

Let X be a random variable \hookrightarrow data generating process

Let x be a realization (data).

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

Discrete r.v.'s X

$$|\text{Supp}[X]| \leq |\mathbb{N}| \quad \leftarrow \text{countable infinity.}$$

\uparrow
size (# of elements)

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)

$$F(x) := P(X \leq x) = \sum_{\substack{y: y \in \text{Supp}[X] \\ \text{and} \\ y \leq x}} p(y)$$

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

Continuous r.v.'s

$$|\text{Supp}[X]| = |\mathbb{R}| \quad \text{uncountable infinity}$$

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

$$f(x) := \frac{d}{dx} [F(x)] \quad \text{probability density function (PDF)} \quad \text{Supp}[X] := \{x: f(x) > 0\}$$

By Fundamental Thm of Calculus $a \leq b$,

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

Examples of Discrete r.v.'s

"Bernoulli"

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

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

"Parameters" A tuning knob

For Bernoulli, p is the "probability of success"

$p \in (0, 1)$ (if 1 is the definition of success).

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

"degenerate" are deterministic

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

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

Parameter Space the set of all parameter values which are not degenerate.

For Binomial $p \in (0, 1)$

$$n \in \{1, 2, \dots\} = \mathbb{N}$$

Continuous r.v. Examples

$$X \sim \text{Exp}(\lambda) := \lambda e^{-\lambda x}$$

$$\text{Supp}[X] = (0, \infty) \quad \text{fix}$$

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

$$X \sim \mathcal{N}(\mu, \sigma^2) = \frac{1}{\sqrt{2\pi} \sigma} \cdot e^{-\frac{1}{2\sigma^2}(x-\mu)^2}$$

$$\text{Supp}[X] = \mathbb{R}$$

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

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

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

Let θ be the unknown parameter.

And $\vec{\theta} = [\theta_1, \dots, \theta_k]$ parameters.

And Θ be the parameter space.

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

$$X \sim \mathcal{N}(\theta_1, \theta_2)$$

$\vec{\theta}$

Parametric Model

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

↑
space of
parameter-model

↑
PMF/
PDF

↑
PMF "needs"
the parameter value

If X_1, X_2, \dots, X_n are random variable $p(X_1, X_2, \dots, X_n; \vec{\theta})$ is the joint mass function / joint density function.

$$p(X_1, X_2, \dots, X_n; \vec{\theta}) \stackrel{\text{If } X_1, \dots, X_n \text{ ind (independent)}}{=} p_1(X_1; \vec{\theta}) \cdot p_2(X_2; \vec{\theta}) \cdot \dots \cdot p_n(X_n; \vec{\theta}) = \prod_{i=1}^n p_i(X_i; \vec{\theta})$$

$$\stackrel{\text{If } X_1, \dots, X_n \text{ iid} \rightarrow \text{independent \& identically distributed.}}{=} \prod_{i=1}^n p(X_i; \vec{\theta})$$

In the real world, you observe :

$$x = \langle 0, 0, 1, 0, 1, 0 \rangle$$

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

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

Three Goals of Inference

- (I) Point Estimation. Provide a best single guess of θ (denoted $\hat{\theta}$).
- (II) Confidence set. Provide range of likely θ
- (III) Theory Test. Test plausibility that θ is some value or belongs to some sets.