

Summary

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In this class,

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



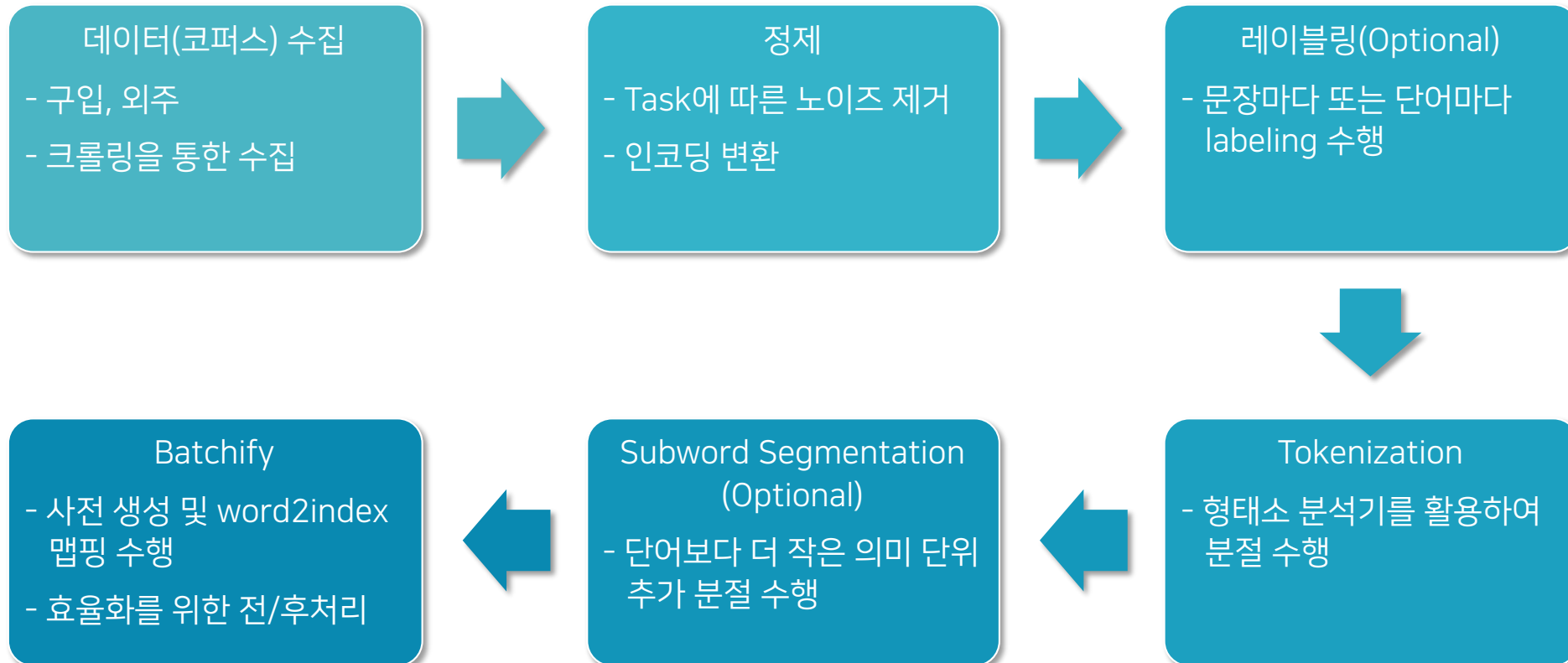
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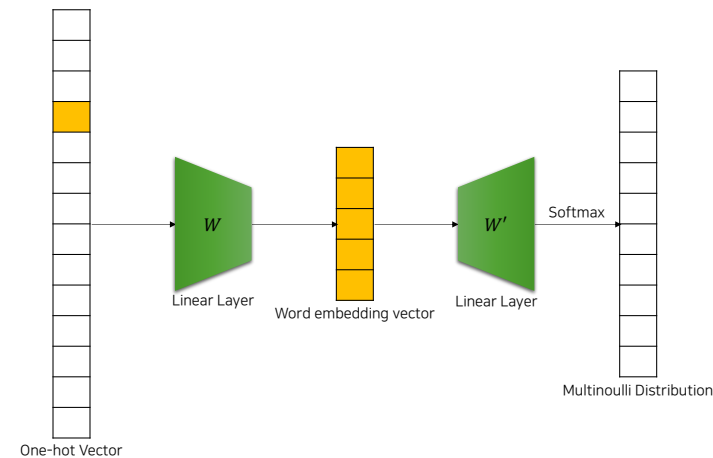
