

# CS 107, Probability, Spring 2019

## Lecture 15

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- Repeated, Independent Trials: Multinomial Distribution
- Some Applications of the Conditional Probabilities

## The Secretary/Marriage Problem

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**Question:** When to stop?

**Answer:** Reject the first  $\frac{1}{e} \approx 0.368$  percent of the candidates, and then choose the first candidate who will be better than every applicant interviewed so far.

# Repeated Indep Trials: Multinomial Probabilities

## Multinomial Probabilities

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$$P_n(k_1, k_2, \dots, k_m) = \binom{n}{k_1, k_2, \dots, k_m} \cdot p_1^{k_1} \cdot p_2^{k_2} \cdot \dots \cdot p_m^{k_m},$$

where  $p_k = \mathbb{P}(A_k)$ ,  $k = 1, \dots, m$ .

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Note that in the binomial case, i.e., when  $m = 2$ , we get the Binomial Probabilities formula.

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- The probabilities of Simple Events:  $p_1 = \mathbb{P}(A_1) = \frac{2}{10} = \frac{1}{5}$ ,  $p_2 = \mathbb{P}(A_2) = \frac{5}{10} = \frac{1}{2}$ ,  $p_3 = \mathbb{P}(A_3) = \frac{3}{10}$ , and

$$p_1 + p_2 + p_3 = 1$$

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- We will not have any green balls shown? OTB



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Now, it remains to calculate this conditional probability.

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Explanation: for  $\mathbb{P}(\text{like } B | \text{like } A)$ , our "universe", new Sample Space consists of all persons who liked  $A$ . Among them we want to measure the probability of liking  $B$ .

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- Take into consideration the probability  $\mathbb{P}(\text{like } A)$  - if this is small, then the recommendation can be not effective (think about the diagnosis case!). So put a threshold on this probability, make predictions if  $\mathbb{P}(\text{like } A)$  is larger than some fixed value



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Say, when we ask Google some question, say, "what is the age of the universe", it gives <https://www.google.com/search?q=what+is+the+age+of+the+universe>

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- Yerekhaner@ partavor en mecanal.
- Angliayic lordur u lorduhiner ekel en vor aysor mer yerkrum iroq sirum en mer yerkir@ mer joghovrdin.

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Here the order matters, so we calculate the Probability of having the first word  $w_1$ , then followed by  $w_2$  and so on.

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Now, using the Conditional Probabilities, we can write

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- **Model 1: Unigram Model** - we assume that the words are independent. In this case

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Here the Probabilities can be calculated by (if we have a large corpus, large dataset of sentences):

$$\mathbb{P}(w_k) = \frac{\text{\#Sentences containing } w_k}{\text{\#All Sentences}}.$$

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- **Model 2: Bi-gram Model** - we assume that each words depends only on the preceding one, the previous word. In this case

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and

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Of course, I am oversimplifying things 😊

And you can consider 3-gram, n-gram models too!

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$$\mathbb{P}(w_k | w_1, w_2, \dots, w_{k-1}) = \mathbb{P}(w_k | w_{k-1}),$$

and this is the celebrated **Markov Chain Model**, which we will cover a little bit at the end of our course!