

# CS540 Introduction to Artificial Intelligence

## Lecture 9

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Based on lecture slides by Jerry Zhu, Yingyu Liang, and Charles Dyer

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# Spam or Ham?

Admin

training data from SMS Spam Collection

Dataset.

- Scratch
- Q1
- Which one of the following messages is a spam?
  - A: Go until jurong point, crazy.. Available only in bugis n great world la e buffet... Cine there got a...
  - B: Ok lar... Joking wif u oni...
  - C: Free entry in 2a wkly comp to win FA Cup final tkts 21st May 2005. Text FA to 87121 to receive entr...
  - D: U dun say so early hor... Uc already then say...
  - E: Nah I don't think he goes to usf, he lives around here though

grade → P1, P2 ..

→ code, output → P1S, P2S ..

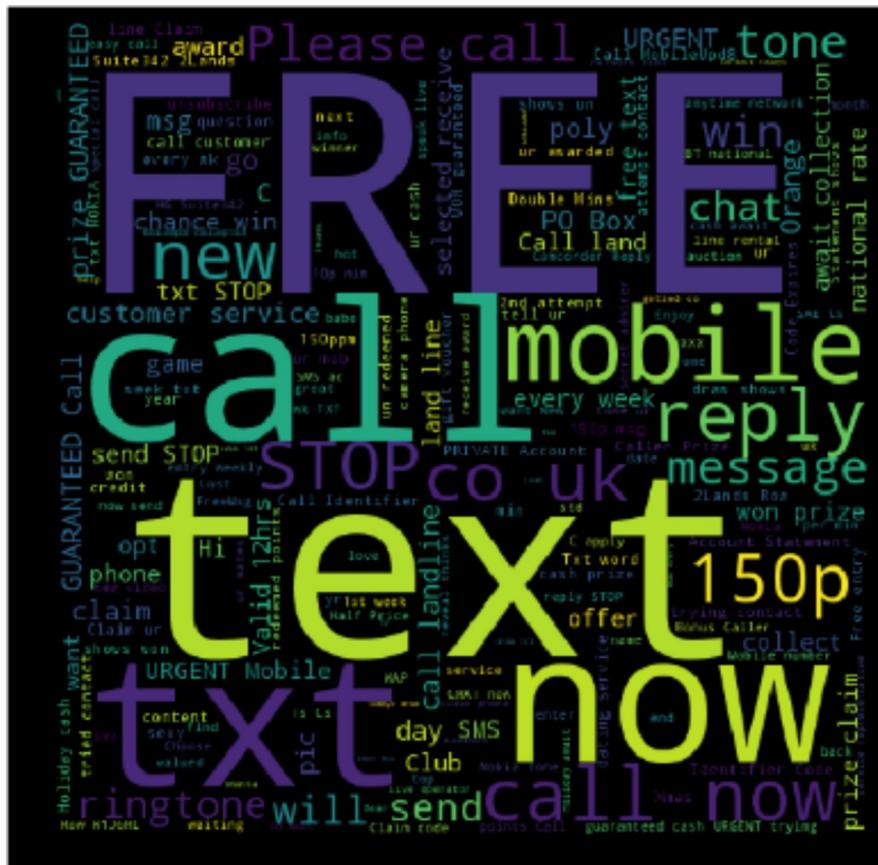
# Spam or Ham? Visualization

$\sqrt{P^2 - m^2}$

## Admin

ham

P3



# Bag of Words Features

## Definition

- Given a document  $i$  and vocabulary with size  $m$ , let  $c_{ij}$  be the count of the word  $j$  in the document  $i$  for  $j = 1, 2, \dots, m$ .
- Bag of words representation of a document has features that are the count of each word divided by the total number of words in the document.

