

Assignment 1

Learning distributed word representations

Jimmy Ba

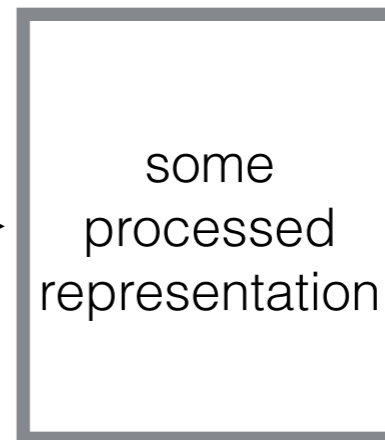
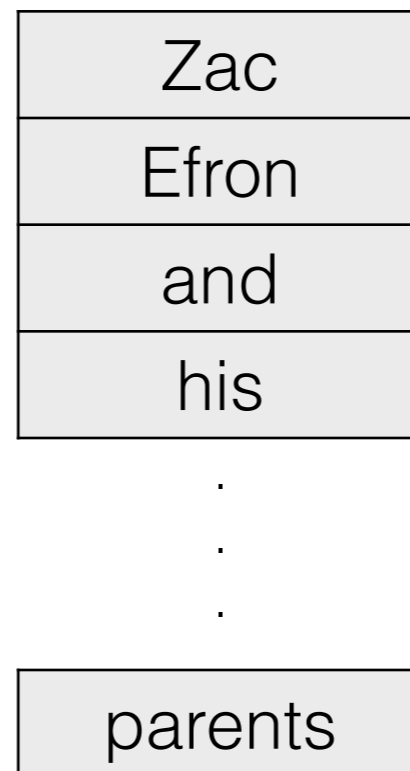
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Background

- Text and language play central role in a wide range of computer science and engineering problems
- Applications that depend on language understanding/processing includes: speech processing, search/query internet, social media, recommendation system, artificial intelligence and many others

Motivation

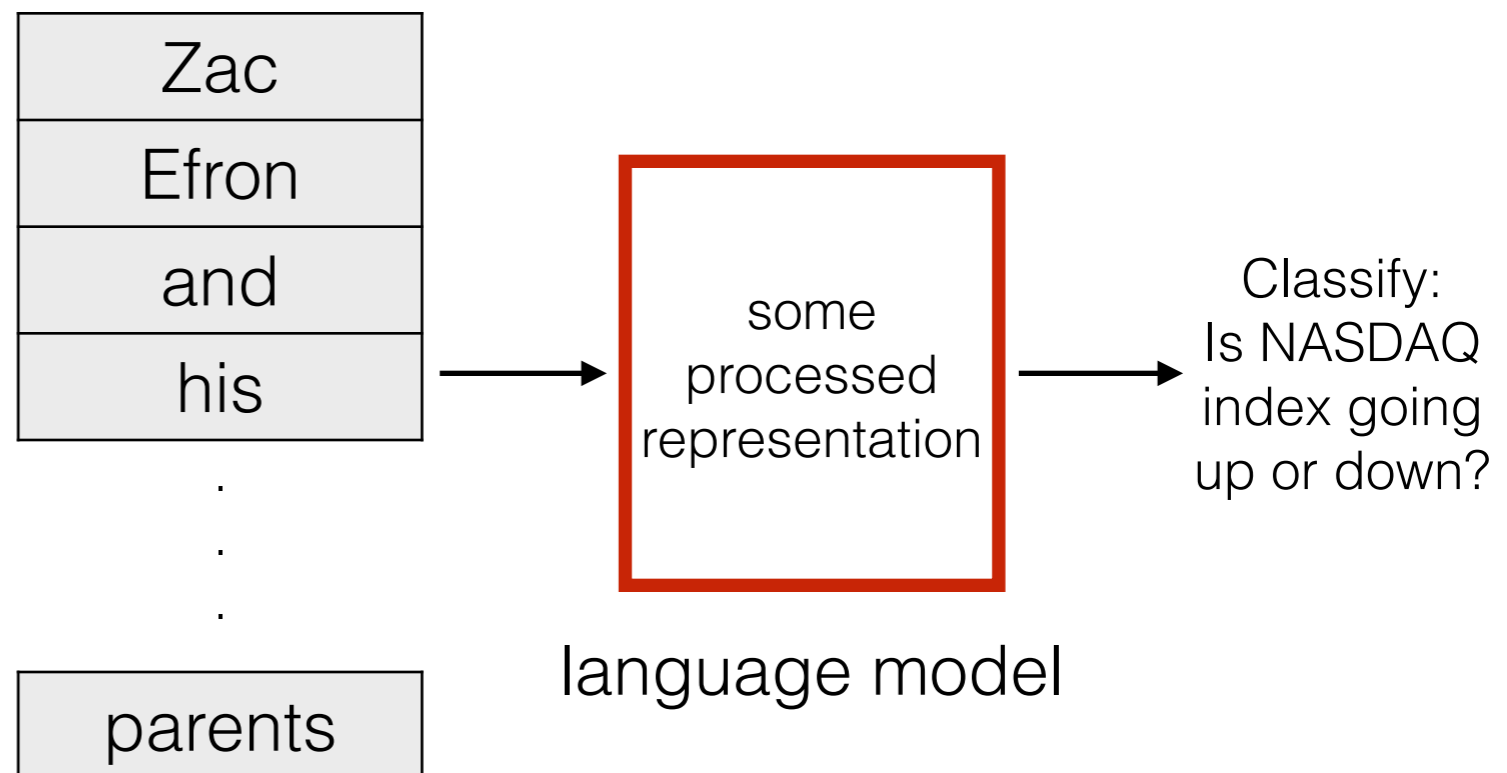
- Getting meaningful representations from text data are often the key component in Google search engine or your next big start-up ideas