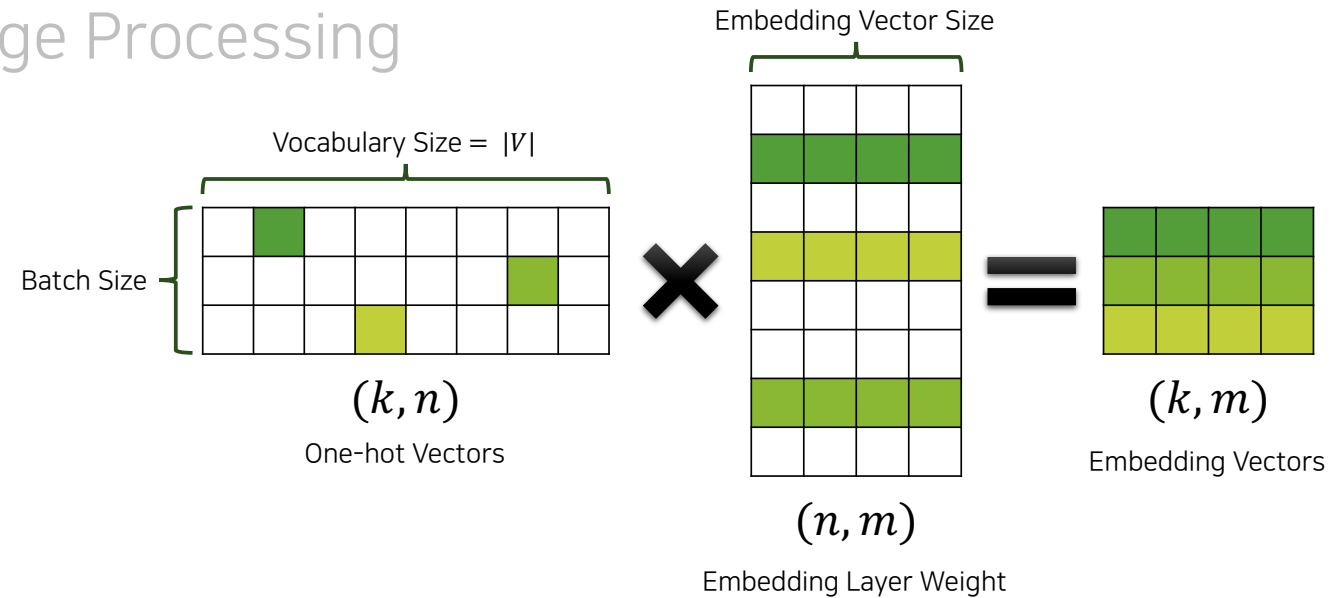
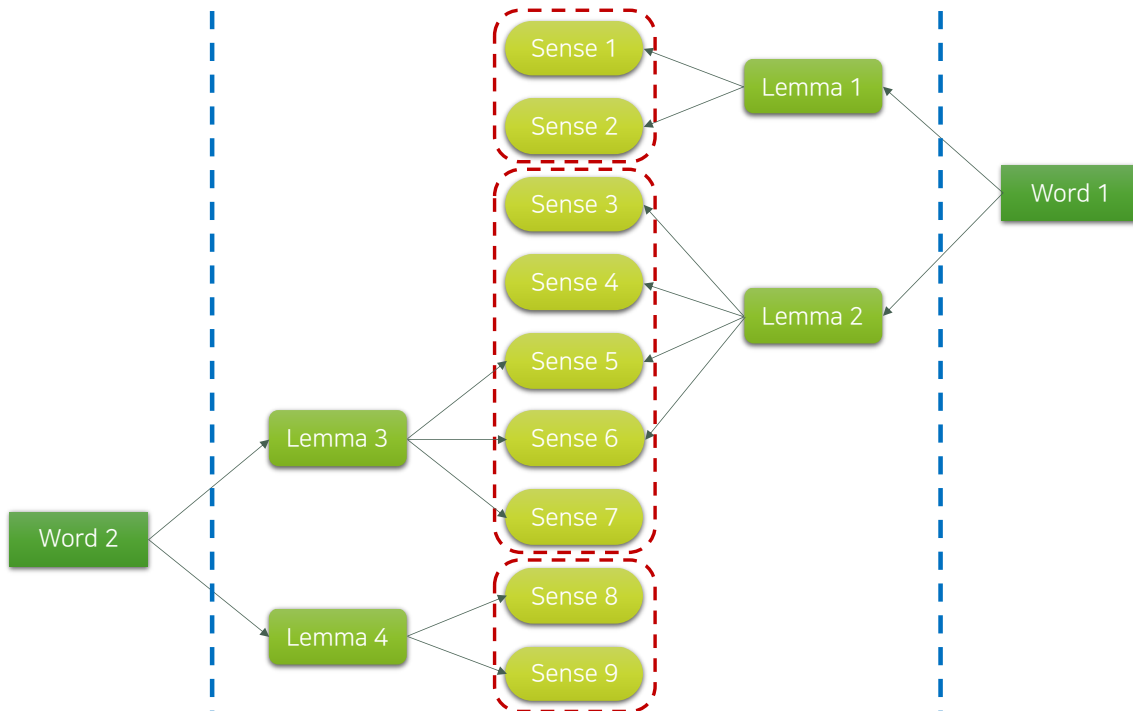
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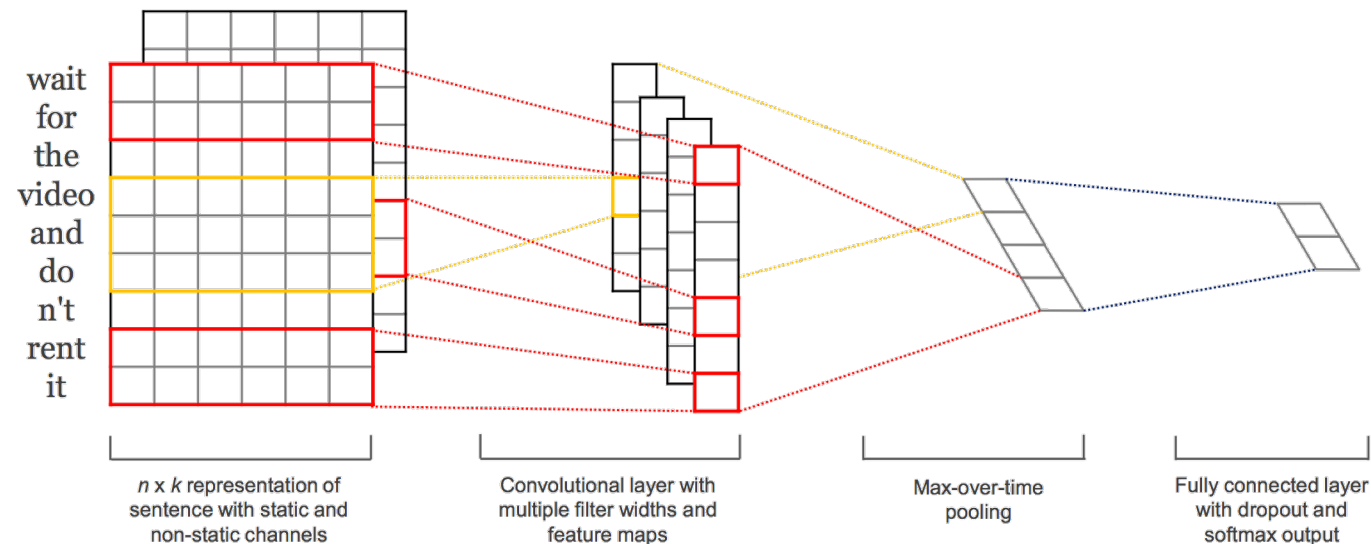
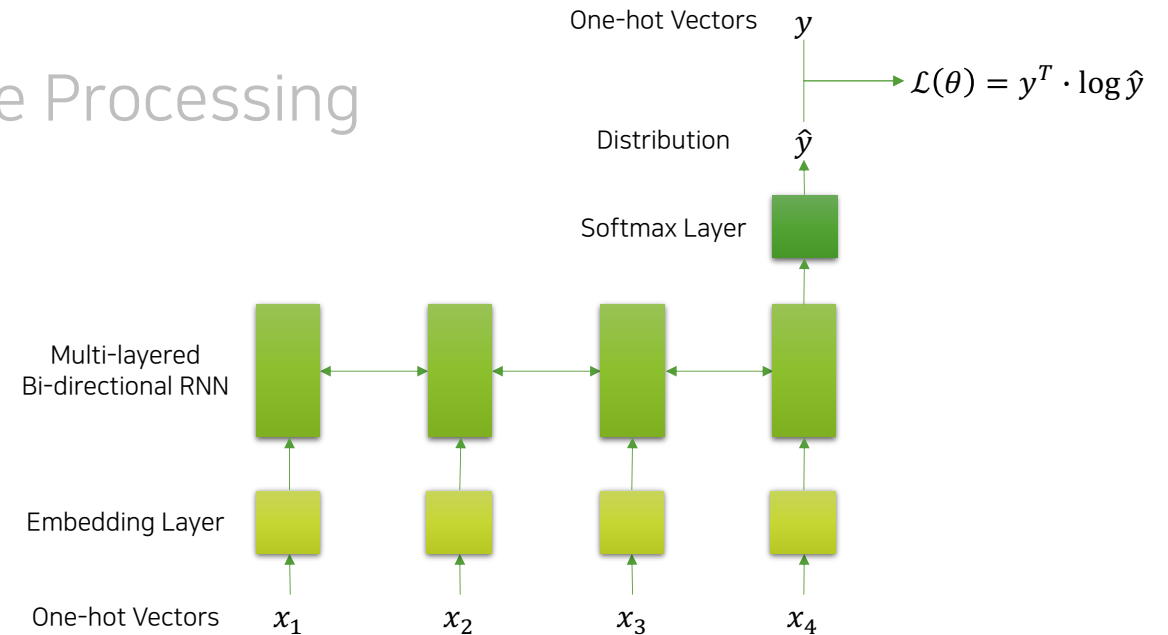
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$$\hat{\theta} = \operatorname{argmax}_{\theta \in \Theta} \mathbb{E}_{\mathbf{x} \sim P(\mathbf{x})} \left[\mathbb{E}_{\mathbf{y} \sim P(\mathbf{y}|\mathbf{x})} \left[\log P(\mathbf{y}|\mathbf{x}; \theta) \right] \right]$$