$$x_{ij} = \frac{c_{ij}}{\sum_{j'=1}^m c_{ij'}}$$

# Bag of Words Features Example

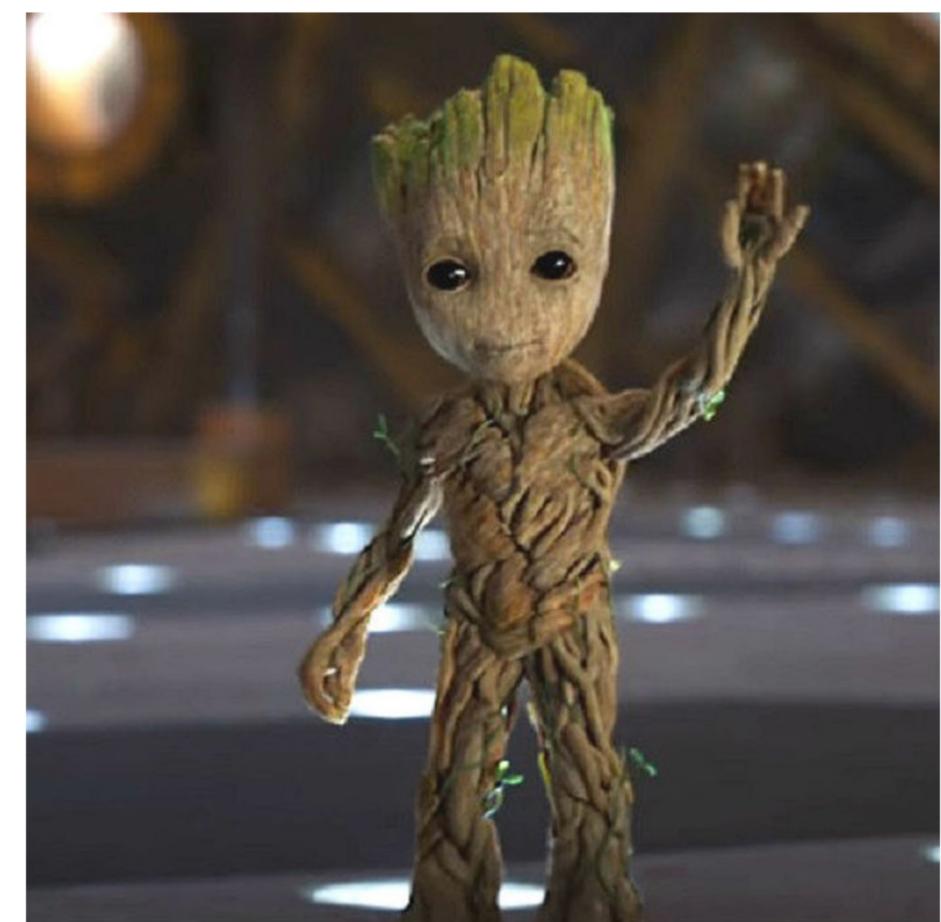
Definition

instance 1:

[ I am Groot ] am Groot --  
we are Groot  $\times 10$

Instance 2:

I am  
Groot



$x_1$

I  $(10/33)$   
am  $(10/33)$   
Groot  $(11/33)$   
we  $(1/33)$   
fine  $(1/33)$

$x_2$

$(1/3)$   
 $(1/3)$   
 $(1/3)$   
 $(0)$   
 $(0)$

## Unigram Model

Definition

- Unigram models assume independence.

$$\underbrace{\mathbb{P}\{z_1, z_2, \dots, z_d\}}_{\text{ }} = \prod_{t=1}^d \underbrace{\mathbb{P}\{z_t\}}_{\text{ }}$$

- In general, two events  $A$  and  $B$  are independent if:

$$\mathbb{P}\{A|B\} = \mathbb{P}\{A\} \text{ or } \mathbb{P}\{A, B\} = \mathbb{P}\{A\} \mathbb{P}\{B\}$$

- For sequence of words, independence means:

$$\underbrace{\mathbb{P}\{z_t | z_{t-1}, z_{t-2}, \dots, z_1\}}_{\text{ }} = \underbrace{\mathbb{P}\{z_t\}}_{\text{ }}$$

# Maximum Likelihood Estimation

## Definition

- $\mathbb{P}\{z_t\}$  can be estimated by the count of the word  $z_t$ .

$$\hat{\mathbb{P}}\{z_t\} = \frac{c_{z_t}}{m} \sum_{z=1}^m c_z$$

- This is called the maximum likelihood estimator because it maximizes the probability of observing the sentences in the training set.

