Introduction to Word2Vec

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t.me/cvision

Outline

- Word Vectors
- Word2Vec



http://rohanvarma.me/Word2Vec/

http://mccormickml.com/2016/04/19/word2vec-tutorial-the-skip-gram-model/

http://mccormickml.com/2017/01/11/word2vec-tutorial-part-2-negative-sampling/

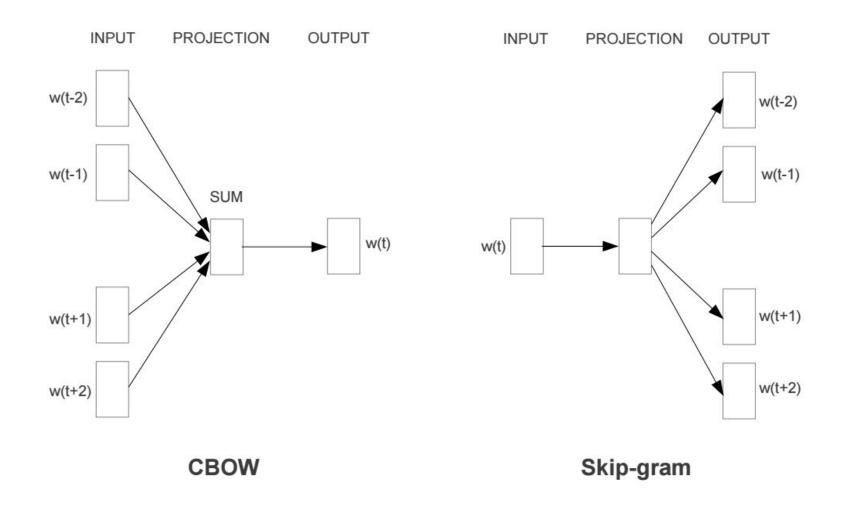
http://ruder.io/word-embeddings-1/

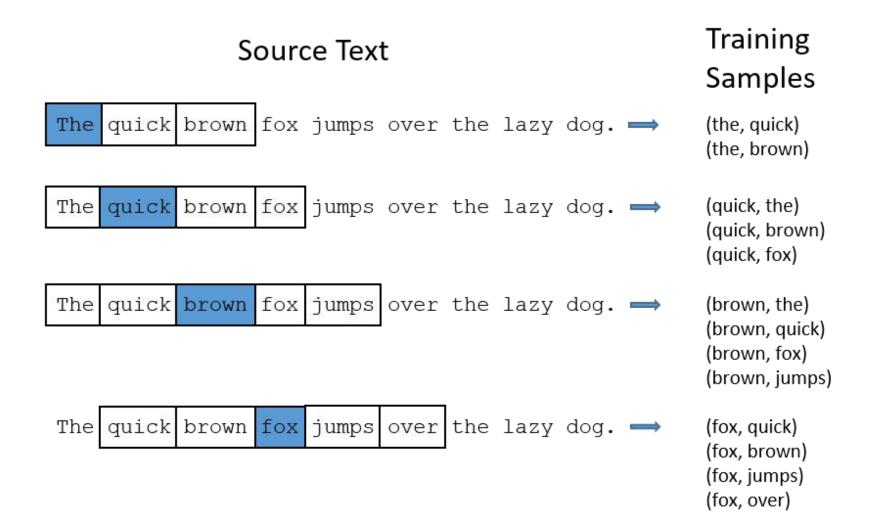
Word Vectors

- Word Vectors, Word Embeddings or Distributed
 Representation of Words, generally refer to a Dense
 Vector representation of a word, as compared to a sparse (ie
 one-hot) traditional representation
- Word2Vec is one of Distributed Representation of Words methods