Classify:
Is NASDAQ
index going
up or down?

Motivation

- Getting meaningful representations from text data are often the key component in Google search engine or your next big start-up ideas



Language Model

- We need to represent text data in a way that is “easy” for the later stage classification problem or learning algorithms
- “Easy”: Be able to handle large scale vocabulary and words have similar syntactic/semantic meaning should be close in the representation space

Language Model

one-of-K encoding

binary encoding

“Zac”

1	0	0	0	0	0	0	0
---	---	---	---	---	---	---	---

0	0	0
---	---	---

“Efron”

0	1	0	0	0	0	0	0
---	---	---	---	---	---	---	---

1	0	0
---	---	---

“and”

0	0	1	0	0	0	0	0
---	---	---	---	---	---	---	---

0	1	0
---	---	---

“his”

0	0	0	1	0	0	0	0
---	---	---	---	---	---	---	---

1	1	0
---	---	---

.

.

.

.

.

.

“parents”

0	0	0	0	0	0	0	1
---	---	---	---	---	---	---	---

1	1	1
---	---	---

Language Model

one-of-K encoding

binary encoding

“Zac”

1	0	0	0	0	0	0	0
---	---	---	---	---	---	---	---

0	0	0
---	---	---

“Efron”

0	1	0	0	0	0	0	0
---	---	---	---	---	---	---	---

1	0	0
---	---	---

“and”

0	0	1	0	0	0	0	0
---	---	---	---	---	---	---	---

0	1	0
---	---	---

“his”

0	0	0	1	0	0	0	0
---	---	---	---	---	---	---	---

1	1	0
---	---	---

.

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“parents”

0	0	0	0	0	0	0	0	1
---	---	---	---	---	---	---	---	---

1	1	1
---	---	---

← vocabulary size →

← $\log(\text{vocabulary size})$ →

Language Model

one-of-K encoding

distributed encoding

“Zac”

1	0	0	0	0	0	0	0
---	---	---	---	---	---	---	---

1.5	0.1	-0.1	2.1
-----	-----	------	-----

“Efron”

0	1	0	0	0	0	0	0
---	---	---	---	---	---	---	---

0.7	-0.1	0.3	0.4
-----	------	-----	-----

“and”

0	0	1	0	0	0	0	0
---	---	---	---	---	---	---	---

0.1	1.6	-1.9	1.1
-----	-----	------	-----

“his”

0	0	0	1	0	0	0	0
---	---	---	---	---	---	---	---

3.5	0.2	1.1	-2.5
-----	-----	-----	------

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“parents”

