

QUIZ • 30 MIN

Natural Language Processing & Word Embeddings

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Natural Language Processing & Word Embeddings

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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.

1 / 1 point

- ☐ True
- ☒ False

✔ Correct

The dimension of word vectors is usually smaller than the size of the vocabulary. Most common sizes for word vectors ranges between 50 and 400.

2. What is t-SNE?

1 / 1 point

- ☐ A linear transformation that allows us to solve analogies on word vectors
- ☒ A non-linear dimensionality reduction technique
- ☐ A supervised learning algorithm for learning word embeddings
- ☐ An open-source sequence modeling library

✔ Correct

Yes

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.

1 / 1 point

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 $y = 1$.

- ☒ True
- ☐ False

✔ Correct

Yes, word vectors empower your model with an incredible ability to generalize. The vector for "ecstatic" would contain a positive/happy connotation which will probably make your model classify the sentence as a "1".

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

1 / 1 point

☒ $e_{\text{dog}} - e_{\text{girl}} \approx e_{\text{breathes}} - e_{\text{cancer}}$

✔ Correct

Yes!

☐ $e_{\text{dog}} - e_{\text{girl}} \approx e_{\text{cancer}} - e_{\text{breathes}}$

☒ $e_{\text{dog}} - e_{\text{breathes}} \approx e_{\text{girl}} - e_{\text{cancer}}$

✔ Correct

Yes!

☐ $e_{\text{dog}} - e_{\text{breathes}} \approx e_{\text{cancer}} - e_{\text{girl}}$

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

1 / 1 point

- ☒ It is computationally wasteful.
- ☐ The correct formula is $\vec{E}^T * \sigma_{1234}$.
- ☐ This doesn't handle unknown words (<UNK>).
- ☐ None of the above: calling the Python snippet as described above is fine.

✔ Correct

Yes, the element-wise multiplication will be extremely inefficient.

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.

1 / 1 point

- ☒ True
- ☐ False

✔ Correct

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.

1 / 1 point

- ☒ c and t are chosen to be nearby words.
- ☐ c is a sequence of several words immediately before t .
- ☐ c is the one word that comes immediately before t .
- ☐ c is the sequence of all the words in the sentence before t .

✔ Correct

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

1 / 1 point

$$P(t \mid c) = \frac{e^{\theta_c^T e_t}}{\sum_{j \in \text{VOCAB}} e^{e_j^T}}$$

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

☒ θ_t and e_c are both 500 dimensional vectors.

✔ Correct

☐ θ_t and e_c are both 10000 dimensional vectors.

☒ θ_t and e_c are both trained with an optimization algorithm such as Adam or gradient descent.

✔ Correct

☐ After training, we should expect θ_t to be very close to e_c when t and c are the same word.

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

1 / 1 point

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

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

☐ θ_i and e_j should be initialized to 0 at the beginning of training.

☒ θ_i and e_j should be initialized randomly at the beginning of training.

✔ Correct

☒ X_{ij} is the number of times word j appears in the context of word i .

✔ Correct

☒ The weighting function $f(\cdot)$ must satisfy $f(0) = 0$.

✔ Correct

The weighting function helps prevent learning only from extremely common word pairs. It is not necessary that it satisfies this function.

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 circumstance would you expect the word embeddings to be helpful?

1 / 1 point

- ☒ $m_1 \gg m_2$
- ☐ $m_1 \ll m_2$

✔ Correct