#### **Last lecture**

Gaussian (normal) Distribution (Ch 3.6.2)

Example

The Central Limit Theorem and Gaussian Approximation (Ch 3.6.3)

- Definition
- CDF Approximation
- Examples

# Agenda

The Central Limit Theorem and Gaussian Approximation (Ch 3.6.3)

Examples

ML estimation for continuous RVs (Ch 3.7)

- Definition
- Examples

Functions of a random variable (Ch 3.8)

• Find CDF/ PDF of g(X) (Ch 3.8.1)

### **Example**

 $X \sim Bin(n = 1000, p = 0.5)$ , Using Gaussian approximation, find

$$K \text{ s.t. } P\{X \ge K\} \approx 0.01 = Q(2.325)$$

• 
$$\mu_X = 500$$
 ,  $\sigma_X = \sqrt{np(1-p)} = \sqrt{250} \approx 15.8$ 

• 
$$P\{X \geq K\} =$$

$$X = 15.82 + 500$$
 $Z = \frac{x - 500}{15.8}$ 

• What if n = 1000000?

$$P\{\frac{X-500}{15.8} \geq \frac{K-500}{15.8}\} \approx P\{Z \geq \frac{K-500-0.5}{15.8} \geq 0.01$$

$$K-500-0.5=2.325$$
  $K=537.26$ 

Mx = 5000000  $\sigma_{x} = \sqrt{1}M \times 0.5 \times 0.5 = 500$  K = 5000000 = 0.5 = 2.325 K = 50(16)

Example 
$$x - np \times Jn$$
 Side note  $P\{|\hat{p}-p| \leq \frac{\alpha}{2\sqrt{n}}\} \geq 1 - \frac{1}{\alpha^2}$ 

We want to estimate p with  $\hat{p} = \frac{X}{2}$ 

• Find  $P\{|\hat{p}-p|<\delta\}$  in terms of  $n, p, \delta$ , and  $\Phi$ 

• Find  $\delta$  w/ 99% confidence if p=0.5, n=1000. Given that

 $\Phi(2.58) \approx 0.995$ 

• What if p = 0.1?

$$P\{Z \leq S \frac{n}{P(I-P)}\} - P\{Z \leq -S \frac{n}{P(I-P)}\}$$

$$=\Phi(c)-\Phi(-c)$$

Ans, 
$$\Phi\left(\sqrt[3]{p_{1}p_{1}}\right) - \Phi\left(-\sqrt[3]{p_{1}p_{1}}\right) = 2\Phi\left(\sqrt[3]{p_{1}p_{1}}\right)$$
  
 $2\Phi(c) - 1 = 0.99$ ,  $\Phi(c) = 0.995$ 

$$\frac{3}{3} = \frac{1000}{5} = \frac{258}{5} = 0.04$$

$$\frac{3}{C^{2}} = \frac{1000}{1000} = \frac{1}{2.58} = 0.025$$

Midtern Related

Nov. 3rd 7 PM

HWK5 -- HWK8 --> Gaussian

X cover CLT & Est.



#### **Definition**

Recall for discrete RV X, given observation u, ML is to find  $\theta$ maximizing  $p_{\theta}(u)$ 

- But for continuous RV,  $p_{\theta}(u) = 0$  Instead, ML maximize  $f_{\theta}(u)$  because

• 
$$f_{\theta}(u) \approx \frac{1}{\epsilon} P\{u - \frac{\epsilon}{2} < X < u + \frac{\epsilon}{2}\}$$
  
•  $\hat{\theta}_{ML}(u) \triangleq argmax_{\theta} f_{\theta}(u)$ 

### **Example**

 $T \sim Exp(\lambda)$  where  $\lambda$  is unknown. We want to estimate  $\lambda$  with observation t

