

Neural networks for text

What is text?

You can think of text as a sequence of

- Characters
- **Words**
- Phrases and named entities
- Sentences
- Paragraphs
- ...

Bag of words way (sparse)

~100k columns

*any word with
one hot encoded vector.
every word is
vectorized independently.*

very



| good | movie | very | a | did | like |
|------|-------|------|---|-----|------|
|------|-------|------|---|-----|------|

| | | | | | |
|---|---|---|---|---|---|
| 0 | 0 | 1 | 0 | 0 | 0 |
|---|---|---|---|---|---|

good



| | | | | | |
|---|---|---|---|---|---|
| 1 | 0 | 0 | 0 | 0 | 0 |
|---|---|---|---|---|---|

movie



| | | | | | |
|---|---|---|---|---|---|
| 0 | 1 | 0 | 0 | 0 | 0 |
|---|---|---|---|---|---|

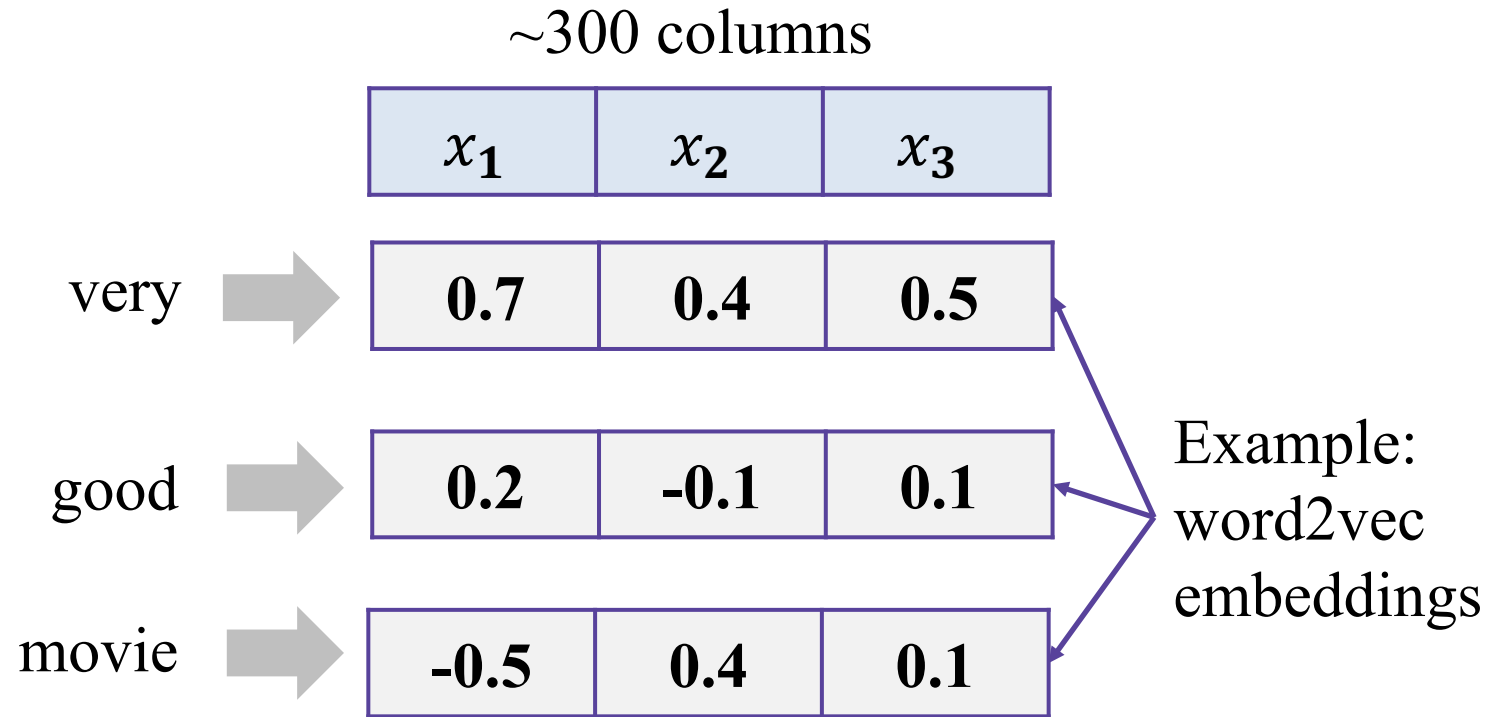
Bag of words way (sparse)

