



Department of Computer Science  
UNIVERSITY OF COLORADO **BOULDER**



# Probability Distributions: Categorical and Poisson

Introduction to Data Science Algorithms

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## Categorical distribution

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- The **categorical** distribution generalizes Bernoulli distribution over any number of values
  - Rolling a die
  - Selecting a card from a deck
- AKA *discrete* distribution.
  - Most general type of discrete distribution
  - specify all (but one) of the probabilities in the distribution

## Categorical distribution

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- If the categorical distribution is over  $K$  possible outcomes, then the distribution has  $K$  parameters.
- We will denote the parameters with a  $K$ -dimensional vector  $\vec{\theta}$ .
- The probability mass function can be written as:

$$f(x) = \prod_{k=1}^K \theta_k^{[x=k]}$$

where the expression  $[x = k]$  evaluates to 1 if the statement is true and 0 otherwise.

- All this really says is that the probability of outcome  $x$  is equal to  $\theta_x$ .
- The number of *free parameters* is  $K - 1$ , since if you know  $K - 1$  of the parameters, the  $K$ th parameter is constrained to sum to 1.

## Categorical distribution

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- Example: the roll of an unweighted die

$$P(X = 1) = \frac{1}{6}$$

$$P(X = 2) = \frac{1}{6}$$

$$P(X = 3) = \frac{1}{6}$$

$$P(X = 4) = \frac{1}{6}$$

$$P(X = 5) = \frac{1}{6}$$

$$P(X = 6) = \frac{1}{6}$$

- If all outcomes have equal probability, this is called the *uniform* distribution.
- General notation:  $P(X = x) = \theta_x$

## Sampling from a categorical distribution

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- How to randomly select a value distributed according to a categorical distribution?
- The idea is similar to randomly selected a Bernoulli-distributed value.
- Algorithm:
  - 1 Randomly generate a number between 0 and 1  
 $r = \text{random}(0, 1)$
  - 2 For  $k = 1, \dots, K$ :
    - Return smallest  $r$  s.t.  $r < \sum_{i=1}^k \theta_i$

## Sampling from a categorical distribution

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- Example: simulating the roll of a die

$$P(X=1) = \theta_1 = 0.166667$$

$$P(X=2) = \theta_2 = 0.166667$$

$$P(X=3) = \theta_3 = 0.166667$$

$$P(X=4) = \theta_4 = 0.166667$$

$$P(X=5) = \theta_5 = 0.166667$$

$$P(X=6) = \theta_6 = 0.166667$$

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Random number in  $(0, 1)$ :

$r = 0.452383$

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$$r < \theta_1?$$



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Random number in  $(0, 1)$ :

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$$r < \theta_1?$$

$$r < \theta_1 + \theta_2?$$

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Random number in  $(0, 1)$ :

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$$r < \theta_1 + \theta_2?$$

$$r < \theta_1 + \theta_2 + \theta_3?$$

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Random number in  $(0, 1)$ :

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$$r < \theta_1?$$

$$r < \theta_1 + \theta_2?$$

$$r < \theta_1 + \theta_2 + \theta_3?$$

- Return  $X = 3$

## Sampling from a categorical distribution

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- Example 2: rolling a *biased* die

$$P(X=1) = \theta_1 = 0.01$$

$$P(X=2) = \theta_2 = 0.01$$

$$P(X=3) = \theta_3 = 0.01$$

$$P(X=4) = \theta_4 = 0.01$$

$$P(X=5) = \theta_5 = 0.01$$

$$P(X=6) = \theta_6 = 0.95$$

## Sampling from a categorical distribution

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Random number in  $(0, 1)$ :  
 $r = 0.209581$

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$$P(X=6) = \theta_6 = 0.95$$

Random number in  $(0, 1)$ :

$$r = 0.209581$$

$$r < \theta_1?$$

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Random number in  $(0, 1)$ :

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$$r < \theta_1?$$

$$r < \theta_1 + \theta_2?$$

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$$r < \theta_1?$$

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$$r < \theta_1?$$

$$r < \theta_1 + \theta_2?$$

$$r < \theta_1 + \theta_2 + \theta_3?$$

$$r < \theta_1 + \theta_2 + \theta_3 + \theta_4?$$

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$$r < \theta_1 + \theta_2 + \theta_3 + \theta_4 + \theta_5?$$

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Random number in  $(0, 1)$ :

$$r = 0.209581$$

$$r < \theta_1?$$

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$$r < \theta_1 + \theta_2 + \theta_3 + \theta_4?$$

$$r < \theta_1 + \theta_2 + \theta_3 + \theta_4 + \theta_5?$$

$$r <$$

$$\theta_1 + \theta_2 + \theta_3 + \theta_4 + \theta_5 + \theta_6?$$

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Random number in  $(0, 1)$ :