# Bigram Model

## Definition

- Bigram models assume Markov property.

$$\underbrace{\mathbb{P}\{z_1, z_2, \dots, z_d\}}_{\text{Bigram model}} = \underbrace{\mathbb{P}\{z_1\}}_{\text{Initial state}} \prod_{t=2}^d \underbrace{\mathbb{P}\{z_t | z_{t-1}\}}_{\text{Transition probability}}$$

- Markov property means the distribution of an element in the sequence only depends on the previous element.

$$\mathbb{P}\{z_t | \underbrace{z_{t-1}, z_{t-2}, \dots, z_1}_{\text{Previous elements}}\} = \mathbb{P}\{z_t | z_{t-1}\}$$

# Bigram Model Estimation

## Definition

- Using the conditional probability formula,  $\hat{P}\{z_t | z_{t-1}\}$ , called transition probabilities, can be estimated by counting all bigrams and unigrams.

$$\hat{P}\{z_t | z_{t-1}\} = \frac{c_{z_{t-1}, z_t}}{c_{z_{t-1}}} \quad \text{total}$$
$$\Pr\{z_t | z_{t-1}\} = \frac{\Pr\{z_t, z_{t-1}\}}{\Pr\{z_{t-1}\}}$$

# Unigram MLE Probability 1

## Quiz

- Given the training data "I am Iron Man", "I love you 3000", "I love you mom", "Tell my family I love them", 18 words in total. With the unigram model, what is the probability of observing a new sentence "I love"?

$$\Pr[\{I\}] \cdot \Pr[\{\text{love}\}] \\ = \frac{4}{18} \cdot \frac{3}{18}$$

# Bigram MLE Probability 1

## Quiz

- Given the training data "I am Iron Man", "I love you 3000", "I love you mom", "Tell my family I love them", 18 words in total. With the bigram model, what is the probability of observing  $Z_2 = \text{"love"}$  given the sentence starts with  $Z_1 = \text{"I"}$ ?

$$\Pr \{ \text{love} | \text{I} \} = \frac{C_2 \text{ love}}{C_1} = \frac{3}{4}$$

# Unigram MLE Probability 2

## Quiz

Q2

- Given the training data "I am Groot am I", with the **unigram** model, what is the probability of observing a new sentence "I am I"?
- A:  $\frac{2}{5}$
- B:  $\frac{2}{25}$
- C:  $\frac{4}{25}$
- D:  $\frac{4}{125}$
- E:  $\frac{8}{125}$

$$\frac{2}{5} \cdot \frac{2}{5} \cdot \frac{2}{5} = \frac{8}{125}$$

# Bigram MLE Probability 2

## Quiz

- Given the training data "I am Groot am I", with the bigram model, what is the probability of observing a new sentence "I am I" given the first word is "I"?

A:  $\frac{1}{2}$

B:  $\frac{1}{4}$

C:  $\frac{1}{5}$

D:  $\frac{1}{10}$

E:  $\frac{4}{25}$

$P_{C \mid S_{\text{an}}} | I \rangle = \frac{C_{I \mid \text{an}}}{C_{\text{an}}} = \boxed{\frac{1}{1}} \text{ or } \frac{1}{2}$

 $P_{C \mid S_{\text{an}}} | I \rangle = \frac{C_{\text{an} \mid I}}{C_{\text{an}}} = \boxed{\frac{1}{2}}$ 

$I \text{ after an}$

Q3

# Bigram MLE Probability 3

## Quiz

- Given the training data "I am Groot am I", with the bigram model, what is the probability of observing a new sentence "I am Groot" given the first word is "I"?

- A:  $\frac{1}{2}$
- B:  $\frac{1}{4}$
- C:  $\frac{1}{5}$
- D:  $\frac{1}{10}$
- E:  $\frac{4}{25}$

Q4 -A

# Transition Matrix

## Definition

- These probabilities can be stored in a matrix called transition matrix of a Markov Chain. The number on row  $j$  column  $j'$  is the estimated probability  $\hat{P}\{j'|j\}$ . If there are 3 tokens  $\{1, 2, 3\}$ , the transition matrix is the following.