~100k columns

| | good | movie | very | a | did | like |
|-----------------|------|-------|------|---|-----|------|
| very | 0 | 0 | 1 | 0 | 0 | 0 |
| good | 1 | 0 | 0 | 0 | 0 | 0 |
| movie | 0 | 1 | 0 | 0 | 0 | 0 |
| very good movie | 1 | 1 | 1 | 0 | 0 | 0 |

Bag of words representation
is a sum of sparse one-hot-encoded vectors

Neural way (dense)

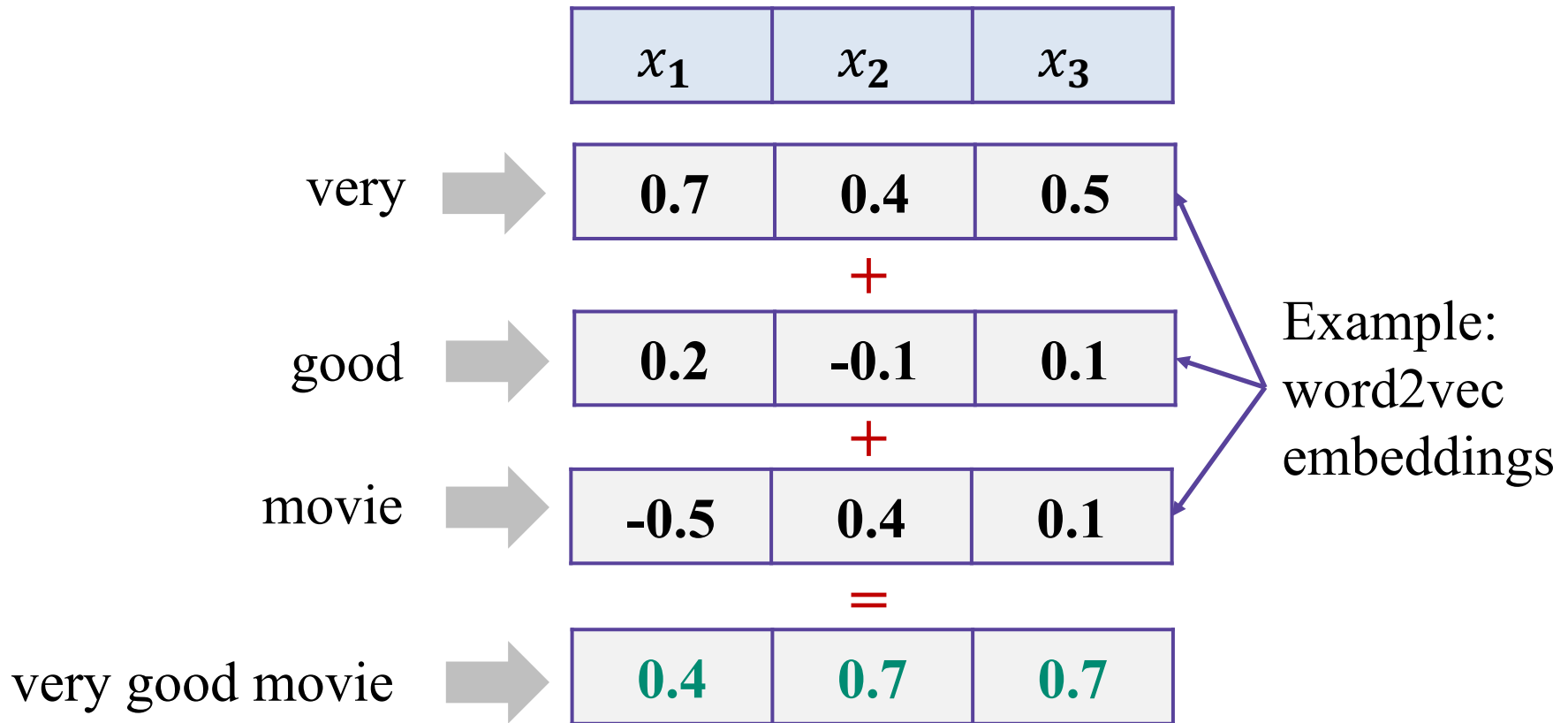


Word2vec property:

Words that have similar context tend to have collinear vectors

Neural way (dense)

~300 columns



Sum of word2vec vectors
can be a good text descriptor already!

