

Word Embeddings

- 1-hot representation

- word embedding

e.g.

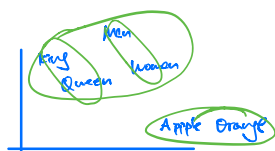
	Man (53911)	Woman (9853)	King (4914)	Queen (7157)
Gender	-1	1	-0.95	0.97
Royal	0.01	0.02	0.93	0.95
Age	203	0.02	0.7	0.69
...				

300

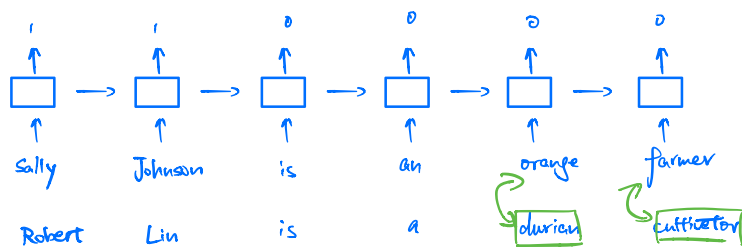
↑
↓

e_{53911} e_{9853}

300D → 2D Visualization



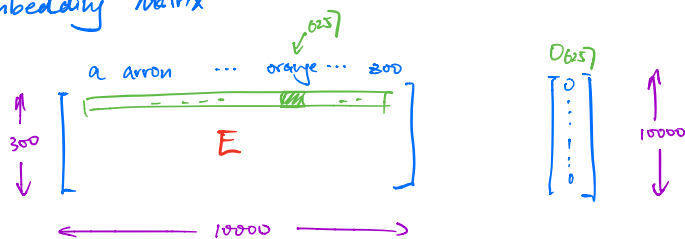
NER example



⇒ transfer learning

1. learn word embeddings from large text corpus (1-100B words)
(or download pre-trained embedding)
2. transfer embedding to new task w/ smaller training set
(say 100 k words)
3. Optional. Fine-tune word embeddings w/ new data

Embedding Matrix



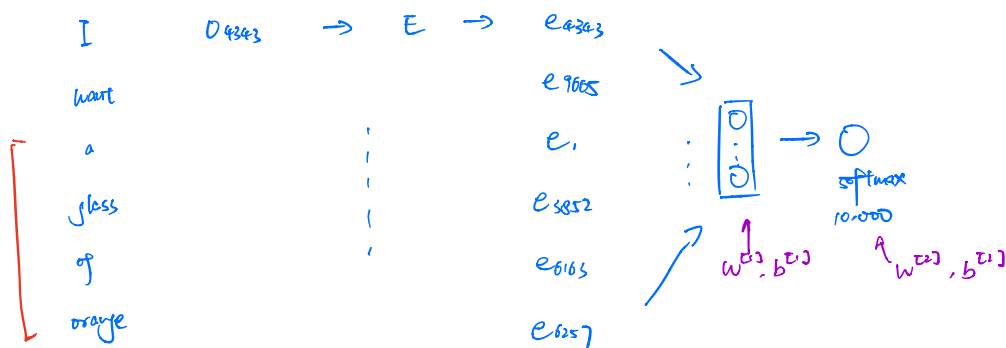
$$E \cdot D_{625} = \begin{bmatrix} 12 \\ \vdots \\ 0 \end{bmatrix} = e_{625}$$

$(300, 10K) \quad (10K, 1) \quad (300, 1)$

Learning word embeddings

e.g. I want a glass of orange — .

4343 9665 1 3852 6163 6257



$300 \times 6 = 1800$ → 1200 window

backprop to learn E

- Context : last 4 words
- 4 words on left & right
- last 1 word
- nearby 1 word

Word2Vec

Skip-gram

e.g. I want a glass of orange juice to go along my cereal.

<u>Context</u>	<u>target</u>	← sampled from nearby window
orange	juice	
orange	glass	
orange	my	

Model:

Vocab size = 10,000 K

Context c ("orange") → Target t ("juice")
6257 4834

$O_c \rightarrow \textcircled{E} \rightarrow e_c \rightarrow \text{softmax} \rightarrow \hat{y}$ ← one-hot
 $e_c = E O_c$

Softmax: $P(t|c) = \frac{e^{\theta_t^T e_c}}{\sum_{j=1}^{10000} e^{\theta_j^T e_c}}$ θ_t = param associated w/ output t

$$\mathcal{L}(\hat{y}, y) = - \sum_{i=1}^{10000} y_i \log \hat{y}_i$$

Problem:

Calc denominator of softmax is slow

Solution:

hierarchical softmax

negative sampling

Negative Sampling

e.g. I want a glass of orange juice to go along my cereal.

x		y
Context	word	target ?
orange	juice	1
orange	king	0
orange	book	0
orange	the	0

$k = 5 - 20$ smaller dataset

$k = 2 - 5$ larger dataset

softmax:
$$P(t|c) = \frac{e^{\theta_t^T e_c}}{\sum_{j=1}^{10000} e^{\theta_j^T e_c}}$$
 } 10000-way softmax

$\Rightarrow P(y=1 | c, t) = \sigma(\theta_t^T e_c)$



much faster
10,000 logistic regression

Select negative sampling

$$P(w_i) = \frac{f(w_i)^{3/4}}{\sum_{j=1}^{10000} f(w_j)^{3/4}}$$

between uniform distribution & word distribution

GloVe word vectors

e.g. I want a glass of orange juice to go along my cereal.

c, t

$x_{ij} = \# \text{ times } j \text{ appears in context of } i$

$x_{ij} = x_{ji}$

Symmetric $\Rightarrow e_w^{\text{final}} = \frac{e_w + \theta_w}{2}$

Model:

minimize
$$\sum_{i=1}^{10000} \sum_{j=1}^{10000} f(x_{ij}) (\theta_i^T e_j + b_i + b_j' - \log x_{ij})^2$$

weighting term $f(x_{ij}) = 0$ if $x_{ij} = 0$