05_week2_quiz

Natural Language Processing & Word Embeddings

测验, 10 个问题 第 1 个问题 1 point

1。第1个问题

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

第2个问题

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point

2。第2个问题

What is t-SNE?

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

第3个问题

1

point

3。第3个问题

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.

x (input text) y (happy?)

I'm feeling wonderful today! 1

I'm bummed my cat is ill. 0

Really enjoying this!

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

第4个问题

1

point

4。第4个问题

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

eboy-egirl≈ebrother-esister

eboy-egirl≈esister-ebrother

eboy-ebrother≈egirl-esister

eboy-ebrother≈esister-egirl

第5个问题

1

point

5。第5个问题

Let E be an embedding matrix, and let e1234be a one-hot vector corresponding to word 1234. Then to get the embedding of word 1234, why don't we call E*e1234in Python?

It is computationally wasteful.

The correct formula is ET*e1234

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

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

第6个问题

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6。第6个问题

When learning word embeddings, we create an artificial task of estimating P(target 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

第7个问题

point

False

7。第7个问题

In the word2vec algorithm, you estimate P(t|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 is a sequence of several words immediately before t.

c and t are chosen to be nearby words.

第 8 个问题 1 point

8。第8个问题

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

 $P(t|c)=e\theta t Tec \Sigma t'=110000e\theta t' Tec$

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

Ot and ec are both 500 dimensional vectors.

θt and ecare both 10000 dimensional vectors.

θt and ec are both trained with an optimization algorithm such as Adam or gradient descent.

After training, we should expect θt to be very close to eC when t and C are the same word.

第 9 个问题 1 point

9。第9个问题

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

 $min\sum_{i=1}^{j=1} 10,000\sum_{j=1}^{j=1} 10,000f(Xij)(\theta iTej+bi+bj'-logXij)2$

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

θi and ej should be initialized randomly at the beginning of training.

Xij is the number of times word i appears in the context of word j.

θi and ej should be initialized to 0 at the beginning of training.

The weighting function f(.) must satisfy f(0)=0

第 10 个问题

1

point

10。第 10 个问题

You have trained word embeddings using a text dataset of m1 words. You are considering using these word embeddings for a language task, for which you have a separate labeled dataset of m2 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?

m1 >> m2

 $m1 \ll m2$

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