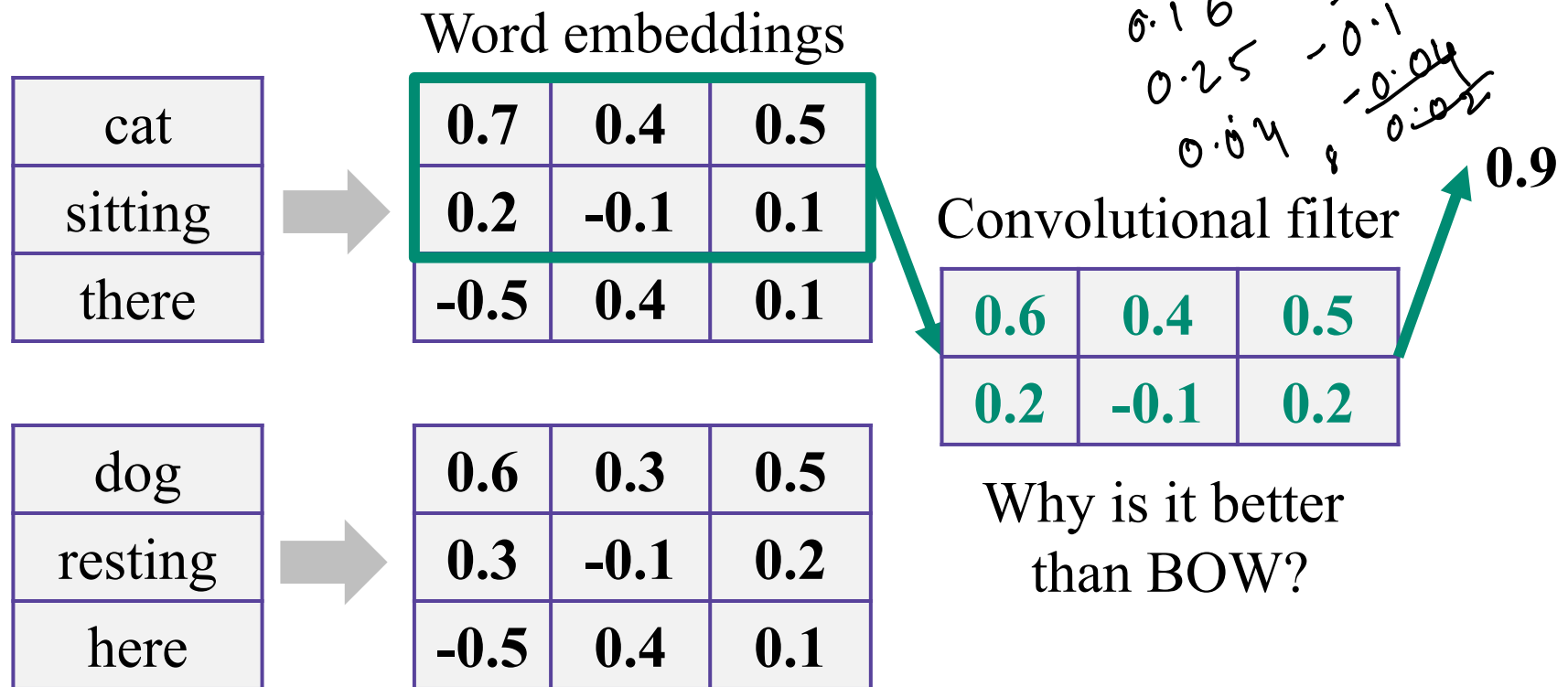
A better way: 1D convolutions

| | | | | |
|-----------------|--|-------------|-------------|------------|
| Word embeddings | | | | |
| cat | | 0.7 | 0.4 | 0.5 |
| sitting | | 0.2 | -0.1 | 0.1 |
| there | | -0.5 | 0.4 | 0.1 |

| | | | |
|---------|-------------|-------------|------------|
| dog | 0.6 | 0.3 | 0.5 |
| resting | 0.3 | -0.1 | 0.2 |
| here | -0.5 | 0.4 | 0.1 |

