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Lecture 1

- A discrete random variable (r.v) has probability mass function (PMF)

$P(x) := P(X=x)$ and the r.v $X \sim P(x)$ where x is the "realized value"

- The Cumulative distribution function (CDF) is $F(x) := P(X \leq x)$ and the complementary CDF or "survival function" is $S(x) = 1 - F(x)$

This r.v has "support" given by $\text{support}[X] := \{x : p(x) > 0, x \in \mathbb{R}\}$

$|\text{support}[X]| \leq |\mathbb{N}|$ Countably infinite at most. Sets of this size are called "discrete" sets.
 \uparrow # element in a set

- The support and the PMF are related by the following identity:

$$\sum_{x \in \text{supp}[X]} P(x) = 1$$

- The most "fundamental" r.v is the Bernoulli:

$$X \sim \text{Bern}(p) := \frac{p^x (1-p)^{1-x}}{P(x)} \quad \text{with } \text{supp}[X] = \{0, 1\}$$

- Let defined the "indicator function"

$$\mathbb{1}_A := \begin{cases} 1 & \text{if } A \\ 0 & \text{if } A^c \end{cases} \Rightarrow X \sim \text{Bern}(p) := p^x (1-p)^{1-x} \mathbb{1}_{x \in \{0,1\}}$$

$$\Rightarrow \sum_{x \in \mathbb{R}} P(x) = 1$$

What if $p=1$?

$$X \sim \text{Bern}(1) = 1^x 0^{1-x} \mathbb{1}_{x \in \{0,1\}} = \mathbb{1}_{x=1} = \mathbb{1}_{x=1}$$

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$$X \sim \text{Deg}(1) = \{1\} \text{ w.p. } 1 \quad X \sim \text{Deg}(c) := \{1_{X=c}\}$$

↖ degenerate

$$X \sim \text{Bern}(0) = \text{Deg}(0)$$

• The convention in this class is that parameter values (p is the parameter of the Bernoulli) that yield degenerate RV's are not part of the legal "parameter space".

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

• If we have more than one RV x_1, x_2, \dots, x_n we can group them together in column vector:

$$\vec{x} = [x_1, x_2, \dots, x_n]^T \text{ and define the "joint mass function" (JMF) as: } P_{\vec{x}}^{(\vec{x})} = P_{x_1, \dots, x_n}^{(x_1, \dots, x_n)} \text{ valid for } \vec{x} \in \mathbb{R}^n \text{ and } \sum_{\vec{x} \in \mathbb{R}^n} P(\vec{x}) = 1$$

$$\forall x \cdot P_{x_1}^{(x_1)} = P_{x_2}^{(x_2)} = \dots = P_{x_n}^{(x_n)}$$

• If $x_1 \stackrel{d}{=} x_2 \stackrel{d}{=} \dots \stackrel{d}{=} x_n$ this denotes "equal in distribution" meaning their PMF's are the same. However, this offers no simplification of the JMF unless...

$x_1, x_2, \dots, x_n \stackrel{iid}{\sim}$ that means independent and identically distributed.

$$\Rightarrow P_{\vec{x}}^{(\vec{x})} = \prod_{i=1}^n P(x_i)$$

Let $x_1, x_2 \stackrel{iid}{\sim} \text{Bern}(p)$, let $T_2 = f(x_1, x_2) = x_1 + x_2 \sim P_T(t)$ denote

$$P_T^{(t)} = P_{x_1}^{(x_1)} * P_{x_2}^{(x_2)}$$

* convolution operator

$$\text{Supp}[T_2] = \{0, 1, 2\}$$