

Lec 1

note: # : number

Let  $X$  be a r.v. Let  $x$  be a realization (data)  
data gathering process

$$x \in \underset{\text{Support}}{\text{Supp}[X]}$$

Discrete random variable  $X$  probability mass function

$$|\text{Supp}[X]| \leq \underset{\substack{\text{numbers of} \\ \text{elements}}}{N} \underset{\substack{\text{natural} \\ \text{\#}}}{\infty} \text{ countable infinity}$$

(PMF)

$$p(x) = P(X=x)$$

Definition:  $p: \mathbb{R} \rightarrow [0, 1]$

$$P: \text{supp}[X] \rightarrow [0, 1]$$

Cumulative Distribution function  
(cdf)

$$F(x) = P(X \leq x) = \sum p(y)$$

$$F: \mathbb{R} \rightarrow [0, 1] \quad \left\{ y: y \in \text{supp}(X) \text{ and } y \leq x \right\}$$

Continuous r.v.'s  $|\text{supp}(X)| = |\mathbb{R}|$  uncountable infinity

$$f(x) = \frac{d}{dx}(F(x)) \text{ pdf}$$

$$\text{if } a \leq b \quad P(X \in [a, b]) = F(b) - F(a) = \int_a^b f(x) dx$$

## Example of Discrete

①  $X \sim \text{Bern}(p)$  "Bernoulli" =  $\underbrace{p^x (1-p)^{1-x}}_{p(x)}$  pmf only 2 #  
 $p(x) = \begin{cases} p & x=1 \\ 1-p & x=0 \end{cases}$   $\text{Supp}(X) = \{0, 1\}$

②  $X \sim \text{Binom}(n, p) := \underbrace{\binom{n}{x} p^x (1-p)^{n-x}}_{\text{pmf}}$   $\text{Supp}(X) = \{0, 1, \dots, n\}$   $n+1$  #

"parameter" a thinking knob

For 1,  $p$  is the probability of success (if 1 is the def of success)  
 $p \in 0, 1$   $X \sim \text{Bern}(0) = \text{Deg}(0)$  "Degenerate"  $\hookrightarrow$  not random

$$X \sim \text{Bern}(1) = \text{Deg}(1) = \{ \text{always 1 with prob 1} \}$$

$p=0$  or  $p=1$  are degenerate cases, we exclude them from parameter space

parameter space: the set of all parameter values which are not degenerate.

Parameter space for Binomial  $p \in (0, 1)$   
 $n \in \{1, 2, \dots\} = \mathbb{N}$

Continuous r.v's

$X \sim \text{Exp}(\lambda) := \lambda e^{-\lambda x} f(x)$  exponential  
 $\text{Supp}(x) \in (0, \infty) \quad \lambda \in (0, \infty)$

$X \sim N(\mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(X-\mu)^2}{2\sigma^2}}$  normal r.v's  $\text{Supp}(x) = \mathbb{R}$   
 $\mu \in \mathbb{R} \quad \sigma^2 \in (0, +\infty) \quad \sigma^2 = \text{Var}(X) := E[(X-\mu)^2]$

Let  $\theta$  be the unknown parameter  
and  $\underbrace{\vec{\theta}}_{\text{vector}}, \dots, \dots$  parameters. and  $H$  be parameter space

$X \sim \text{Bern}(\theta) := \theta^x (1-\theta)^{1-x}$

$X \sim N(\theta_1, \theta_2)$  parameter model

$\underbrace{\vec{F}}_{\downarrow} := \left\{ \underbrace{P(X; \theta)}_{\text{pmf}} : \theta \in H \right\}$

space of parameter model

If  $X_1, X_2, \dots, X_n$  are r.v's

$P(X_1, \dots, X_n; \vec{\theta})$  is joint mass function (Jmf)  
joint density function (Jdf)

If  $X_1, \dots, X_n$  indep then

$$P(X_1, \dots, X_n; \vec{\theta}) = P(X_1, \vec{\theta}) \cdot P(X_2, \vec{\theta}) \cdot \dots \cdot P(X_n, \vec{\theta}) = \prod_{i=1}^n P_i(X_i, \vec{\theta})$$

if iid

$$P(X_1, \dots, X_n; \vec{\theta}) = \prod_{i=1}^n P(X_i, \vec{\theta})$$

In the real world, you observe "data"  $X = \langle 0, 0, 1, 0, 1, 0 \rangle$

Step I: pick  $\hat{\theta}$  pick a model, beyond the scope of this course.

Step II: Inference (figure out the value of  $\vec{\theta}$ )

The of inference

(I) point estimation provide a <sup>single</sup> guessing of  $\theta$

(II) Confident set provide a range of likely  $\theta$  values

(III) Theory testing test possibility that  $\theta$  is some value or belongs to some set.