Embeddings (1)

Neural Machine Translation

One-Hot-Encoding

```
sen1 = 'Deep Learning is interesting for science and economy.'
sen2 = 'The media claims Deep Learning is interesting.'

11 words -> 11 dimensional vectors
( and, claims, Deep, ecomomy, for, interesting, is, Learning, media, science, The)
'Deep' = (0, 0, 1, 0, 0, 0, 0, 0, 0, 0)
```

Bag-of-Words

```
sen1 = 'Deep Learning is interesting for science and economy.'
sen2 = 'The media claims Deep Learning is interesting.'

BOW(sen1) = (1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0)

BOW(sen2) = (0, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1)

D = sen1 + sen2

BOW(D) = (1, 1, 2, 1, 1, 2, 2, 2, 1, 1, 1)
```

Counter-Based

```
• TF tf(t,d) = f_{t,d}/N \qquad \qquad N: Number\ of\ words\ in\ d  idf(t,D) = \log(\#D/\#d_t) \qquad \#D: Number\ of\ Documents \\ \#d_t: Number\ of\ Docs \\ including\ term\ t
```

```
sen1 = 'Deep Learning is interesting for science and economy.'
sen2 = 'The media claims Deep Learning is interesting.'
```

```
TF('Deep', sen1) = 1/8

TF('Deep', sen2) = 1/7

IDF('Deep', D) = log(2/2) = 0
```

Counter-Based

```
 \begin{array}{ll} \bullet & {\sf TF} & tf(t,d) = f_{t,d}/N & N: {\it Number of words in d} \\ \bullet & {\sf TF-IDF} & idf(t,D) = \log(\#D/\#d_t) & \#D: {\it Number of Documents} \\ & \#d_t: {\it Number of Docs} \\ & including term\ t \\ \\ {\sf sen1} = {\it 'Deep Learning is interesting for science and economy.'} \end{array}
```

```
TF('science', sen1) = 1/8

TF('science', sen2) = 0

IDF('science', D) = log(2/1) = 0.301 -> TF-IDF('science', sen2, D)=0.037
```

sen2 = 'The media claims Deep Learning is interesting.'

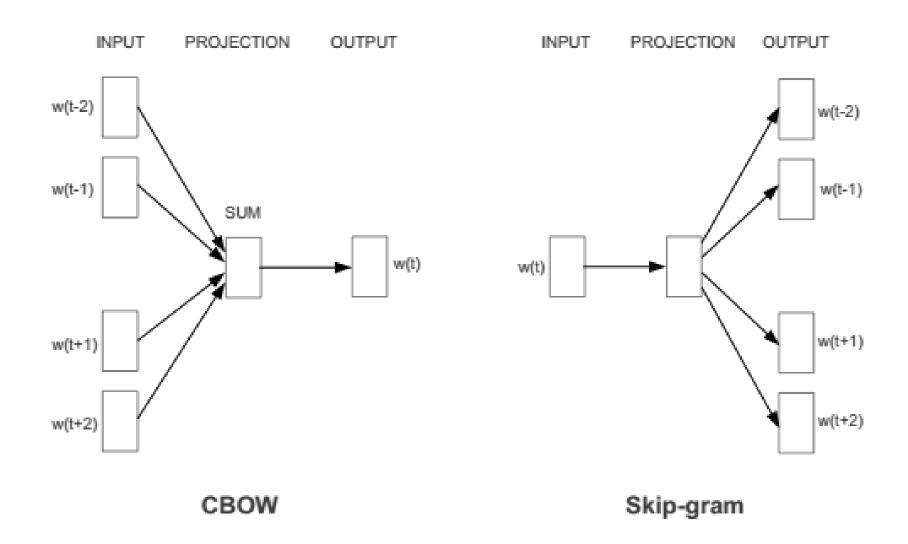
Word2Vec

• CBOW:

Predict current word from bag-of-words of surrounding words

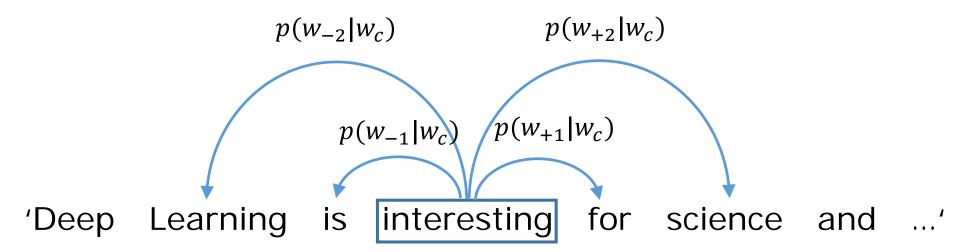
• Skip-Gram:

Predict context words given the current word



Word2Vec

• Minimize: $\log \sigma(u_o^T v_c) + \sum [\log \sigma(-u_j^T v_c)]$



Glove

• Minimize: $\sum_{i,j=1}^{W} f(count(i,j)) (u_i^T v_j - \log count(i,j))^2$

'Deep Learning is interesting for science and ...'