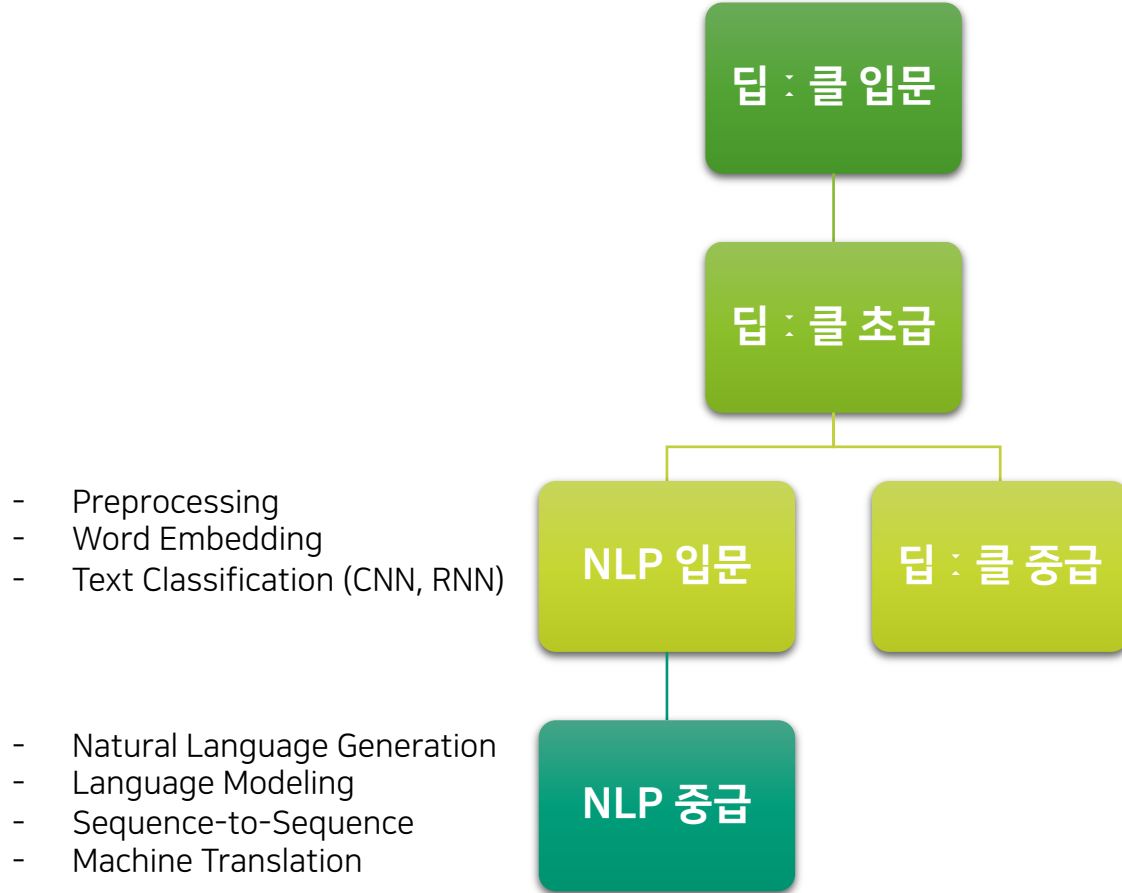
$$\mathcal{D} = \{(x_i, y_i)\}_{i=1}^N$$

$$\begin{aligned} \hat{\theta} &= \operatorname{argmax}_{\theta \in \Theta} \sum_{i=1}^N \log P(y_i | x_i; \theta) \\ &= \operatorname{argmin}_{\theta \in \Theta} - \sum_{i=1}^N \log P(y_i | x_i; \theta) \end{aligned}$$

$$\begin{aligned} \mathcal{L}(\theta) &= - \sum_{i=1}^N \log P(y_i | x_i; \theta) \\ \theta &\leftarrow \theta - \alpha \nabla_{\theta} \mathcal{L}(\theta) \end{aligned}$$



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