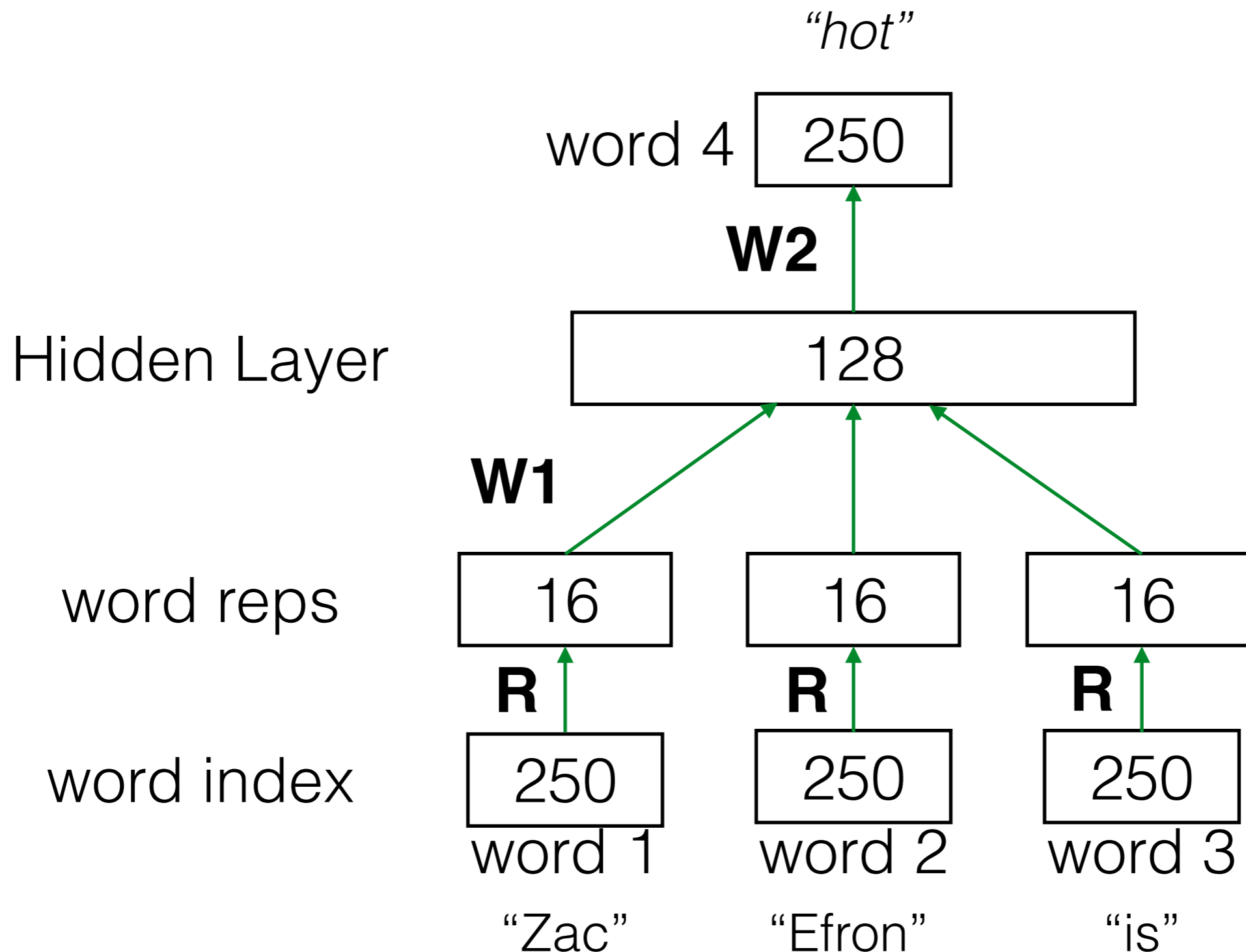
0	0	0	0	0	0	0	1
---	---	---	---	---	---	---	---

