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Machine Learning with Python-From Linear Models to Deep Learning

<u>Help</u>



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Lecture 11. Recurrent Neural Course > Unit 3 Neural networks (2.5 weeks) > Networks 2 > 1. Markov Models 1. Markov Models Markov Models Start of transcript. Skip to the end. Recall: learning to encode/decode · Language modeling This course has been a success (?) Sentiment classification I have seen better lectu Welcome back. Today we're going to be talking about how Machine translation Olen nähnyt parempia to generate sequences using recurrent neural I have seen better lectures networks. luentoja So if you recall what we did last time, we showed how recurrent neural networks can encoding decoding be used to condense a map sequences into vector representations, and those vector representations X CC 0:00 / 0:00 ▶ Speed 1.50x can then be used for all kinds of predictions, Video **Transcripts** Download video file Download SubRip (,srt) file Download Text (.txt) file

Markov Symbols

1/1 point (graded)

To specify a Markov language model, what additional symbols do we need to add to our set of possible symbols?

- a start symbol
- an end symbol

 ✓
- a symbol for unknown words
- a symbol for complicated words



Solution:

As shown in the lecture video, we need start and end symbols in order to specify the bounds of our sentence. We also need a symbol for unknown words as there might be words in the sentence not represented in our Markov Model. Lastly, our Markov model treats words the same regardless of complexity. We're simply representing a table of transition probabilities, so there's no need to create extra symbols for complex words.

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You have used 1 of 2 attempts

1 Answers are displayed within the problem

Transition Probabilities

2/2 points (graded)

			w_i			
		ML	course	is	UNK	<end></end>
	<beg></beg>	0.7	0.1	0.1	0.1	0.0
	ML	0.1	0.5	0.2	0.1	0.1
w_{i-1}	course	0.0	0.0	0.7	0.1	0.2
	is	0.1	0.3	0.0	0.6	0.0
	UNK	0.1	0.2	0.2	0.3	0.2

Using a first order Markov model specified above, what is the probability of generating the following sentence <beg> ML course UNK <end>?

0.007

0.01

0.003

0.005

Which the following sentences are possible to generate? Choose all those apply.

<beg> course ML is UNK <end>

<beg> <end>

course is ML <end>

<beg>ML course <end>



Solution:

For the first question,

 $P(ML| < beg >) \times P(course|ML) \times P(UNK|course) \times P(< end > |UNK) = 0.7 \times 0.5 \times 0.1 \times 0.2 = 0.007$

. For the second question, all valid sentences must start with <beg>. The probability of the second word being <end> is zero. In addition, the probability of ML occurring after course is also zero. The correct answer is the only choice with nonzero transition probabilities at each step.

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You have used 2 of 2 attempts

1 Answers are displayed within the problem

Maximum Likelihood

1/1 point (graded)

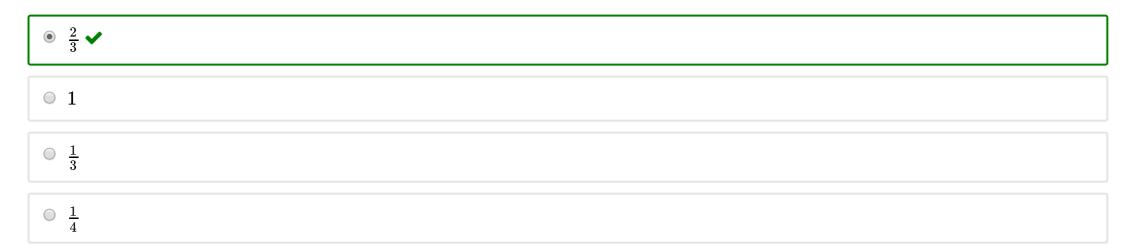
Suppose our training examples are the following three sentences.

ML courses are cool.

Humanities courses are cool.

But some courses are boring.

Using a bigram model, what is the maximum likelihood estimate for the probability that the next word is 'cool', given that the previous word is 'are'?



Solution:

"are" occurs three times in the training corpus, and is followed by "cool" two out of those three times.

