



Midterm 2 Review

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CS 421: Natural Language
Processing

Fall 2019

Exam Format

True/False (30%)

- 15 questions, 2 points each
- No partial credit (autograded)

Multiple Choice (40%)

- 8 questions, 5 points each
- No partial credit (autograded)

Problem Solving (30%)

- 3 questions, 10 points each
- Show work for partial credit (graded manually)

Bonus Question (10%)

- Problem-solving question
- Points added to exam score (max exam score = 100)

Sample Midterm

- Currently available on Piazza
- Solution will be posted after class

What should I study?



- Questions designed based on **slides** and **assignments**
- Problem solving questions similar but not identical to **problem solving examples** (e.g., stepping through a neural network) in slides

What
content will
the exam
cover?

N-Grams

Language Modeling

Word Embeddings

Cosine Similarity

Naive Bayes

Text Classification

Evaluation Metrics

Feedforward Neural Networks

Recurrent Neural Networks

What will I for sure *not* need to memorize?

- Derivatives
- Log values
- Language model interpolation
- Word embedding visualization techniques
- Implementation details for GloVe, ELMo, or BERT
- Backpropagation equations
 - Do have a general sense of how backpropagation works
- Implementation details for LSTM and GRU gates

What should I bring to the exam?

- Pen or pencil
- UIN (you'll need to write it on the first page of the exam)
- This exam will be:
 - Closed note
 - Closed book
 - Closed device
- You will not need a calculator



How long will the exam last?

Full class period, if needed (75
minutes)



True/False

**Multiple
Choice**

Solution Time!

Problem-Solving Questions

Consider the following word embeddings for *pumpkin*, *candy*, and *halloween*.

Compute the cosine similarities between (a) *pumpkin* and *halloween*, and (b) *candy* and *halloween*, to determine whether *pumpkin* or *candy* is closer to *halloween*. If you want, you can leave the final values as fractions rather than converting them to decimals.

	W1	W2	W3
pumpkin	2	1	2
candy	2	2	1
halloween	1	2	1

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$$\cos(x,y) = \frac{x \cdot y}{|x||y|} = \frac{\sum_{i=1}^N x_i y_i}{\sqrt{\sum_{i=1}^N x_i^2} \sqrt{\sum_{i=1}^N y_i^2}}$$

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$$\cos(\text{pumpkin}, \text{halloween}) = \frac{2 \cdot 1 + 1 \cdot 2 + 2 \cdot 1}{\sqrt{2^2 + 1^2 + 2^2} \sqrt{1^2 + 2^2 + 1^2}}$$

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$$\cos(\text{candy}, \text{halloween}) = \frac{2 \cdot 1 + 2 \cdot 2 + 1 \cdot 1}{\sqrt{2^2 + 2^2 + 1^2} \sqrt{1^2 + 2^2 + 1^2}} = \frac{7}{\sqrt{9} \cdot \sqrt{6}}$$

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document

Happy Halloween!

Tomorrow, Halloween candy will be on sale in the candy aisle.

Do you prefer Thanksgiving?

Problem Solving Questions

- Given the following set of documents, compute the TF*IDF for:
 1. Halloween, for the first document
 2. candy, for the second document
- Compute *tf* and *idf* without applying any smoothing operations or log functions. If you want, you can leave the final values as fractions rather than converting them to decimals.

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$$\text{TF*IDF} = \text{count}(\text{term}, \text{doc}) * (\# \text{ docs}) / (\# \text{ docs w/term})$$

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$$TF(\text{Halloween}, \text{doc1}) = 1$$

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$$\text{IDF}(\text{Halloween}) = 3/2$$

$$\text{TF*IDF}(\text{Halloween}, \text{doc1}) = 1 * 3/2 = 3/2$$

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$$TF(\text{Halloween}, \text{doc1}) = 1$$