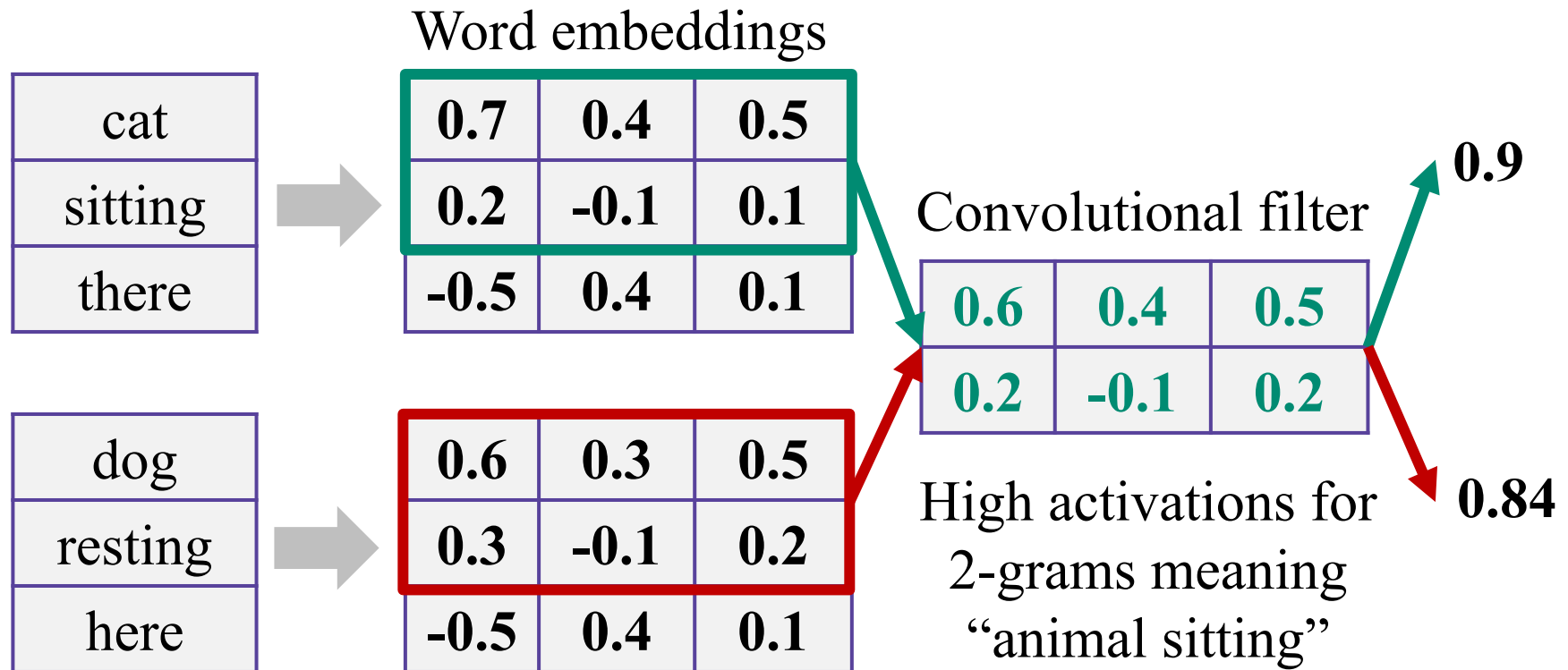
How do we make 2-grams?

A better way: 1D convolutions



- This convolution provides high activations for 2-grams with certain meaning

A better way: 1D convolutions



- This convolution provides high activations for 2-grams with certain meaning
- Word2vec vectors for similar words are similar in terms of cosine distance (similar to dot product)

