

Wed Aug. 26 (Lecture 01)

- A discrete random variable (r.v.) has a probability mass Function (pmf) $p(x) := p(X=x)$ and r.v. $X \sim p(x)$ where x is the realized value. X is the random variable, x is the state the random variable realizes to.
- The cumulative distribution Function (cdf) is $F(x) := p(X \leq x)$
- The survival Function, denoted by S , is defined as $S(x) := 1 - F(x) = P(X > x)$ (called the survival Function since it can be viewed as, "how long will it survive without being picked/occurring")
- This r.v. (X) has support given by $\text{supp}[X] := \{x \mid x \in \mathbb{R}, p(x) > 0\}$
- $|\text{supp}[X]| \leq \aleph_0$
 - Sets this size are called discrete sets
 - discrete r.v. \Rightarrow The number of things it can realize to is a discrete amount
- The pmf and the support are related by the following identity:

$$\sum_{x \in \text{supp}[X]} p(x) = 1 \quad \text{where each } x \in \text{supp}[X] \text{ is independent}$$

(The probability something happens is guaranteed)

The most fundamental r.v. is the Bernoulli r.v.

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

What about $p(7) = p^7 (1-p)^{-6}$??

$\hookrightarrow 7$ is not in the support. It is annoying to have to give 2 pieces of information (pmf and the support)

Let's define the indicator Function: $\mathbb{1}_A := \begin{cases} 1 & \text{if } A \\ 0 & \text{if } A^c \end{cases}$

$$\Rightarrow X \sim \text{Bern}(p) = \underbrace{p^x (1-p)^{1-x}}_{p(x)} \mathbb{1}_{x \in \{0,1\}} \quad \Rightarrow \sum_{x \in \mathbb{R}} p(x) = 1 \quad \text{since every } x \in \mathbb{R} \text{ is now defined}$$

(How this helps is instead of providing the support will be revealed next lecture)

What if $p=1$?

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

calling this a random variable when it is deterministic!?

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

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

$$X \sim \text{Deg}(c) = \mathbb{1}_{x=c}$$

The convention in this class is that parameter values (p is the parameter for the Bernoulli) that yield degenerate r.v.s are not part of the legal parameter space.

$\hookrightarrow p \in (0, 1)$ (otherwise something [not] happens)

If we have more than one r.v. X_1, X_2, \dots, X_n , we can group them together in a column vector: $\vec{X} := [X_1, X_2, \dots, X_n]^T$

and define the joint mass function as:

$$P_{\vec{X}}(\vec{x}) = P_{X_1, X_2, \dots, X_n}(x_1, \dots, x_n) \text{ valid for } \vec{x} \in \mathbb{R}^n \text{ and } \sum_{\vec{x} \in \mathbb{R}^n} P(\vec{x}) = 1$$

(multiple function pmf)

If X_1, X_2, \dots, X_n are independent, then

$$P_{\vec{X}}(\vec{x}) = P_{X_1}(x_1) \cdot P_{X_2}(x_2) \cdot \dots \cdot P_{X_n}(x_n) = \prod_{i=1}^n P_{X_i}(x_i) \quad \text{"multiplication rule"}$$

If $X_1 \stackrel{d}{=} X_2 \stackrel{d}{=} \dots \stackrel{d}{=} X_n \Rightarrow$ This denotes "equal distribution" meaning their pmf's are the same. However this offers no simplification of the pmf unless...

$\forall x, P_{X_1}(x) = P_{X_2}(x) = \dots = P_{X_n}(x)$ $X_1, X_2, \dots, X_n \stackrel{i.i.d.}{\sim}$ (independent and equally distributed)

$$\hookrightarrow P_{\vec{X}}(\vec{x}) = \prod_{i=1}^n p(x_i)$$

Let $X_1, X_2 \stackrel{i.i.d.}{\sim} \text{Bern}(p)$. Let $T_2 = f(X_1, X_2) = X_1 + X_2 \sim p_{T_2}(\cdot)$

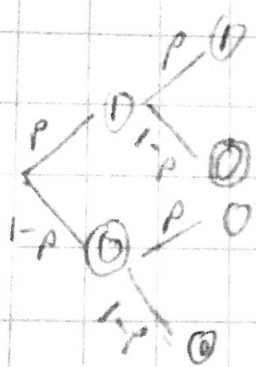
(Function of 2 r.v. and returns a r.v.)

Sometimes denoted as $p_{T_2}(t) = P_{X_1}(x) * P_{X_2}(x)$ \rightarrow convolution operator

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

0, 1 or 2 successes

(probability tree on the next page)



$P_{X_1, X_2} (x_1, x_2) \quad (I_{MF})$

P^2
 $P(1-P)$
 $P(1-P)$
 $(1-P)^2$

T

2
 1
 1
 0

$T \sim \begin{cases} 2 & w.p & P^2 \\ 1 & w.p & 2P(1-P) \\ 0 & w.p & (1-P)^2 \end{cases}$