-2.1	-3.3	-2.7	1.9
------	------	------	-----

← vocabulary size →

← embedding size (constant) →

Neural Language Model



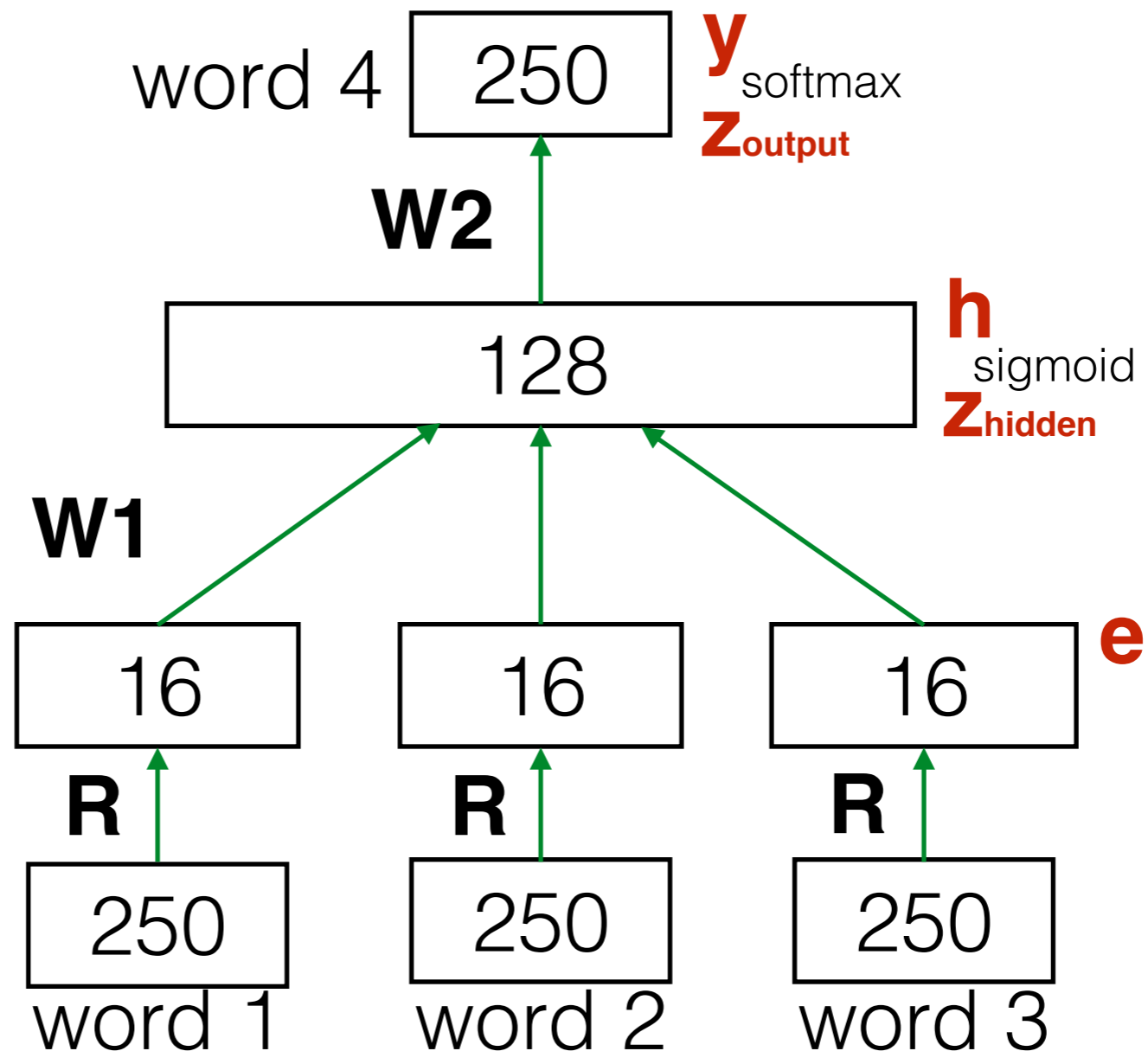
Neural Language Model

cross entropy
cost: $C(W)$

