

C5W2-Quiz-NLP-Word-Embedding

1. Suppose you learn a word embedding for a vocabulary of 10000 words. Then the embedding vectors should be 10000 dimensional, so as to capture the full range of variation and meaning in those words.

☐ True

☒ False

2. What is t-SNE?

☒ A non-linear dimensionality reduction technique

☐ An open-source sequence modeling library

☐ A linear transformation that allows us to solve analogies on word vectors

☐ A supervised learning algorithm for learning word embeddings

3. Suppose you download a pre-trained word embedding which has been trained on a huge corpus of text. You then use this word embedding to train an RNN for a language task of recognizing if someone is happy from a short snippet of text, using a small training set.

Aa x (input text)	# y (happy?)
I'm feeling wonderful today!	1
I'm bummed my cat is ill.	0
Really enjoying this!	1

Then even if the word “ecstatic” does not appear in your small training set, your RNN might reasonably be expected to recognize “I’m ecstatic” as deserving a label=1.

☒ True

☐ False

4. Which of these equations do you think should hold for a good word embedding? (Check all that apply)

☐ $e_{boy}-e_{girl} \approx e_{sister}-e_{brother}$

☐ $e_{boy}-e_{brother} \approx e_{sister}-e_{girl}$

☒ ~~$e_{boy}-e_{girl} \approx e_{brother}-e_{sister}$~~

☒ ~~$e_{boy}-e_{brother} \approx e_{girl}-e_{sister}$~~

5. Let E be an embedding matrix, and let o_{1234} be a one-hot vector corresponding to word 1234. Then to get the embedding of word 1234, why don't we call $E * o_{1234}$ in Python?

☐ The correct formula is $ET * o_{1234}$.

☐ This doesn't handle unknown words (<UNK>).

☐ None of the above: calling the Python snippet as described above is fine.

☒ ~~It is computationally wasteful.~~

6. When learning word embeddings, we create an artificial task of estimating $P(\text{target} \mid \text{context})$. It is okay if we do poorly on this artificial prediction task; the more important by-product of this task is that we learn a useful set of word embeddings.

☒ ~~True~~

☐ False

7. In the word2vec algorithm, you estimate $P(t \mid c)$, where t is the target word and c is a context word. How are t and c chosen from the training set? Pick the best answer.

☐ c is the one word that comes immediately before t .

☐ c is the sequence of all the words in the sentence before t .

☒ ~~c and t are chosen to be nearby words.~~

☐ c is a sequence of several words immediately before t .

8. Suppose you have a 10000 word vocabulary, and are learning 500-dimensional word embeddings. The word2vec model uses the following softmax function:

$$P(t \mid c) = \frac{e^{\theta_t^T e_c}}{\sum_{t'=1}^{10000} e^{\theta_{t'}^T e_c}}$$

Which of these statements are correct? Check all that apply.

- ☒ ~~θ_t and e_c are both trained with an optimization algorithm such as Adam or gradient descent.~~
- ☐ After training, we should expect θ_t to be very close to e_c when t and c are the same word.
- ☒ ~~θ_t and e_c are both 500 dimensional vectors.~~
- ☐ θ_t and e_c are both 10000 dimensional vectors.

9. Suppose you have a 10000 word vocabulary, and are learning 500-dimensional word embeddings. The GloVe model minimizes this objective:

$$\min \sum_{i=1}^{10,000} \sum_{j=1}^{10,000} f(X_{ij})(\theta_i^T e_j + b_i + b_j - \log X_{ij})^2$$

Which of these statements are correct? Check all that apply.

- ☒ ~~θ_i and e_j should be initialized randomly at the beginning of training.~~
- ☒ ~~X_{ij} is the number of times word j appears in the context of word i .~~
- ☐ θ_i and e_j should be initialized to 0 at the beginning of training.
- ☒ ~~The weighting function $f(\cdot)$ must satisfy $f(0)=0$.~~

10. You have trained word embeddings using a text dataset of m_1 words. You are considering using these word embeddings for a language task, for which you have a separate labeled dataset of m_2 words. Keeping in mind that using word embeddings is a form of transfer learning, under which of these circumstances would you expect the word embeddings to be helpful?

- ☐ $m_1 \ll m_2$
- ☒ ~~$m_1 \gg m_2$~~