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[Unit 5 Reinforcement Learning \(2 weeks\)](#)

[Lecture 19: Applications: Natural Language Processing](#)

> [6. Word Embeddings](#)

## 6. Word Embeddings

### Word Embeddings

**Sparsity**

- Features are often represented as high-dimensional, sparse vectors

# Training Sentences	unigram (Seen Arcs)	bigrams (Seen Arcs)
0	0.85	0.30
1000	0.88	0.35
5000	0.92	0.45
10000	0.95	0.55
15000	0.98	0.60

— unigram  
— bigrams

▶ 0:00 / 0:00

▶ 1.50x



## Video

[Download video file](#)

## Understanding Word Embeddings

1/1 point (graded)

Which of the following options is correct about word embeddings presented in the lecture.

- ☐ The goal of word embeddings is to increase the sparsity of the encoded input word features
- ☐ We would like similar words to have word embeddings that are far apart in order to minimize word sense disambiguation
- ☒ One way to learn word embeddings is by minimizing cosine similarity between words with related meaning. ✓
- ☐ To do a good job, word embeddings have to manually encoded by a natural language domain expert

### Solution:

We would like to learn word embeddings that are much less sparse than one hot vector based encoding because reducing the sparsity of input features lowers the sample complexity (number of training examples required to do an accurate task) of the downstream text classification task.

In order to do the above, we should cluster the similar or related words together in the embedding dimension space. For instance, the words "dog" and "samoyed" must have similar embedding representations than "dog" and "lipstick"

Word embeddings are practically very useful because they can be learnt without any significant manual effort and they generalize well to completely new tasks.

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You have used 1 of 2 attempts

 Answers are displayed within the problem

## Sentence Embeddings vs bag-of-words

1/1 point (graded)

Consider the following two sentences with very different meanings:

(1) I ate pizza with my friend

(2) I ate my friend with pizza

☒ Bag of words encoding approach would lead to identical encodings for both these sentences ✓

☐ Bag of words encoding approach would be able to successfully differentiate the above two sentences

☐ Recurrent neural network (e.g., LSTMs) based approach would lead to identical encodings for both these sentences

☒ Recurrent neural network (e.g., LSTMs) based approach would be able to successfully differentiate the above two sentences ✓

**Solution:**

Bag of words approach sums up all the word embeddings in order to encode an input sentence. Hence, it cannot capture the ordering of these words within a sentence.

LSTM or Recurrent Neural network based approaches encode an input sentence into a context vector capturing more than just a summation of its constituent parts together.

Submit

You have used 1 of 1 attempt

**i** Answers are displayed within the problem

## Discussion

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? [Staff] Issue with Understanding Word Embeddings

2

I may be wrong, but I think there is a problem with the question, as I think there is no valid choice right now. Maybe there is a typo somewhere?

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