

# **Statistical Sciences**

# DoSS Summer Bootcamp Probability Module 3

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# Recap

#### Learnt in last module:

- Independence of events
  - ▶ Pairwise independence, mutual independence
  - ▷ Conditional independence
- Random variables
- Distribution functions
- Density functions and mass functions
- Independence of random variables



### **Outline**

- Discrete probability

  - Combinatorics
  - Common discrete random variables
- Continuous probability
  - ▶ Geometric probability
- Exponential family



### Example:

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- Roll a die,  $P(\{1\}) = 1/6$

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### Classical probability

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### Example:

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- Roll a die,  $P(\{1\}) = 1/6$

### Classical probability

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#### Remark:

For some event  $A \in \mathcal{A}$ ,  $\mathbb{P}(A)$  can be computed as the proportion:

$$\mathbb{P}(A) = \frac{\#\{\text{outcomes that satisfies } A\}}{\#\{\text{all the possible outcomes}\}}$$



### Converting the probability into counting problems

### Permutations

For balls numbered 1 to n, choose r of them without replacement and record the order, the number of all the possible arrangements is

$$P(n,r) = n(n-1)\cdots(n-r+1) = \frac{n!}{(n-r)!}$$

#### Remark:

Order matters.

(1,2) and (2,1) are considered different.



### Converting the probability into counting problems

#### **Combinations**

For balls numbered 1 to n, choose r of them without replacement regardless the order, the number of all the possible arrangements is

$$\binom{n}{r} = C_r^n = \frac{\binom{n!}{r!(n-r)!}}{\binom{n!}{r!}} = \frac{\binom{n!}{r(n,r)}}{\binom{n!}{r!}}$$

Remark:

Order does not matter.

(1,2) and (2,1) are considered the same.

There we ignore order

r!



### **Examples:**

Forming committees: A school has 600 girls and 400 boys. A committee of 5 members is formed. What is the probability that it will contain exactly 4 girls?



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### Hypergeometric distribution

Randomly sample n objects without replacement from a source which contains a successes and N-a failures, denote X as the number of successes. Then

$$\mathbb{P}(X=x) = \frac{\binom{a}{x}\binom{N-a}{n-x}}{\binom{N}{n}}.$$



7 / 25

#### Common discrete random variables

### Bernoulli distribution

$$\Omega = \{ \mathsf{failure}, \, \mathsf{success} \}, \, X : \Omega \to \{0,1\}, \, \mathsf{and} \,$$

$$\mathbb{P}(X=1)=\rho, \quad \mathbb{P}(X=0)=1-\rho.$$

Write  $X \sim Bernoulli(p)$ .  $\times \sim \beta_{em}(p)$ .





#### Common discrete random variables

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$$\mathbb{P}(X=1)=p, \quad \mathbb{P}(X=0)=1-p.$$

Write  $X \sim Bernoulli(p)$ .

### **Example:**

- Toss a coin → p = 5
- Choose correct answer from A, B, C, D  $\rightarrow 4$



Common discrete random variables

### Binomial distribution

Consider n independent Bernoulli trials with success probability  $p \in (0,1)$ , denote the number of successes as X. Then X can take values in  $\{0, 1, \dots, n\}$ , and

$$\mathbb{P}(X=x) = \binom{n}{x} p^{(1-p)} (1-p)^{(-x)}$$

Write  $X \sim B(n, p)$ .



#### Common discrete random variables

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$$\mathbb{P}(X=x) = \binom{n}{x} p^{x} (1-p)^{n-x}.$$

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### **Example:**

- Toss a coin 100 times
- Choose correct answer from A, B, C, D for 20 questions



#### Common discrete random variables

### Geometric distribution

Keep doing independent Bernoulli trials with success probability  $p \in (0,1)$  until the first success happens. Denote the number of trials as X. Then X can take values in  $\{1, \dots, \infty\}$ , and

$$\mathbb{P}(X=x)=p(1-p)^{x-1}$$

Write  $X \sim Geo(p)$ .





