

Lecture 1

01/28/20

Let X be a random variable \rightarrow data generating process
 Let x be a realization (data)

$$X \in \text{Supp}[X] = \{x: P(x) > 0\}$$

"Support"

Discrete Random Variables: $|\text{supp}(x)| \leq |\mathbb{N}|$

\nwarrow \downarrow
 Size (# of elements) natural #'s

Probability Mass function (PMF): $P(x) = P(X=x)$

Two definitions $P: \mathbb{R} \rightarrow [0, 1]$
 $P: \text{supp}(x) \rightarrow (0, 1]$

Cumulative Distribution Function (CDF)

$$F(x) = P(X \leq x) = \sum_{\{y: y \in \text{supp}(x) \text{ and } y \leq x\}} P(y)$$

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

Continuous R.V.'s

$|\text{supp}(x)| = |\mathbb{R}| \rightarrow$ uncountable infinite

CDF is defined the same as for discrete r.v.'s

$$f(x) = \frac{d}{dx} [F(x)], \quad \text{supp}(x) = \{x: f(x) > 0\}$$

Probability density function (PDF)

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By fundamental Thm. of calculus

$$P(X \in [a, b]) = F(b) - F(a) = \int_a^b f(x) dx$$

Examples of Discrete r.v.'s

$$(I) X \sim \text{Bern}(p) = \underbrace{p^x (1-p)^{1-x}}_{p(x)}, \text{Supp}(x) = \{0, 1\}$$

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

"Parameters" → numerical characteristics that define a distribution

For Bernoulli, p is the "probability of success" (If 1 is the definition of success)

$p \in (0, 1)$

$$X \sim \text{Bern}(0) = \text{Deg}(0) = \{0 \text{ w.p. } 1\}$$

Degenerate \Rightarrow r.v. that does NOT make things happen

$$X \sim \text{Bern}(1) = \text{Deg}(1) = \{1 \text{ w.p. } 1\}$$

$p=0$ or $p=1$ are degenerate cases. As a convention, exclude degenerate from the parameter space.

Parameter Space The set of all parameter values which are non degenerate

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

Continuous r.v. Examples

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

$$\lambda \in (0, \infty)$$

$$X \sim N(\mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \cdot e^{-\frac{1}{2\sigma^2}(x-\mu)^2} \quad \text{Supp}(x) = \mathbb{R}$$

$$\mu \in \mathbb{R}$$

$$\sigma^2 \in (0, \infty)$$

$$\text{Var}(x) = E[(x-\mu)^2]$$

Let θ be the unknown parameter and
 $\vec{\theta}$ be multiple unknown parameters and
 Θ "Capital Theta" be the parameter space

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

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

Parametric Model

$$\tilde{F} = \{ p(x; \theta) : \theta \in \Theta \}$$

\downarrow Space of parametric models \searrow PMF/PDF \downarrow Capital Theta

If x_1, x_2, \dots, x_n are random variables and $P(x_1, x_2, \dots, x_n; \vec{\theta})$ is the joint mass function / joint density function

If x_1, \dots, x_n iid

$$\Rightarrow P(x_1, x_2, \dots, x_n; \vec{\theta}) = P_1(x_1; \vec{\theta}) P_2(x_2; \vec{\theta}) \cdot \dots \cdot P_n(x_n; \vec{\theta})$$

$$= \prod_{i=1}^n P(x_i; \vec{\theta}) \quad \text{If } x_1, \dots, x_n \text{ iid} = \prod_{i=1}^n P(x_i; \vec{\theta})$$

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

Step I pick \tilde{F} . Pick a model. (Beyond the scope of this course)

Step II Inference (i.e. figure out the role of θ)

Three goals of Inference

- I) Point Estimation. Provide best single guess of θ
- II) Confidence Set. Provide a range of possible θ values
- III) Theory Testing. Test plausibility that θ is some value or belongs to some set