

# Language Model & Distributed Representation (3)

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### **Outlines**



- 1. NNLM
- 2. CBOW
- 3. Skip-gram
- 4. Hierarchical softmax
  - & Negative sampling
- 5. GloVe

- GloVe: Global Vectors for Word Representation
  - 1. Constructing co-ocurrence matrix from corpus

with decreasing weighting (The further the two words are, the less weight they have in the total count)

$$decay = 1/d$$

2. Constructing approximations between word vector and co-occurrence matrix

$$w_i^T ilde{w_j} + b_i + ilde{b_j} = \log(X_{ij})$$

| Probability and Ratio | k = solid            | k = gas            | k = water          | k = fashion         |
|-----------------------|----------------------|--------------------|--------------------|---------------------|
| P(k ice)              | $1.9 \times 10^{-4}$ | $6.6\times10^{-5}$ | $3.0\times10^{-3}$ | $1.7\times10^{-5}$  |
| P(k steam)            | $2.2 \times 10^{-5}$ | $7.8\times10^{-4}$ | $2.2\times10^{-3}$ | $1.8\times 10^{-5}$ |
| P(k ice)/P(k steam)   | 8.9                  | $8.5\times10^{-2}$ | 1.36               | 0.96                |

Pennington J, Socher R, Manning C D. Glove: Global vectors for word representation[C]//Proceedings of the 2014 conference on empirical methods in natural language processing (EMNLP). 2014: 1532-1543.

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3. Constructing Loss function and training:

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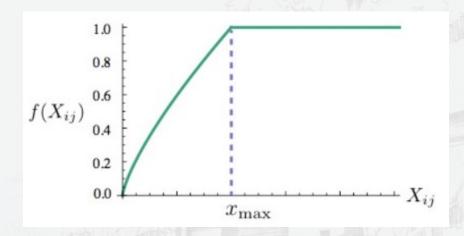
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$$J = \sum_{i,j=1}^V f(X_{ij}) (w_i^T ilde{w_j} + b_i + ilde{b_j} - \log(X_{ij}))^2$$

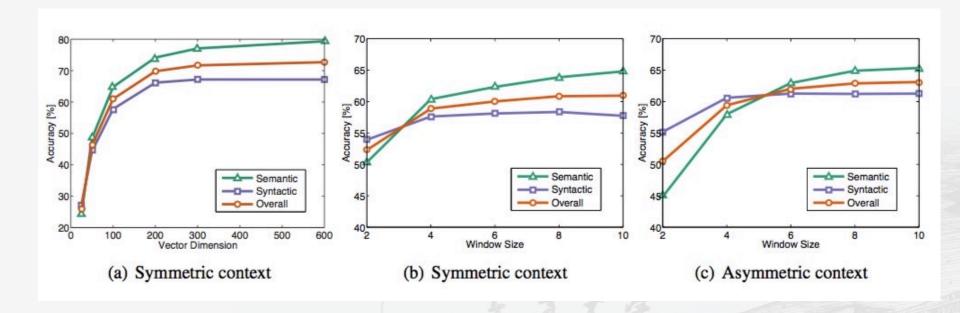
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where 
$$f(X_{ij}) = \begin{cases} (x/x_{max})^{\alpha} & \text{if } x < x_{max} \\ 1 & \text{otherwise} \end{cases}$$



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- 共采用了三个指标: 语义准确度, 语法准确度以及总体准确度;
- Vector Dimension在300时能达到最佳;
- · 而context Windows size大致在6到10之间。

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## Q & A