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$$r < \theta_1?$$

$$r < \theta_1 + \theta_2?$$

$$r < \theta_1 + \theta_2 + \theta_3?$$

$$r < \theta_1 + \theta_2 + \theta_3 + \theta_4?$$

$$r < \theta_1 + \theta_2 + \theta_3 + \theta_4 + \theta_5?$$

$$r <$$

$$\theta_1 + \theta_2 + \theta_3 + \theta_4 + \theta_5 + \theta_6?$$

- Return  $X = 6$

- We will always return  $X = 6$  unless our random number  $r < 0.05$ .
  - 6 is the most probable outcome

## Multinomial distribution

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- Recall: the binomial distribution is the number of successes from multiple Bernoulli success/fail events
- The **multinomial** distribution is the number of different outcomes from multiple *categorical* events
  - It is a generalization of the binomial distribution to more than two possible outcomes
  - As with the binomial distribution, each categorical event is assumed to be independent
  - **Bernoulli : binomial :: categorical : multinomial**
- Examples:
  - The number of times each face of a die turned up after 50 rolls
  - The number of times each suit is drawn from a deck of cards after 10 draws

## Multinomial distribution

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- Notation: let  $\vec{X}$  be a vector of length  $K$ , where  $X_k$  is a random variable that describes the number of times that the  $k$ th value was the outcome out of  $N$  categorical trials.
  - The possible values of each  $X_k$  are integers from 0 to  $N$
  - All  $X_k$  values must sum to  $N$ :  $\sum_{k=1}^K X_k = N$
- Example: if we roll a die 10 times, suppose it comes up with the following values:  
 $\vec{X} = \langle 1, 0, 3, 2, 1, 3 \rangle$ 
  - $X_1 = 1$
  - $X_2 = 0$
  - $X_3 = 3$
  - $X_4 = 2$
  - $X_5 = 1$
  - $X_6 = 3$
- The multinomial distribution is a *joint* distribution over multiple random variables:  $P(X_1, X_2, \dots, X_K)$

## Multinomial distribution

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- Suppose we roll a die 3 times. There are 216 ( $6^3$ ) possible outcomes:

$$P(111) = P(1)P(1)P(1) = 0.00463$$

$$P(112) = P(1)P(1)P(2) = 0.00463$$

$$P(113) = P(1)P(1)P(3) = 0.00463$$

$$P(114) = P(1)P(1)P(4) = 0.00463$$

$$P(115) = P(1)P(1)P(5) = 0.00463$$

$$P(116) = P(1)P(1)P(6) = 0.00463$$

...

...

...

$$P(665) = P(6)P(6)P(5) = 0.00463$$

$$P(666) = P(6)P(6)P(6) = 0.00463$$

- What is the probability of a particular vector of counts after 3 rolls?



## Multinomial distribution

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- Example 1:  $\vec{X} = \langle 0, 1, 0, 0, 2, 0 \rangle$

## Multinomial distribution

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- Example 1:  $\vec{X} = \langle 0, 1, 0, 0, 2, 0 \rangle$ 
  - $P(\vec{X}) = P(255) + P(525) + P(552) = 0.01389$

## Multinomial distribution

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- What is the probability of a particular vector of counts after 3 rolls?
- Example 1:  $\vec{X} = \langle 0, 1, 0, 0, 2, 0 \rangle$ 
  - $P(\vec{X}) = P(255) + P(525) + P(552) = 0.01389$
- Example 2:  $\vec{X} = \langle 0, 0, 1, 1, 1, 0 \rangle$

## Multinomial distribution

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- What is the probability of a particular vector of counts after 3 rolls?
- Example 1:  $\vec{X} = \langle 0, 1, 0, 0, 2, 0 \rangle$ 
  - $P(\vec{X}) = P(255) + P(525) + P(552) = 0.01389$
- Example 2:  $\vec{X} = \langle 0, 0, 1, 1, 1, 0 \rangle$ 
  - $P(\vec{X}) = P(345) + P(354) + P(435) + P(453) + P(534) + P(543) = 0.02778$

## Multinomial distribution

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- The probability mass function for the multinomial distribution is:

$$f(\vec{x}) = \frac{N!}{\underbrace{\prod_{k=1}^K x_k!}_{\text{Generalization of binomial coefficient}}} \prod_{k=1}^K \theta_k^{x_k}$$

- Like categorical distribution, multinomial has a  $K$ -length parameter vector  $\vec{\theta}$  encoding the probability of each outcome.
- Like binomial, the multinomial distribution has a additional parameter  $N$ , which is the number of events.

## Multinomial distribution: summary

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- Categorical distribution is multinomial when  $N = 1$ .
- Sampling from a multinomial: same code repeated  $N$  times.
  - Remember that each categorical trial is independent.
  - Question: Does this mean the count values (i.e., each  $X_1$ ,  $X_2$ , etc.) are independent?

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    - No! If  $N = 3$  and  $X_1 = 2$ , then  $X_2$  can be no larger than 1 (must sum to  $N$ ).

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  - Question: Does this mean the count values (i.e., each  $X_1$ ,  $X_2$ , etc.) are independent?
    - No! If  $N = 3$  and  $X_1 = 2$ , then  $X_2$  can be no larger than 1 (must sum to  $N$ ).
- Remember this analogy:
  - **Bernoulli : binomial :: categorical : multinomial**