observation 
$$t$$
•  $f_T(t) = \lambda e^{-\lambda t}$ 

• Extrema happens at  $\frac{df_T(t)}{d\lambda} = 0$ 

$$e^{-\lambda t} + -(\lambda t)e^{-\lambda t} = 0$$

$$(1 - \lambda t)e^{-\lambda t} = 0$$

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#### **Example**

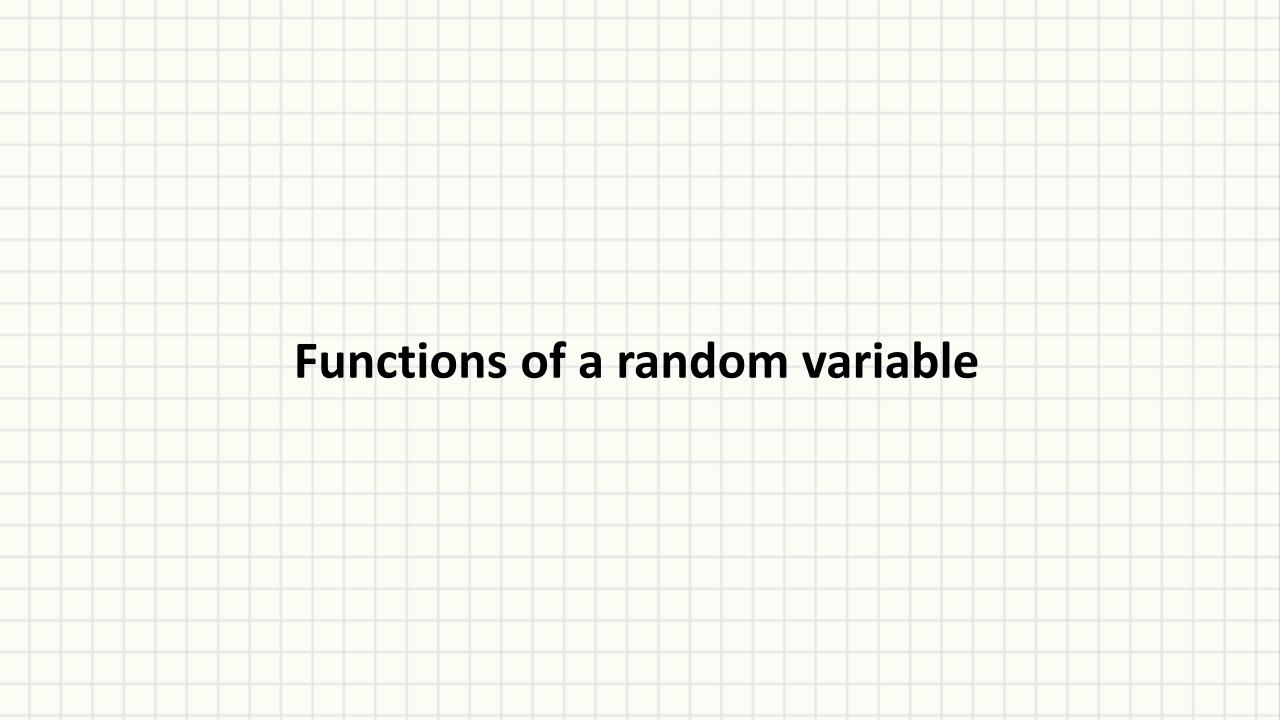
 $X \sim Uni([0,b])$  where b is unknown. We want to estimate b with

observation 
$$u$$

$$f_X(u) = \begin{cases} f & f \\ f & f \end{cases}$$

Extrema happens at

$$b = \mu \Rightarrow f_{\chi}(\mu) = \frac{1}{\mu}$$



# Find CDF/ PDF of g(X)

Motivation – I know X follows some distribution

- but what about Y = g(X)?  $f_Y$ ,  $u_Y$ ,  $\sigma_Y^2$
- 1. Scope the problem Find support of X and Y, are they continuous or discrete?  $\{C^2, f_X(C) > 0\}$
- 2. Find  $F_Y(c)$  from integrating  $f_X(x)$  over  $\{x: g(x) \le c\}$ 
  - If Y is discrete, normally we can find pmf  $p_Y(c)$
- 3. Get  $f_Y = F_Y'$

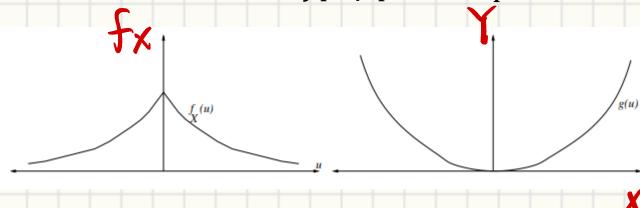
## **Examples**

- 1. Find support and continuity
- 2.  $F_Y(c) = \int_{x:f(x) \le c} f_X(x) dx$ 3.  $f_Y = F_Y'$

RV X follows  $f_X(u) = \frac{e^{-|u|}}{2}$  for  $u \in R$ .  $Y = X^2$ . Find  $f_Y$ ,  $\mu_Y$  and  $\sigma_Y^2$ 

1. 
$$F_Y =$$

2. 
$$P\{X \le c\} =$$



3.  $f_Y(c) =$ 