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### **Example:**

- Toss a coin until the first head
- Choose answers from A, B, C, D until the first correct answer is picked



#### Common discrete random variables

### Negative binomial distribution

Keep doing independent Bernoulli trials with success probability  $p \in (0,1)$  until the first r success happens. Denote the number of trials as X. Then X can take values in  $\{r, \dots, \infty\}$ , and

$$\mathbb{P}(X=x) = \left(\begin{pmatrix} x-1\\r-1\end{pmatrix}\right) p^r (1-p)^{x-r}.$$

order

Write  $X \sim \text{Neg-bin}(r, p)$ .



#### Common discrete random variables

### Negative binomial distribution

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$$\mathbb{P}(X=x) = \binom{x-1}{r-1} p^r (1-p)^{x-r}.$$

Write  $X \sim \text{Neg-bin}(r, p)$ .

### **Example:**

- Toss a coin until the first 10 heads
- Choose answers from A. B. C. D until the first 3 correct answers are picked



#### Common discrete random variables

### Poisson distribution

Events occur in a fixed interval of time or space with a known constant mean rate  $\lambda$  and independently of the time since the last event, then denote the number of events during the fixed interval as X,

$$\mathbb{P}(X=x)=\frac{\lambda^x}{x!}exp(-\lambda).$$

Write  $X \sim Poisson(\lambda)$ .



12 / 25

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### **Example:**

fixed interval

- The number of patients arriving in an emergency room between 10 and 11 pm
- The number of laser photons hitting a detector in a particular time interval



#### Common discrete random variables

122: Bernoulli

#### Multinomial distribution

For n independent trials each of which leads to a success for exactly one of k categories, with each category having a given fixed success probability  $p_i$ ,  $i=1,\cdots,k$ , denote the number of successes of category i as  $X_i$ ,

$$\mathbb{P}(X_1 = x_1, \cdots, X_k = x_k) = \underbrace{\binom{n}{x_1 x_2 \cdots x_k}}_{p_1} p_1^{x_1} \cdots p_k^{x_k} \quad \text{with } \underbrace{\sum_{i=1}^k x_i = n, \sum_{i=1}^k p_i = 1.}_{p_i}$$

Write  $X \sim Multinomial(n, k, \{p_i\}_{i=1}^k)$ .

$$\begin{pmatrix} x_1 - x_2 \end{pmatrix} \stackrel{\text{def}}{=} \frac{n!}{|\vec{x_1}| \cdots |\vec{x_k}|}$$



#### Common discrete random variables

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#### Remark:

The multinomial distribution gives the probability of any particular combination of numbers of successes for the various categories.



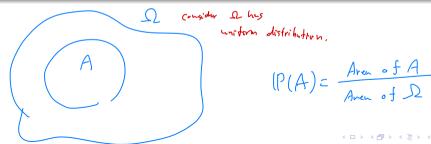
13 / 25

### Example:

- Throw a needle evenly on a circle, the probability that it will lie in the left half.
- A bus arrives every 4 minutes, the probability of waiting for less than 1 minute.

### Geometric probability

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14 / 25

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Common continuous random variables



## (Continuous) Uniform distribution

X takes values in a fixed interval (a, b) evenly,

$$\mathbb{P}(X \le x) = \frac{x - a}{b - a}, \quad a \le x \le b,$$

$$\text{derivative} \quad f(x) = \frac{1}{b - a}, \quad a \le x \le b.$$
Write  $X \sim U(a, b)$ . Unif  $(a, b)$ 

Remark:



#### Common continuous random variables

### Normal distribution

Define random variable X with the probability density function

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} exp(-\frac{(x-\mu)^2}{2\sigma^2})$$
 (2)

Write  $X \sim N(\mu, 6^2)$ .

#### Remark:

Most common distribution in nature 

CL7

#### Common continuous random variables

### Exponential distribution

Define random variable X with the probability density function

$$P(X \le x) = 1 - \exp(-\lambda x), x \ge 0$$
  
 
$$f(x) = \lambda \exp(-\lambda x), x \ge 0$$
 (3)

Write  $X \sim Exp(\lambda)$ .

#### Remark:



#### Common continuous random variables

### Cauchy distribution

Define random variable X with the probability density function

$$f(x;x_0,\gamma) = \frac{1}{\pi\gamma \left[1 + \left(\frac{x-x_0}{\gamma}\right)^2\right]} = \frac{1}{\pi\gamma} \left[\frac{\gamma^2}{(x-x_0)^2 + \gamma^2}\right]$$
(4)

Write  $X \sim Cauchy(x_0, \gamma)$ .

