

1. True/False: Suppose you learn a word embedding for a vocabulary of 20000 words. Then the embedding vectors could be 1000 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 range between 50 and 1000.

2. True/False: t-SNE is a non-linear dimensionality reduction technique.

1 / 1 point

- ☒ True
- ☐ False

✔ Correct

t-SNE is a non-linear dimensionality reduction technique.

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)

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☒ $e_{man} - e_{woman} \approx e_{king} - e_{queen}$

✔ Correct

The order of words is correct in this analogy.

☐ $e_{man} - e_{king} \approx e_{queen} - e_{woman}$

☐ $e_{man} - e_{woman} \approx e_{queen} - e_{king}$

☒ $e_{man} - e_{king} \approx e_{woman} - e_{queen}$

✔ Correct

The order of words is correct in this analogy.

5. True/False: The most computationally efficient formula for Python to get the embedding of word 1021, if C is an embedding matrix, and o_{1021} is a one-hot vector corresponding to word 1021, is $C^T * o_{1021}$.

1 / 1 point

- ☒ False
- ☐ True

✔ Correct

It is computationally wasteful because the element-wise multiplication will be extremely inefficient.

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.

1 / 1 point

- ☒ True
- ☐ False

✔ Correct

7. True/False: In the word2vec algorithm, you estimate $P(t \mid c)$, where t is the target word and c is a context word. t and c are chosen from the training set to be nearby words.

1 / 1 point

- ☒ True
- ☐ False

✔ Correct

Yes, t and c are chosen from the training set to be nearby words.

8. Suppose you have a 10000 word vocabulary, and are learning 100-dimensional word embeddings. The word2vec model uses the following softmax function:

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

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

- ☐ True
- ☒ False

✔ Correct

To review this concept watch the *Word2Vec* lecture.

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

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

True/False: θ_i and e_j should be initialized to 0 at the beginning of training.

- ☐ True
- ☒ False

✔ Correct

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

10. You have trained word embeddings using a text dataset of t_1 words. You are considering using these word embeddings for a language task, for which you have a separate labeled dataset of t_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?

1 / 1 point

- ☒ When t_1 is larger than t_2
- ☐ When t_1 is equal to t_2
- ☐ When t_1 is smaller than t_2

✔ Correct

Transfer embeddings to new tasks with smaller training sets.