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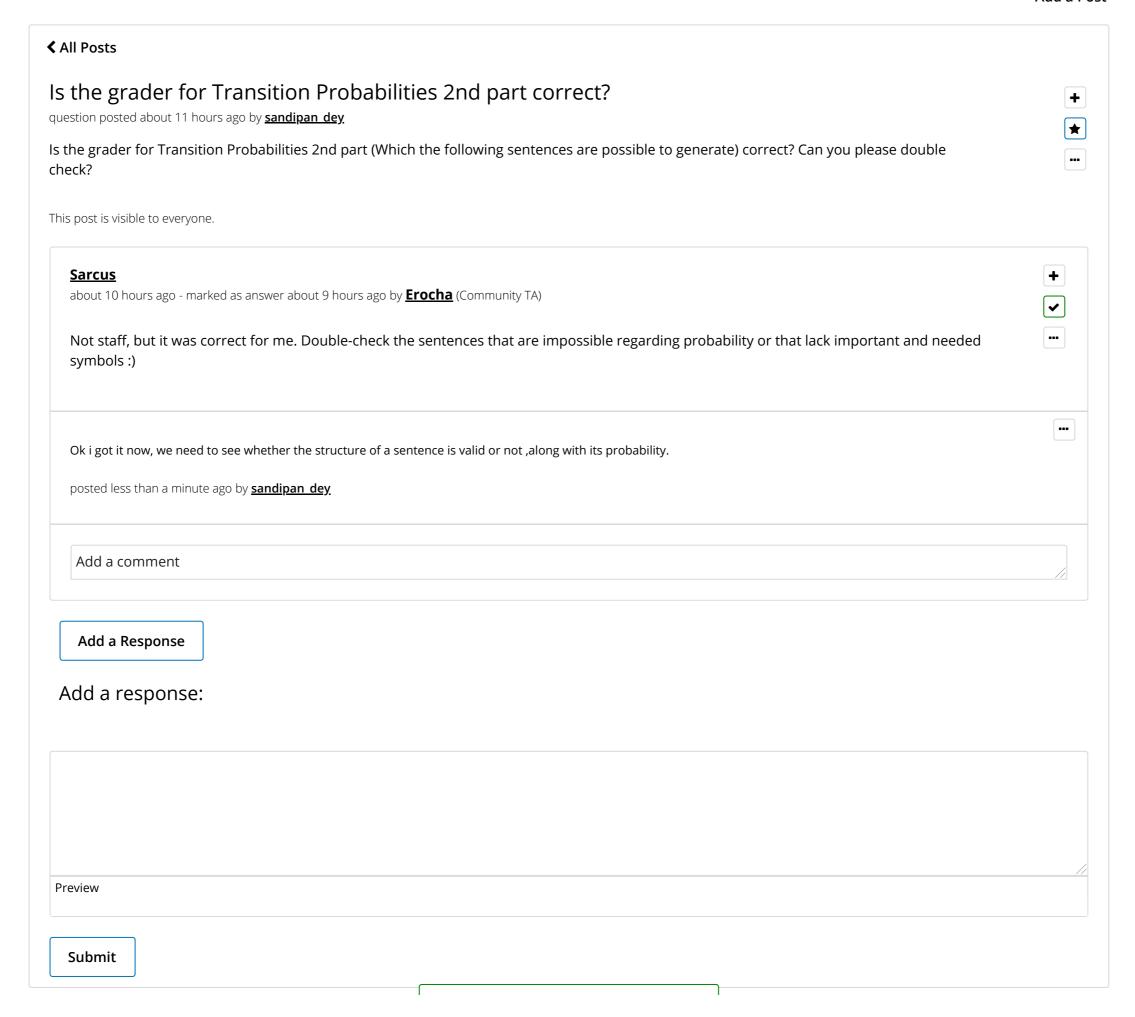
You have used 1 of 2 attempts

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Discussion

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Topic: Unit 3 Neural networks (2.5 weeks):Lecture 11. Recurrent Neural Networks 2 / 1. Markov Models



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