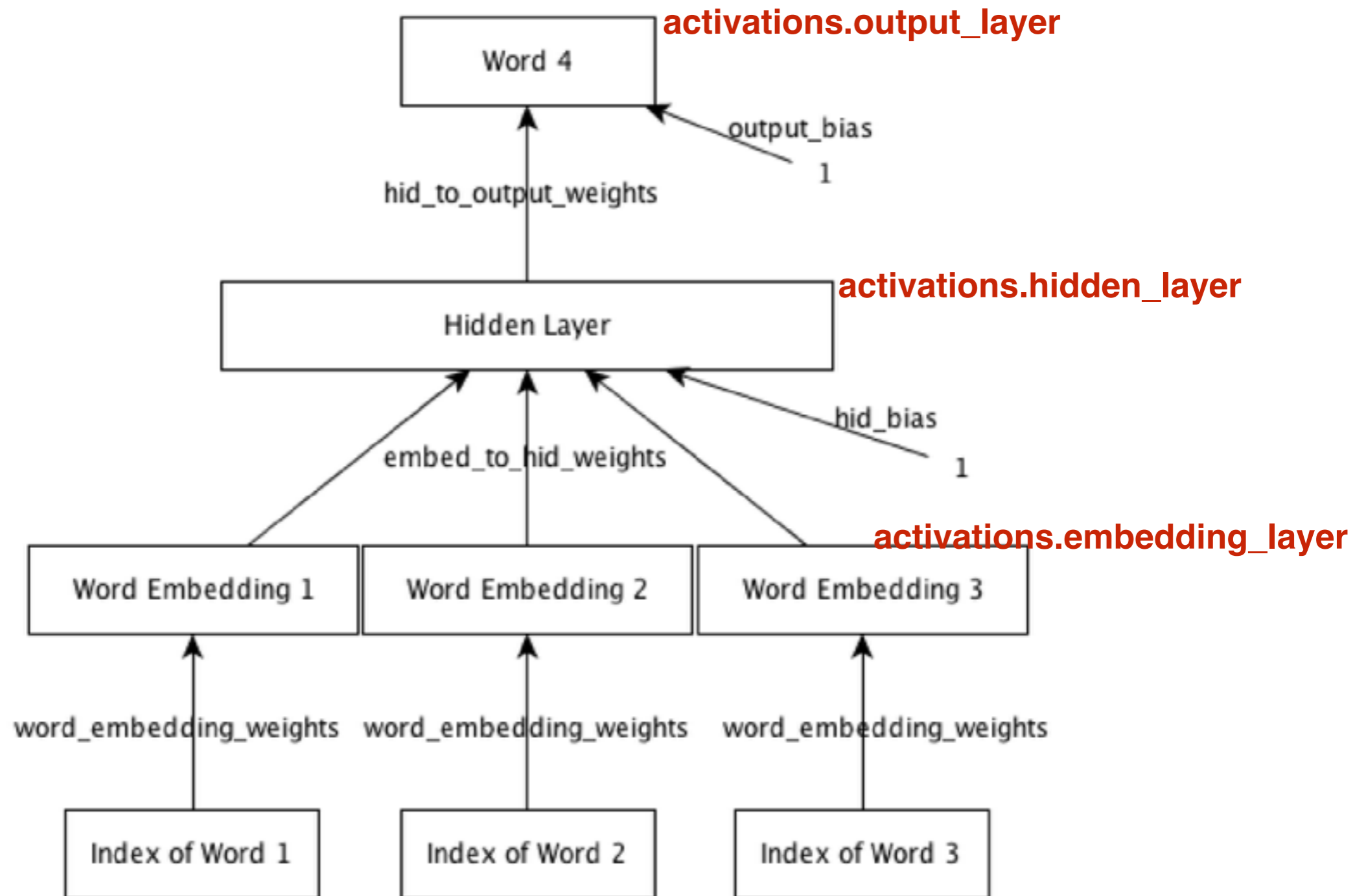
Hidden Layer

word reps

word index



Neural Language Model



Things you need to do in assignment 1

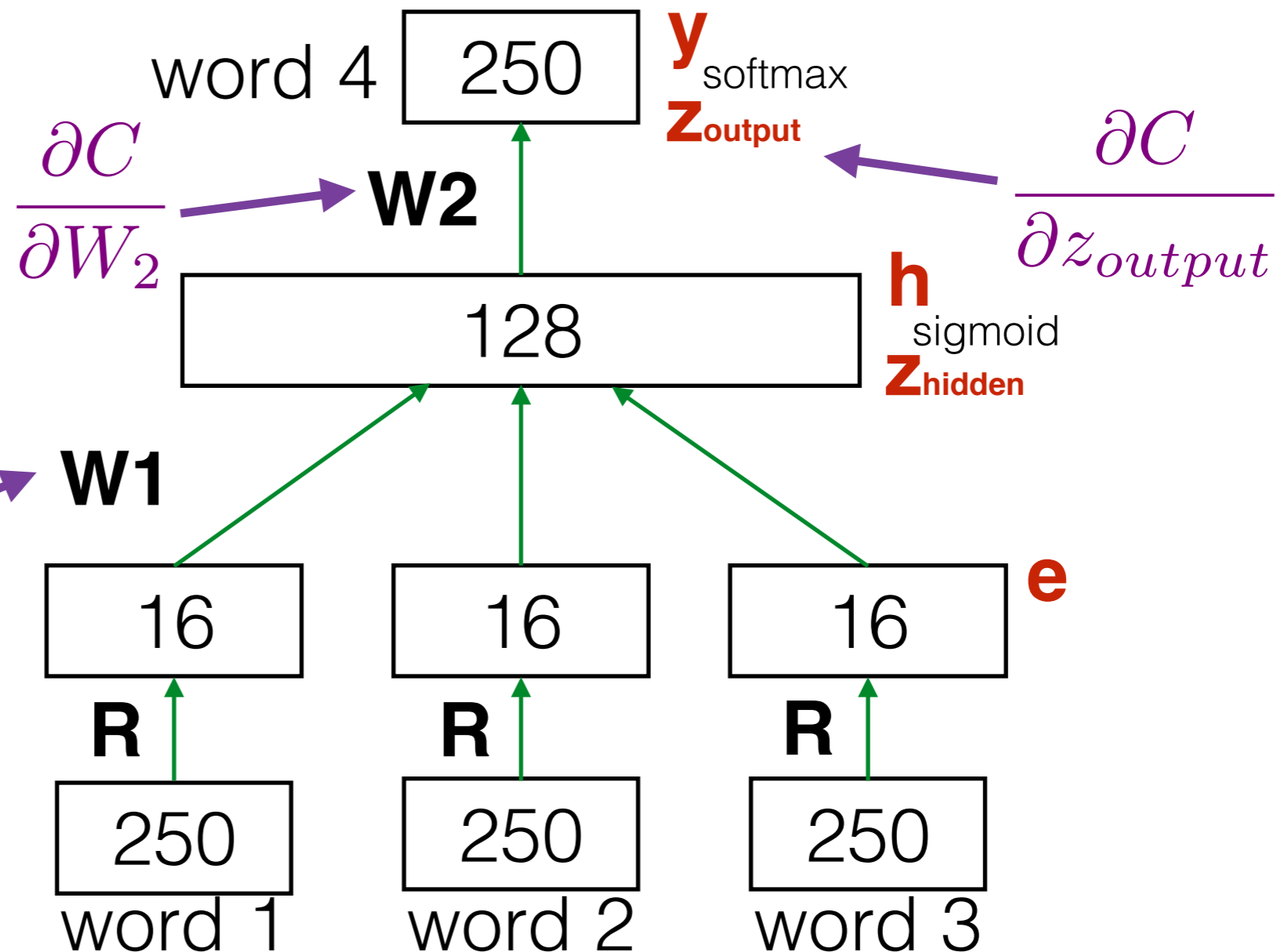
- Part 2

cross entropy
cost: $C(W)$

Hidden Layer

