# Lecture 3: Advanced Word2Vec Methods

2018年7月4日 星期三 下午11:58

#### **SGD**

- Very Sparse: #words in window = 2m+1
- Only have to update Specific vectors

## 1. Negative sampling in skip-gram:

#### **Loss Function:**

$$J_t(\theta) = \log \sigma(u_o^T v_c) + \sum_{j \sim P(w)} [\log \sigma(-u_j^T v_c)]$$

- Randomly choose some words, reduce computation
- Maximize the co-occurrence of real context words
- Minimize the probability of random words (negative samples)

## 2. Window based co-occurrence matrix

counts	1	like	enjoy	deep	learning	NLP	flying	
1	0	2	1	0	0	0	0	0
like	2	0	0	1	0	1	0	0
enjoy	1	0	0	0	0	0	1	0
deep	0	1	0	0	1	0	0	0
learning	0	0	0	1	0	0	0	1
NLP	0	1	0	0	0	0	0	1
flying	0	0	1	0	0	0	0	1
	0	0	0	0	1	1	1	0

- Weakness:
  - Very large dimension
  - Word meaning is not fixed
- Solution: SVD
  - Advantage:
    - Best Rank-k approximation
    - Fast training in small corpus
  - Disadvantage:
    - Time cost O( $mn^{\uparrow}2$ ), restricted in small data
    - Hard to incorporate new words
    - Without information about word similarity
    - Extremely large weight on large counts (the\a\is)

## Interesting patterns

- (Teach Teacher); (Swim Swimmer) seems to have similar distance & direction
- Grammar induction
- Words with the same root cluster together.

#### 3.GloVe:

Combine the best of **Count based** and **Direct prediction** model.

$$J(\theta) = \frac{1}{2} \sum_{i,j=1}^{W} f(P_{ij}) (u_i^T v_j - \log P_{ij})^2$$

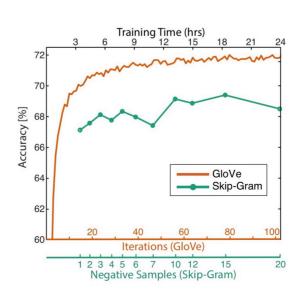
Where:  $\begin{cases} P_{ij} = \frac{x_{ij}}{x_{i}} = \frac{x_{ij}}{\sum_{k} x_{ik}} \\ f(x) = \max(P_{ij}, 1) \end{cases}$  But Frid wett.

让<(Uì,Vj>逼压(ogPi) 其核心是否仍是玩计模型? 引入 J(0) 可用SGD to 連, 然而 它仍需计算其现矩阵, 且无法解决 加入新语料后重新计算 Pi Sii)题... Not Scalable?

随机平梯,是否会平列上下之间语成为中心间很相关的词?

### 4. Evaluation of word vector:

- Intrinsic:
  - o Intermediate subtask
  - What parameter give improvement (change one at a time)
  - o Better understanding the system
  - Word Vector Analogies
    - Semantic: Queen King: Man Woman
    - Syntactic: Short shorter shortest: large larger largest
    - Paris France + Italy = Rome
- Extrinsic:
  - Apply in real task.
  - Time Consuming





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可容贵的我们在什么情略下使用何种证的为此?