$$IDF(\text{Halloween}) = 3/2$$

$$TF*IDF(\text{Halloween}, \text{doc1}) = 1 * 3/2 = 3/2$$

$$TF(\text{candy}, \text{doc2}) = 2$$

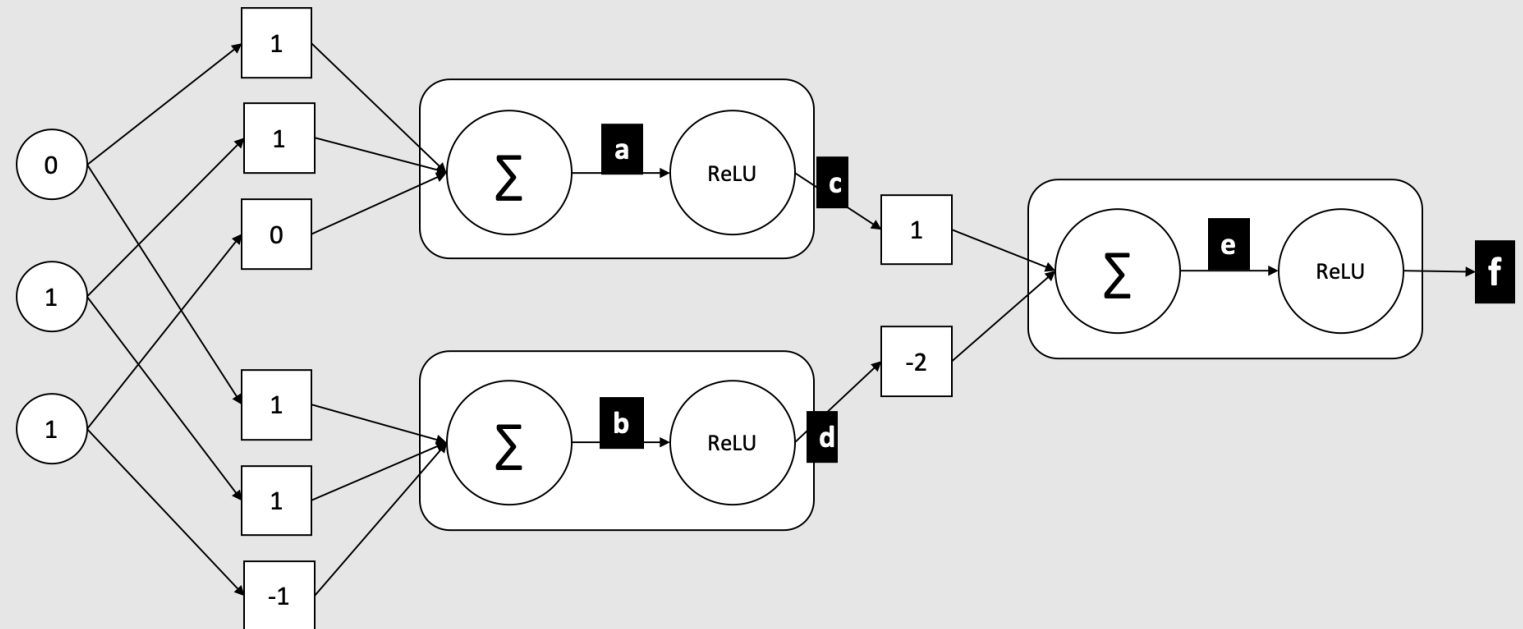
$$IDF(\text{candy}) = 3/1$$

$$TF*IDF(\text{candy}, \text{doc2}) = 2 * 3/1 = 6/1$$

Problem Solving Questions

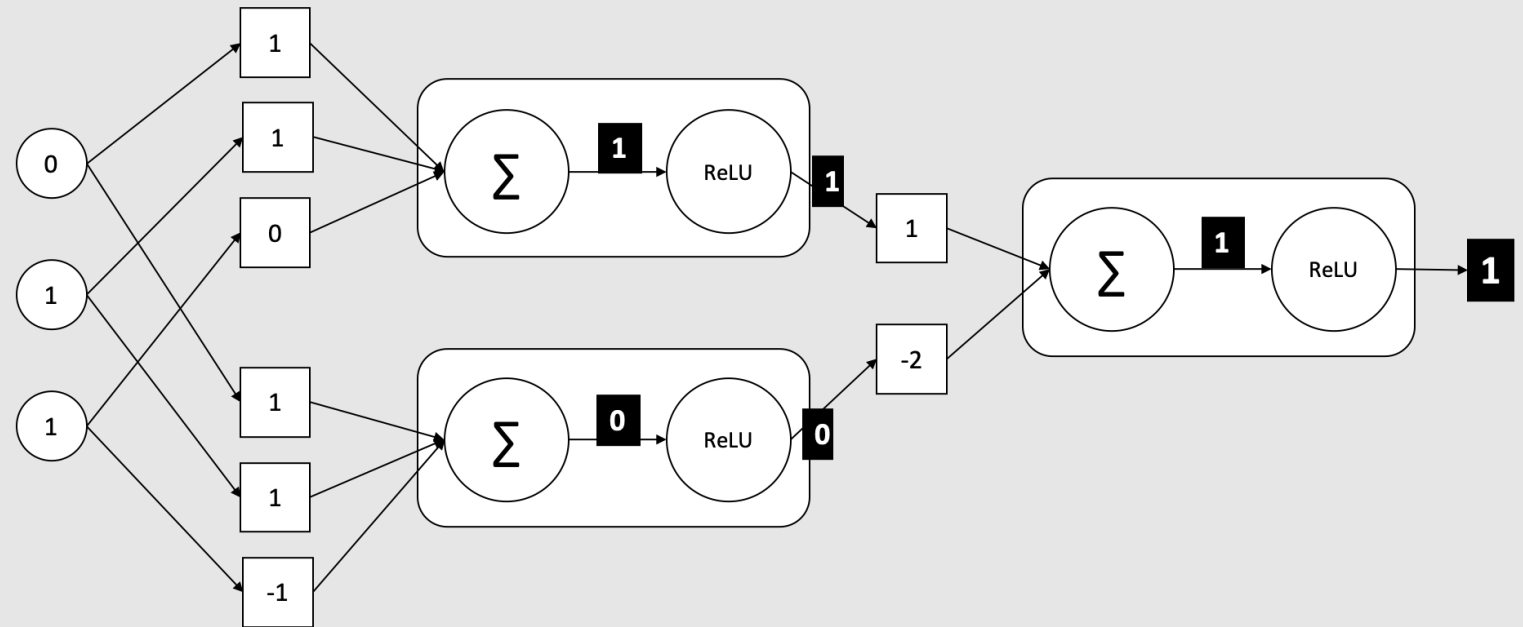
- Given the following set of documents, compute the TF*IDF for:
 1. Halloween, for the first document
 2. candy, for the second document
- Compute *tf* and *idf* without applying any smoothing operations or log functions. If you want, you can leave the final values as fractions rather than converting them to decimals.