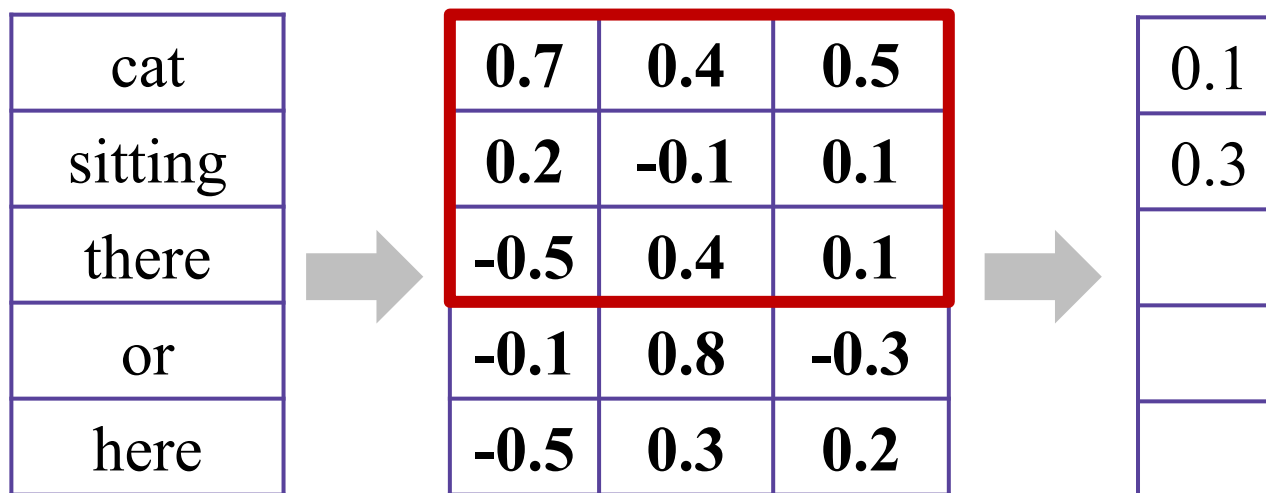
1D convolutions

- Can be extended to 3-grams, 4-grams, etc.
- One filter is not enough, need to track many n-grams
- They are called 1D because we slide the window only in one direction