word reps

word index

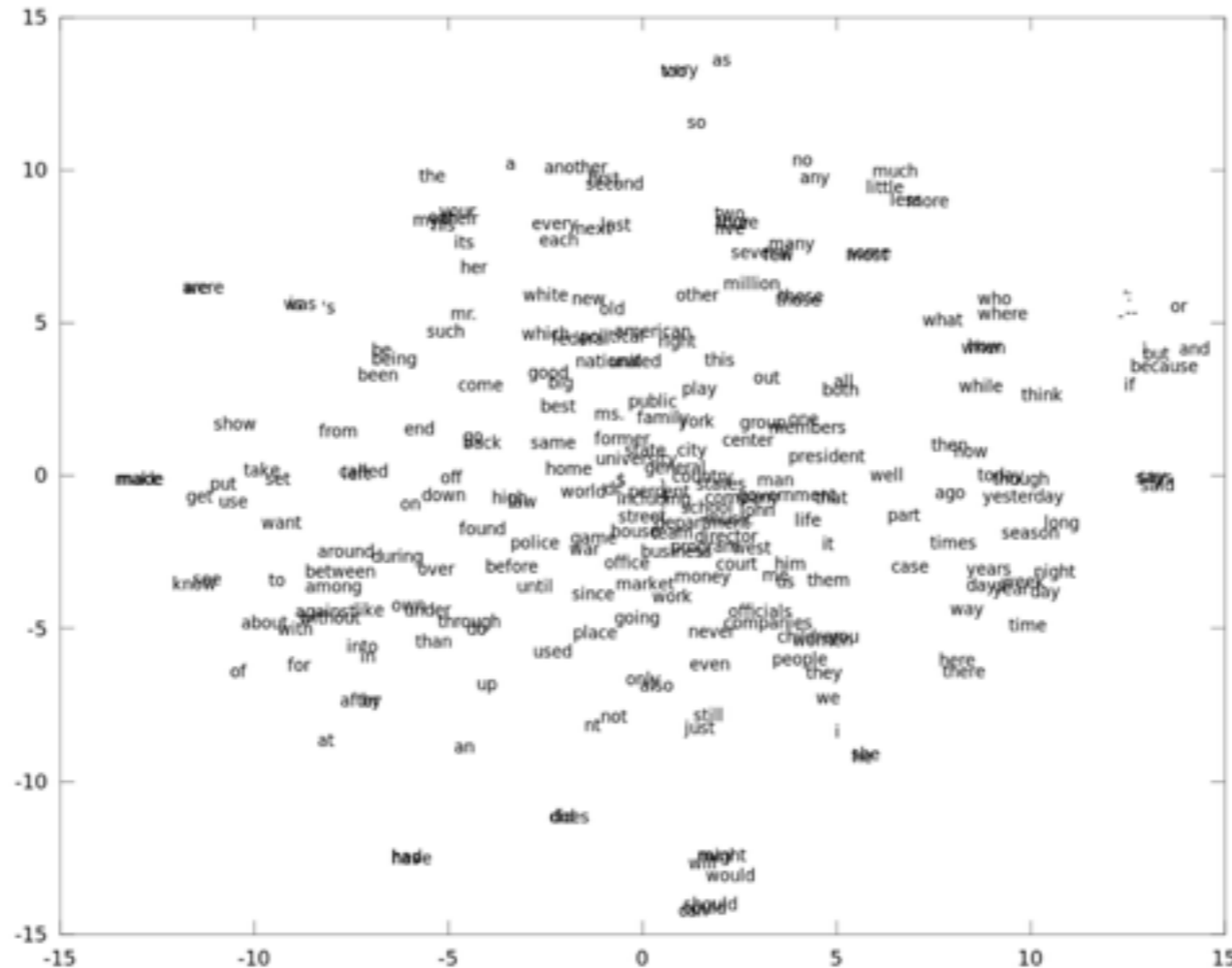


Things you need to do in assignment 1

- Part 2
 - code/derive the partial derivative of cross-entropy cost with respect to softmax input
 - code/derive the gradients of the weight matrix using partial derivatives from backdrop
 - can be done in just 5 simple lines of code and NO for-loops
 - use *checking.check_gradients* to verify the correctness of the 5 lines of code

