sim P(ext{x})} \Big[\mathbb{E}_{ ext{y} \sim P(ext{y}| ext{x})} ig[\log P(ext{y}| ext{x}; heta) ig] \Big]$$

$$\mathcal{D} = \{(x_i,y_i)\}_{i=1}^N$$

$$egin{aligned} \hat{ heta} &= rgmax_{ heta \in \Theta} \sum_{i=1}^N \log P(y_i|x_i; heta) \ &= rgmin_{ heta \in \Theta} - \sum_{i=1}^N \log P(y_i|x_i; heta) \end{aligned}$$

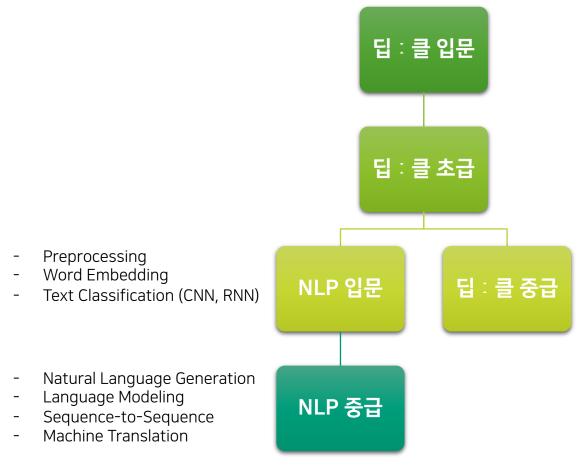
$$\mathcal{L}(heta) = -\sum_{i=1}^N \log P(y_i|x_i; heta)$$







강의 로드맵



- Linear Layer
- Loss Functions (e.g. MSE, BCE Loss)
- Backpropagation & Gradient Descent
- Regression & Classification
- Statistical Approach (e.g. Maximum Likelihood Estimation, MLE)
- Geometric Approach (e.g. Manifold Hypothesis)
- Convolutional / Recurrent Neural Networks (CNN, RNN)
- How to be a professional?
- Advanced Topics (e.g. Transfer Learning)
- Generative Learning (e.g. GAN, VAE)
- Anomaly Detection