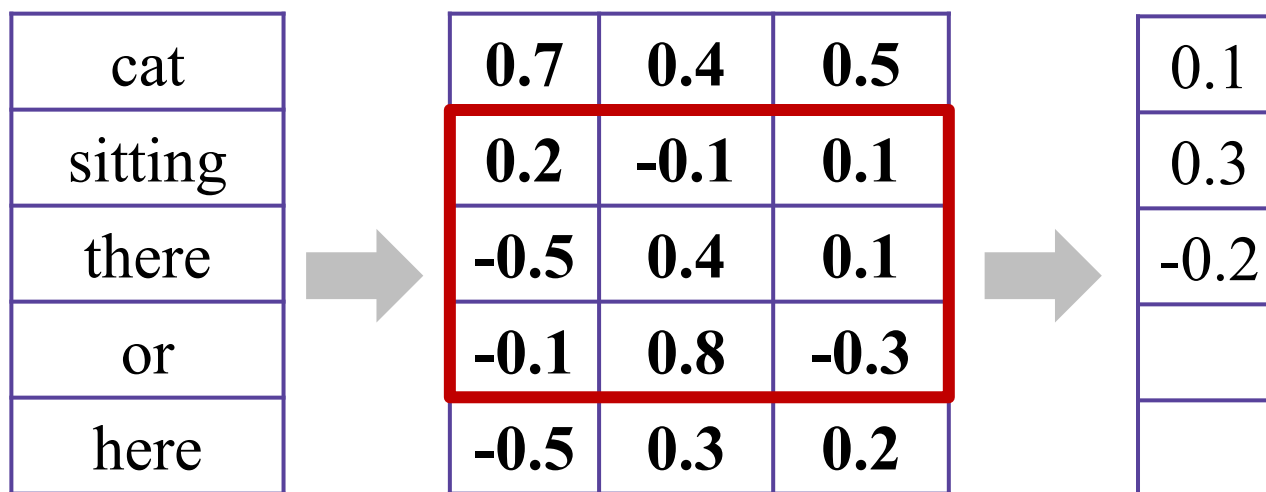
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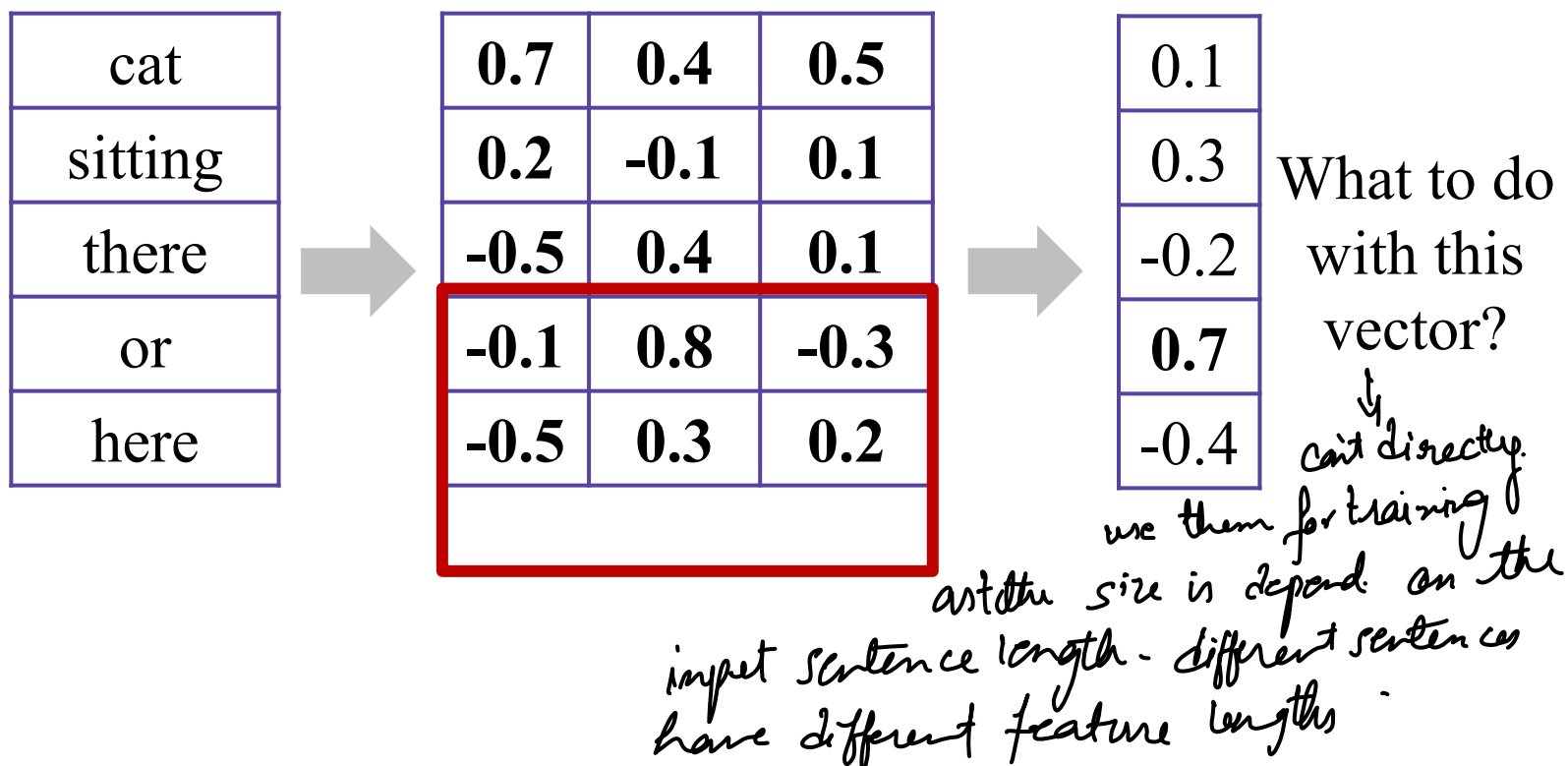
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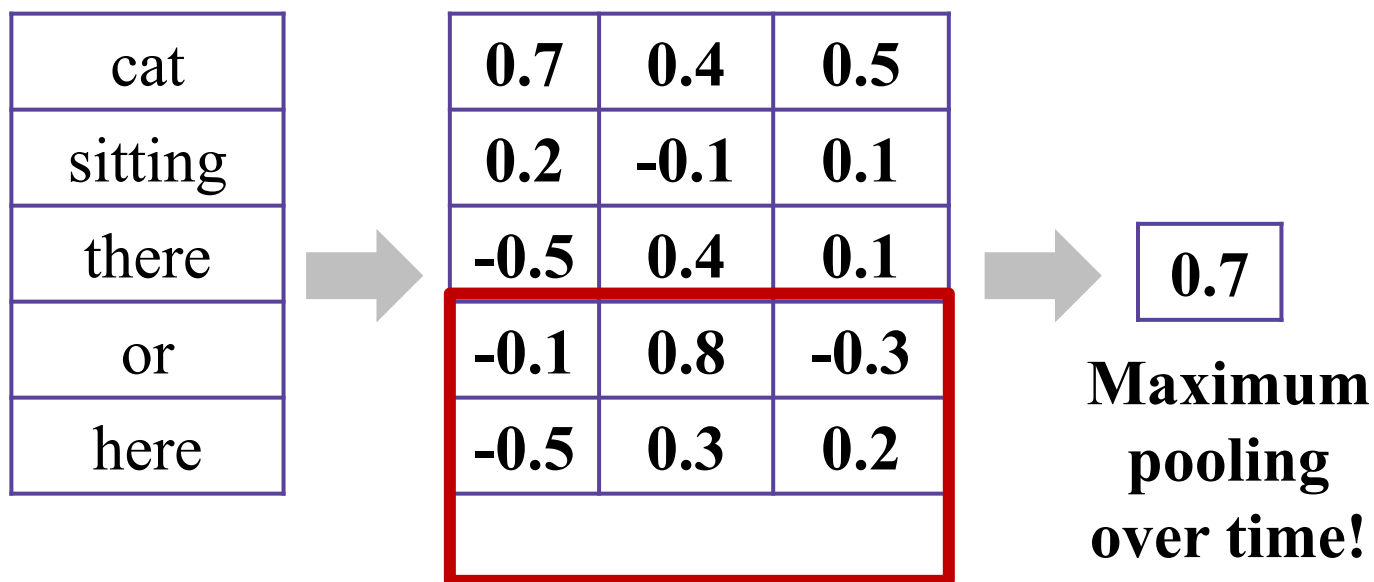
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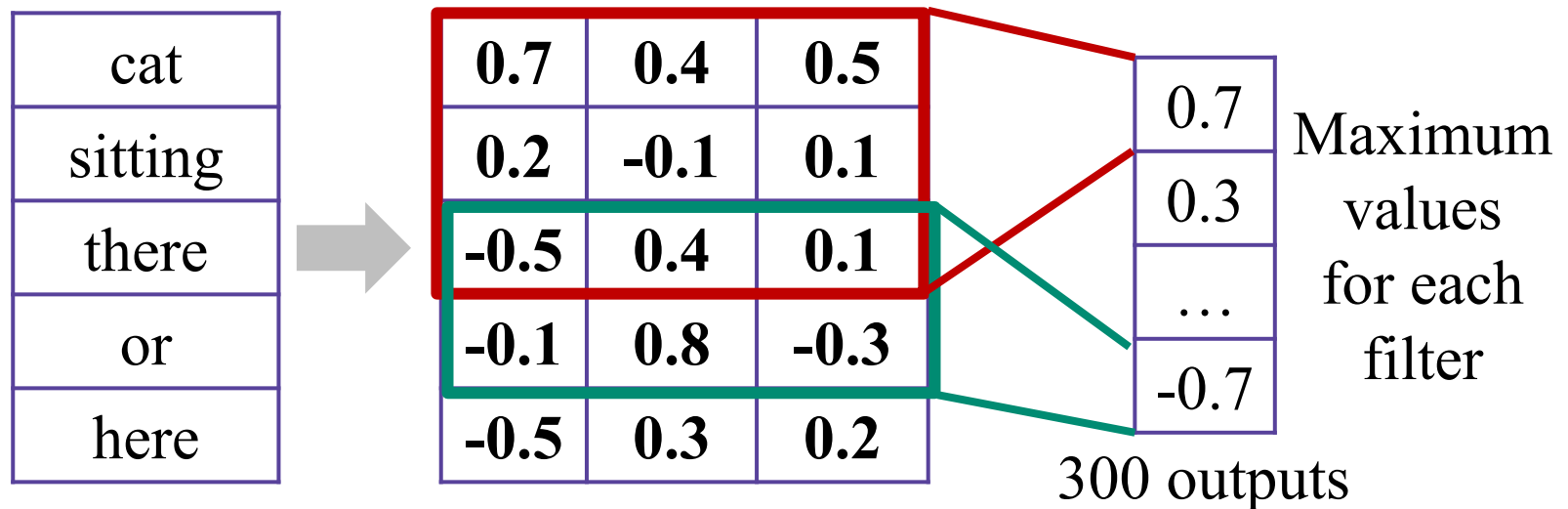
Let's train many filters

Final architecture

- 3,4,5-gram windows with 100 filters each
- MLP on top of these 300 features

Quality comparison on customer reviews (CR)

- Naïve Bayes on top of 1,2-grams – 86.3% accuracy
- 1D convolutions with MLP – 89.6% (+3.8%) accuracy



Summary

- You can just average pre-trained word2vec vectors for your text
- You can do better with 1D convolutions that learn more complex features
- In the next video we'll continue to apply convolutions to text