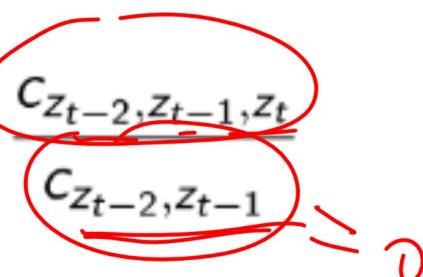

$$\begin{bmatrix} \hat{P}\{1|1\} & \hat{P}\{2|1\} & \hat{P}\{3|1\} \\ \hat{P}\{1|2\} & \hat{P}\{2|2\} & \hat{P}\{3|2\} \\ \hat{P}\{1|3\} & \hat{P}\{2|3\} & \hat{P}\{3|3\} \end{bmatrix}$$

- Given the initial distribution of tokens, the distribution of the next token can be found by multiplying it by the transition probabilities.

# Trigram Model

## Definition

- The same formula can be applied to trigram: sequences of three tokens.

$$\hat{\mathbb{P}} \{z_t | z_{t-1}, z_{t-2}\} = \frac{c_{z_{t-2}, z_{t-1}, z_t}}{c_{z_{t-2}, z_{t-1}}} \quad \text{0}$$


- In a document, it is likely that these longer sequences of tokens never appear. In those cases, the probabilities are  $\frac{0}{0}$ . Because of this, Laplace smoothing adds 1 to all counts.

$$\hat{\mathbb{P}} \{z_t | z_{t-1}, z_{t-2}\} = \frac{c_{z_{t-2}, z_{t-1}, z_t} + 1}{c_{z_{t-2}, z_{t-1}} + m}$$

# of words  
in vocab.

# Laplace Smoothing

## Definition

- Laplace smoothing should be used for bigram and unigram models too.

$$\hat{\mathbb{P}} \{z_t | z_{t-1}\} = \frac{c_{z_{t-1}, z_t} + 1}{c_{z_{t-1}} + m}$$

$$\hat{\mathbb{P}} \{z_t\} = \frac{c_{z_t} + 1}{\sum_{z=1}^m c_z + m}$$

- Aside: Laplace smoothing can also be used in decision tree training to compute entropy.

# Smoothing Example

## Quiz

- Fall 2018 Midterm Q12.
- Given a vocabulary of  $10^6$ , a document with  $10^{12}$  tokens with  $c_{\text{zoodles}} = 3$ . What is the MLE estimation of  $\mathbb{P}\{\text{zoodles}\}$  with and without Laplace smoothing?

$$\frac{c_z + 1}{\sum c_z' + m} = \frac{3 + 1}{10^{12} + 10^6}$$

# Smoothing Example 2

## Quiz

- Given a vocabulary of 5, a document with 30 words with  $c_{\text{Groot}} = \underline{10}$ . What is the MLE estimation of  $\mathbb{P}\{\text{Groot}\}$  with Laplace smoothing?

- A:  $\frac{1}{2}$
- B:  $\frac{11}{35}$
- C:  $\frac{1}{3}$
- D:  $\frac{11}{31}$
- E:  $\frac{1}{4}$

$$\frac{c_z + 1}{n + m}$$

$$\frac{10 + 1}{30 + 5}$$

$$\frac{c_z + 1}{\sum c_z + m}$$

Q5

↑  
# words in vocab.