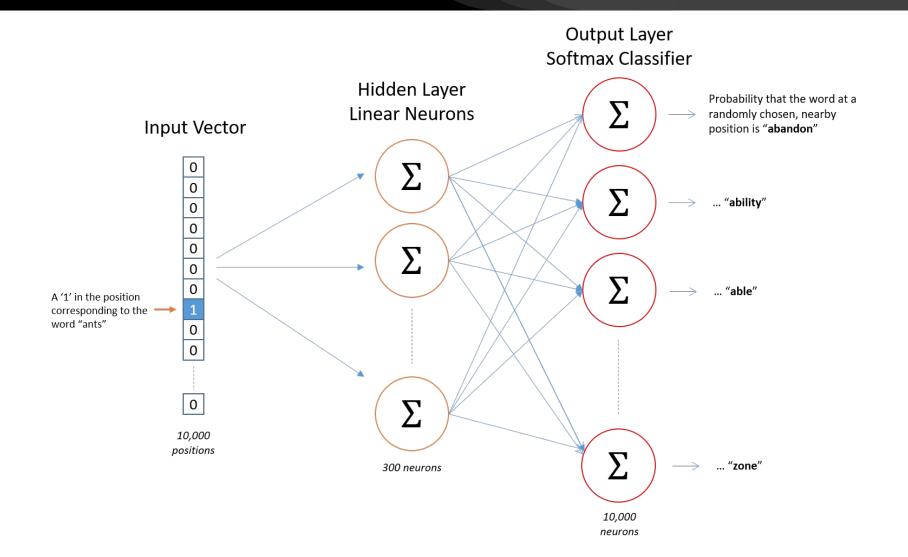
Word2Vec

- Represent words as dense vectors
- Vectors feed into a discriminative model (typically an RNN)
- There are two different implementations of Word2Vec
 - Skip-Gram (SG)
 - Continuous Bag of Words (CBoW)
- Both of these models learn dense vector representation of words, based on the words that surround them (ie, their context)
- Skip-Gram model predicts context (surrounding) words given the current word
- Continuous Bag of Words model predicts the current word based on several surrounding words



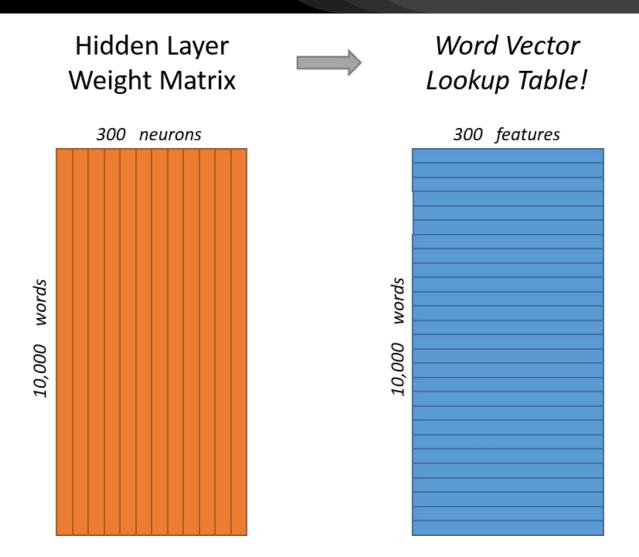


- Word2Vec is a simple neural network with a single hidden layer
- This network train with raw text for specific task (language model), but the goal is actually just to learn the weights of the hidden layer
- The weights of hidden layer are actually the Word Vectors
- For learn Word Vectors with Word2Vec
 - Prepare training documents
 - Build a vocabulary of words from training documents (V unique words)
 - Represent each word as one hot vector (as input)
 - The output of the network have a same size as input (V)



• Hidden layer of this model is really just operating as a lookup table. The output of the hidden layer is just the "word vector" for the input word

$$\begin{bmatrix} 0 & 0 & 0 & 1 & 0 \end{bmatrix} \times \begin{bmatrix} 17 & 24 & 1 \\ 23 & 5 & 7 \\ 4 & 6 & 13 \\ \hline 10 & 12 & 19 \\ 11 & 18 & 25 \end{bmatrix} = \begin{bmatrix} 10 & 12 & 19 \end{bmatrix}$$



- If two different words have very similar "contexts" (that is, what words are likely to appear around them synonyms like "intelligent" and "smart" would have very similar contexts), the model needs to output very similar results for these two word.
- One way for the network to output similar context predictions for these two words is if the word vectors are similar.
- So, if two words have similar contexts, then the network is motivated to learn similar word vectors for these two words!
- Ta da!