Problem Solving Questions



- Given the feedforward neural network with the specified inputs, weights, and activation functions below, compute the network's intermediate (or final) values at the positions a , b , c , d , e , and f of the network that are specified.

Problem Solving Questions



- Given the feedforward neural network with the specified inputs, weights, and activation functions below, compute the network's intermediate (or final) values at the positions *a*, *b*, *c*, *d*, *e*, and *f* of the network that are specified.

Bonus Question

- Given the following contingency table, compute the microaveraged precision across all classes. If you want, you can leave the final values as fractions rather than converting them to decimals.

	Actual A	Actual B	Actual C
Predicted A	1	2	1
Predicted B	2	2	1
Predicted C	2	1	2

Bonus Question

- Given the following contingency table, compute the microaveraged precision across all classes. If you want, you can leave the final values as fractions rather than converting them to decimals.

	Actual A	Actual B	Actual C
Predicted A	1	2	1
Predicted B	2	2	1
Predicted C	2	1	2

	Actual A	Actual Not A
Predicted A	1	3
Predicted Not A	4	6

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	Actual A	Actual B	Actual C
Predicted A	1	2	1
Predicted B	2	2	1
Predicted C	2	1	2

	Actual A	Actual Not A
Predicted A	1	3
Predicted Not A	4	6

	Actual B	Actual Not B
Predicted B	2	3
Predicted Not B	3	6

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- Given the following contingency table, compute the microaveraged precision across all classes. If you want, you can leave the final values as fractions rather than converting them to decimals.

	Actual A	Actual B	Actual C
Predicted A	1	2	1
Predicted B	2	2	1
Predicted C	2	1	2

	Actual A	Actual Not A
Predicted A	1	3
Predicted Not A	4	6

	Actual B	Actual Not B
Predicted B	2	3
Predicted Not B	3	6

	Actual C	Actual Not C
Predicted C	2	3
Predicted Not C	2	7

Bonus Question

- Given the following contingency table, compute the microaveraged precision across all classes. If you want, you can leave the final values as fractions rather than converting them to decimals.

	Actual A	Actual B	Actual C
Predicted A	1	2	1
Predicted B	2	2	1
Predicted C	2	1	2

	Actual A	Actual Not A		Actual x	Actual Not x
Predicted A	1	3			
Predicted Not A	4	6			
	Actual B	Actual Not B			
Predicted B	2	3	Predicted x	5	9
Predicted Not B	3	6			
	Actual C	Actual Not C			
Predicted C	2	3			
Predicted Not C	2	7			

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- Given the following contingency table, compute the microaveraged precision across all classes. If you want, you can leave the final values as fractions rather than converting them to decimals.

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Predicted A	1	2	1
Predicted B	2	2	1
Predicted C	2	1	2

	Actual A	Actual Not A
Predicted A	1	3
Predicted Not A	4	6

	Actual B	Actual Not B
Predicted B	2	3
Predicted Not B	3	6

	Actual C	Actual Not C
Predicted C	2	3
Predicted Not C	2	7

	Actual x	Actual Not x
Predicted x	5	9
Predicted Not x	9	19

$$\text{Precision} = \text{TP} / (\text{TP} + \text{FP}) = 5 / (5 + 9) = 5/14$$

A few remaining details.....



Exam Location: Same
classroom as always (TBH
180G)



Exam Time: Same time as
class (Tuesday from 9:30-
10:45 a.m.)

Good luck!