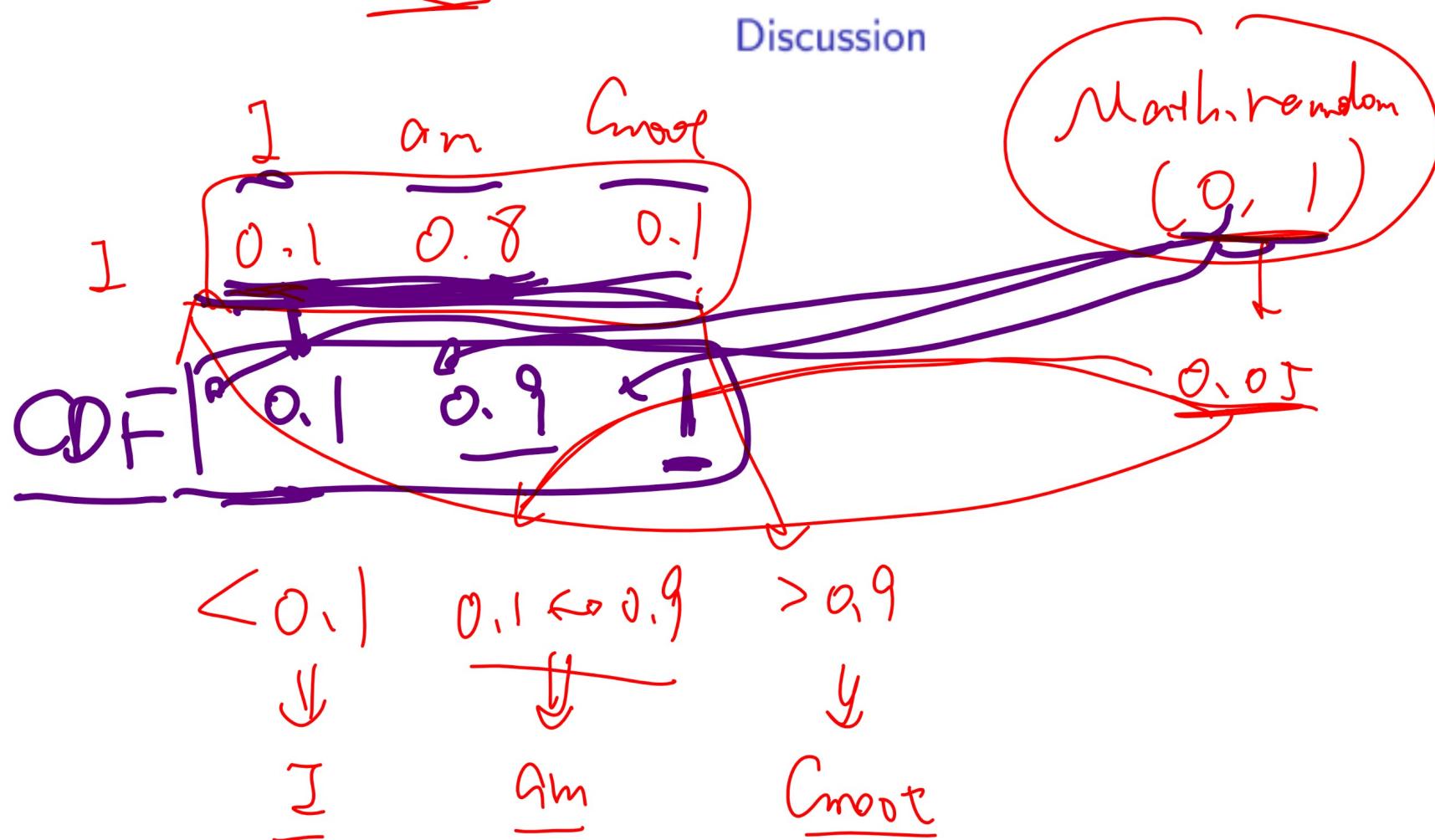
# Smoothing Example 3

## Quiz

- size*
- Given a vocabulary of 5, a document with 30 words with  $c_{I \text{ Groot}} = 0, c_I = 9, c_{\text{Groot}} = 10$ . What is the MLE estimation of  $\mathbb{P}\{\text{Groot} | I\}$  with Laplace smoothing?
  - A:  $\frac{1}{10}$
  - B:  $\frac{1}{11}$
  - C:  $\frac{1}{14}$
  - D:  $\frac{1}{15}$
  - E: 0
- Q6
- $$\frac{c_{I \text{ Groot}} + 1}{c_I + n} = \frac{0 + 1}{9 + 5} = \frac{1}{14}$$
- not overfit.