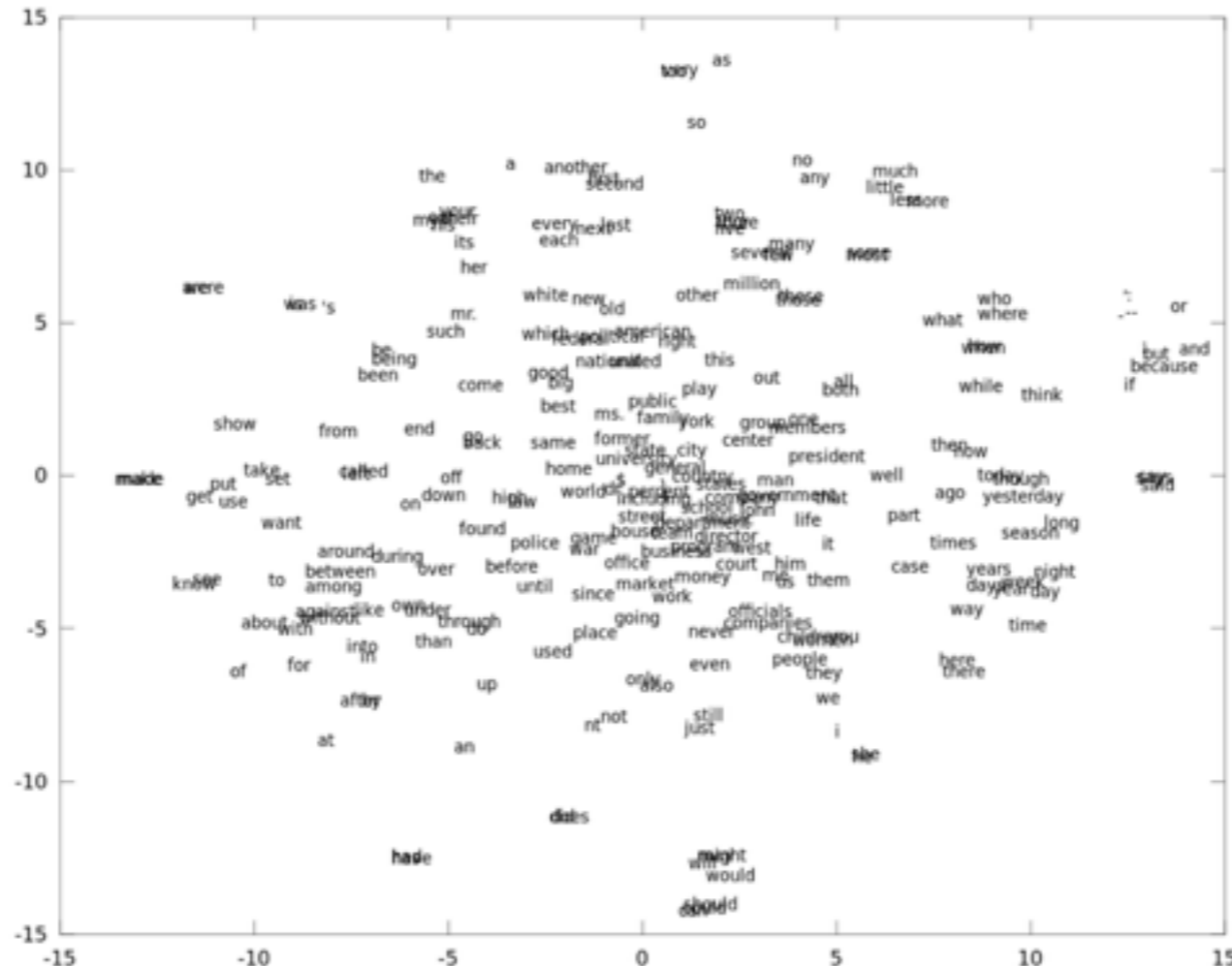
Things you need to do in assignment 1

- Part 3
 - analyze the trained model



t-SNE embedding

- It projects 16D learnt word embedding to 2D for plotting visualization only. (*display_nearest_words*, *word_distance* uses the 16D word embedding)



Word Distance

- It only makes sense to compare the relative distances between words, i.e.
 - $\text{distance}(A, B)$ and $\text{distance}(A, C)$
 - $\text{distance}(A, B)$ and $\text{distance}(A, w)$, $\text{distance}(B, w)$
 - NOT $\text{distance}(A, B)$ and $\text{distance}(C, D)$

Things you need to do in assignment 1

- Part 3
 - Think about how the model would put two words close together in embedding space
 - Think about what the task the model is trying to achieve and how that affects the word representation that is being learned.
 - Think about what kind of similarity the nearest words in the 16D embedding space have

Due: Tuesday, Feb. 3

at the start of lecture