For 
$$\chi_{0}=0$$
,  $\chi=1$ 

$$= \frac{2}{\pi} \int_{-1+\chi^{2}}^{\infty} d\chi = \frac{1}{\pi} \cdot \left[ \log \left( (t\chi^{2}) \right]_{\chi=0}^{\infty} = \infty \right]$$

#### Common continuous random variables

#### Gamma distribution

Define random variable X with the probability density function

$$f(x; \alpha, \beta) = \frac{x^{\alpha - 1} e^{-\beta x} \beta^{\alpha}}{\Gamma(\alpha)} \quad \text{for } x > 0 \quad \alpha, \beta > 0,$$

$$\text{pereneters} \quad \text{normalizing factor so that} \quad \int_{0}^{\infty} P^{\alpha} dx = 1.$$
(5)

Write  $X \sim \Gamma(\alpha, \beta)$ .

Remark: 
$$P(d) = \int_{0}^{\infty} x^{d-1} e^{-Cx} p^{d} dx$$
 change of vertile  $p = 1$ 

Properties of gamma function

$$\cdot \quad |^{?} \left( \frac{1}{2} \right) = \sqrt{\pi}$$

#### Common continuous random variables

#### Beta distribution

Define random variable X with the probability density function

$$f(x; \alpha, \beta) = \frac{1}{B(\alpha, \beta)} x^{\alpha - 1} (1 - x)^{\beta - 1} \quad \text{for } 0 < x < 1 \quad \alpha, \beta > 0,$$

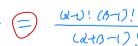
$$P = Beta(\alpha, \beta).$$

$$\text{Normalizy factor is that}$$

$$\int_{0}^{1} f(x) \, dx = 1.$$

Write  $X \sim Beta(\alpha, \beta)$ .

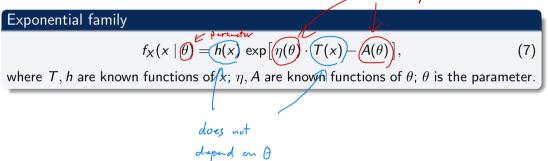
Remark: 
$$B(\alpha,\beta) = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx. = \frac{[\text{Pa}] [\text{Pa}]}{[\text{Pa}] [\text{Pa}]} = \frac{(\text{Pa})! (\text{Pa})!}{(\text{Pa})!}$$



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Consider a set of probability distributions whose pmf (discrete case) or pdf (continuous case) can be expressed in a certain form:





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### Exponential family

$$f_X(x \mid \theta) = h(x) \exp[\eta(\theta) \cdot T(x) - A(\theta)],$$
 (7)

where T, h are known functions of x;  $\eta$ , A are known functions of  $\theta$ ;  $\theta$  is the parameter.

#### Merits:

- Facilitate the computation of some properties
- Bayesian statistics: conjugate prior
- Regression: GLM



### Common distributions in the exponential family:

- Bernoulli / Binomial
- Poisson
- Negative Binomial
- Multinomial
- Exponential
- Normal
- Gamma
- Beta



### Show that Bernoulli distribution belongs to the exponential family:

1 lexp 
$$\left(\begin{array}{c} \log \frac{p}{1-p} \cdot \times - \left(-\log(1-p)\right) \\ \uparrow \\ h(x) \end{array}\right)$$

### **Problem Set**

**Problem 1:** The Robarts library has recently added a new printer which turns out to be defective. The letter "U" has a 30% chance of being printed out as "V", and the letter "V" has a 10% chance of being printed out as "U". Each letter is printed out independently, and all other letters are always correctly printed.

The librarian uses "UNIVERSITY OF TORONTO" as a test phrase, and will make a complaint to the printer factory immediately after the third incorrectly printed test phrase, calculate the probability that there are fewer correctly printed phrases than incorrectly printed phrases when the complaint is made.



### **Problem Set**

**Problem 2:** Compute the mode of Negative binomial distribution with parameter r and p.

(Hint: consider  $\mathbb{P}(X = k + 1)/\mathbb{P}(X = k)$ )

**Problem 3:** Show that normal distribution belongs to the exponential family.

