## Natural Language Processing & Word Embeddings

Quiz, 10 questions

| ~                  | Congratulations! You passed!   | Next Item             |
|--------------------|--|-----------------------|
|                    |  |                       |
| <b>~</b>           | 1 / 1 points   |                       |
| 1.                 |  |                       |
|                    | se you learn a word embedding for a vocabulary of 10000 words. Then the em<br>dimensional, so as to capture the full range of variation and meaning in those |                       |
|                    | True   |                       |
| 0                  | False  |                       |
|                    | ect<br>dimension of word vectors is usually smaller than the size of the vocabulary. I<br>d vectors ranges between 50 and 400.                               | Most common sizes for |
| <b>~</b>           | 1 / 1 points   |                       |
| 2.                 |  |                       |
| What i             | s t-SNE?   |                       |
|                    | A linear transformation that allows us to solve analogies on word vectors  |                       |
| 0                  | A non-linear dimensionality reduction technique  |                       |
| <b>Corr</b><br>Yes | ect  |                       |
|                    | A supervised learning algorithm for learning word embeddings   |                       |
|                    | An open-source sequence modeling library   |                       |
|                    |  |                       |

points

National Janguage Processing & Word Fmbeddings ained on a huge corpus of text. You then Quily se 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!        | 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 classified the sentence as a "1".





1/1 points

4.

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



$$e_{boy} - e_{girl} pprox e_{brother} - e_{sister}$$

#### Correct

Yes!

$$e_{boy} - e_{girl} pprox e_{sister} - e_{brother}$$

### **Un-selected is correct**



$$e_{boy} - e_{brother} pprox e_{girl} - e_{sister}$$

#### Correct

Yes!

| Natural 1 | Language | Processing | & Word | Embeddings |
|-----------|----------|------------|--------|------------|
|           |          |            |        |            |

| Quiz, | 10 | que <b>£i</b> թŋs— | $e_{brother}$ | $\approx$ | $e_{sister}$ | _ | $e_{girl}$ |
|-------|----|--------------------|---------------|-----------|--------------|---|------------|
|-------|----|--------------------|---------------|-----------|--------------|---|------------|

### **Un-selected is correct**



1/1 points

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?



It is computationally wasteful.

#### Correct

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

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



1/1 points

6

When learning word embeddings, we create an artificial task of estimating  $P(target \mid context)$ . 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

Correct

False



1/1 points

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 $t$ |
|---|---|
| Natural Language Processing & Wi                      | ard Embeddings  |

| NaturaloLanguager Prioces single by Word Embeddings   | τ   |
|---|-----|
| Quiz, 10 questions $c$ is a sequence of several words immediately before $t.$   |     |
| igcap c is the one word that comes immediately before $t.$  |     |
| igcap c is the sequence of all the words in the sentence before $t.$  |     |
| $igcup_c$ and $t$ are chosen to be nearby words.  |     |
| Correct   |     |
| 1/1 points  |     |
| Suppose you have a 10000 word vocabulary, and are learning 500-dimensional word embeddings. The word2v model uses the following softmax function: $P(t \mid c) = \frac{e^{\theta_t^T e_c}}{\sum_{t'=1}^{10000} e^{\theta_t^T e_c}}$ | vec |
| Which of these statements are correct? Check all that apply.  |     |
| $igcup_t$ and $e_c$ are both 500 dimensional vectors.   |     |
| Correct   |     |
| $igcup_t$ and $e_c$ are both 10000 dimensional vectors.   |     |
| Un-selected is correct  |     |
| $igcup_t$ and $e_c$ are both trained with an optimization algorithm such as Adam or gradient descent.   |     |
| Correct   |     |

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

**Un-selected is correct** 

# Natural Language Processing & Word Embeddings

Quiz, 10 questions

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.

oxedge and  $e_j$  should be initialized to 0 at the beginning of training.

**Un-selected is correct** 

 $oxedsymbol{ heta}_i$  and  $e_j$  should be initialized randomly at the beginning of training.

Correct

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

Correct

The weighting function f(.) 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.



1/1 points

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?

 $\bigcap$   $m_1 >> m_2$ 

Correct

 $m_1 \ll m_2$ 

# Natural Language Processing & Word Embeddings

Quiz, 10 questions