## CDF Inversion Method Diagram

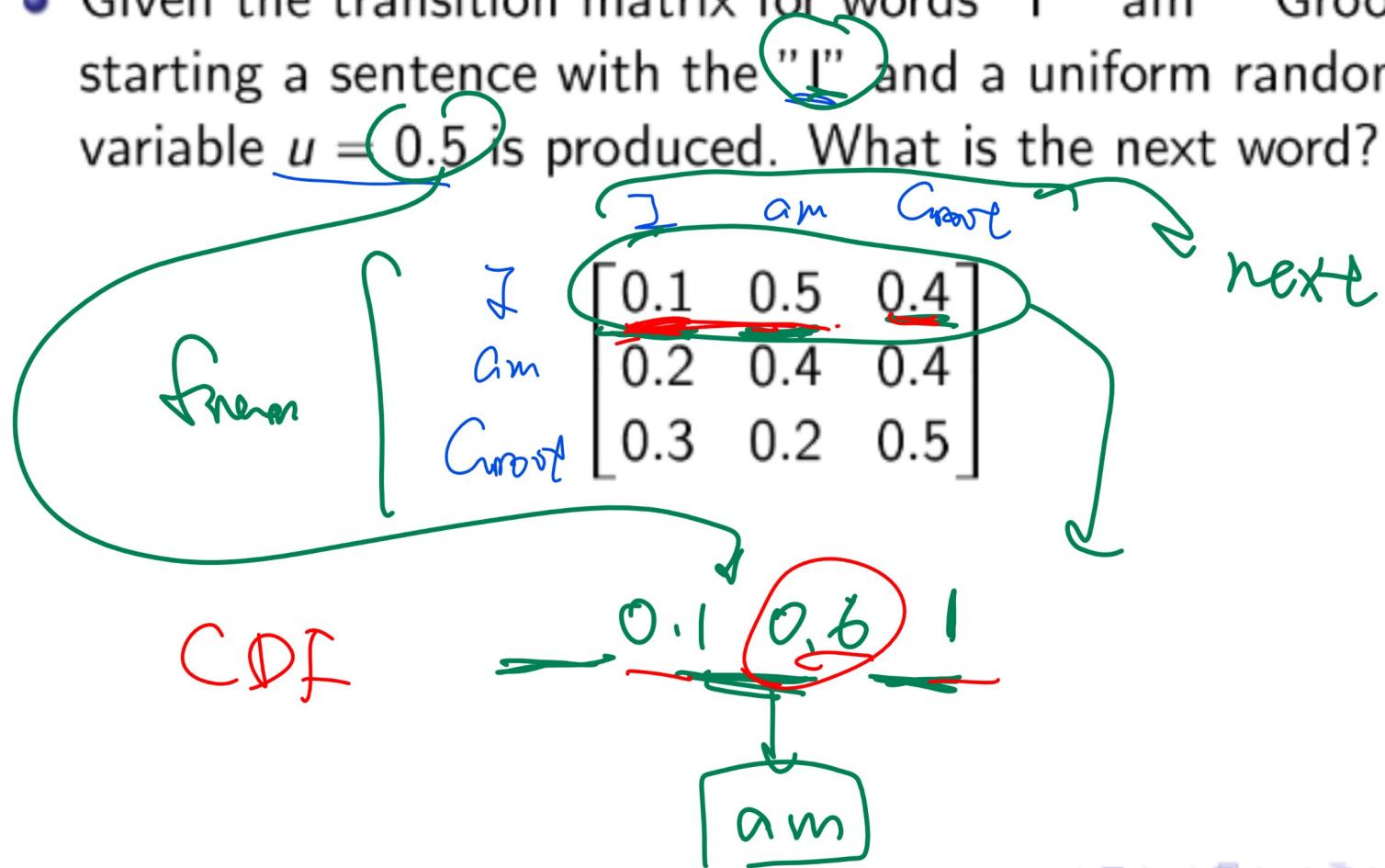
## Discussion



# Generating New Words 1

## Quiz

- Given the transition matrix for words "I" "am" "Groot", starting a sentence with the "I" and a uniform random variable  $u = 0.5$  is produced. What is the next word?



# Generating New Words 2

## Quiz

back @ 6:40

- Given the transition matrix for words "I" "am" "Groot", starting a sentence with the "am" and a uniform random variable  $u = 0.5$  is produced. What is the next word?

I	am	Groot
I	0.1 0.5 0.4	
am	0.2 0.4 0.4	
Groot	0.3 0.2 0.5	

Q7

- A: I, B: am, C: Groot

0.2 0.6 0

I	am	Groot	
I	1/3	1/3	1/3
am	0	0	1
Groot	0	1	0

