

Math 621

Go over
lecture 1 and
2 on syllabus

Exam dates:

- 9/23
- 11/4
- 12/14

HW 20%.

C.P. 5%.

Exam 1 20%.

Exam 2 20%.

Final 35%.

- A discrete random variable (rv) has a prob. mass $f(x)$ (pmf).

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

$X, x \leftarrow$ state of
the random
variable.
 \uparrow
random
variable

- The cumm. distributive $f(x)$ (cdf) is
 $F(x) := P(X \leq x)$

and the complementary (cdf) or "survival function" is $S(x) := P(X > x) = 1 - F(x)$.

- This rv has "support" given by $\text{Supp}[X] := \{x : p(x) > 0, x \in \mathbb{R}\}$. (stuff that could be seen)

$|\text{Supp}[X]| \leq |\mathbb{N}|$ countingly infinite at most

\uparrow
of elts. in a set

Sets this size are called "discrete" sets.

The support and the pmf are related by the following identity:

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

The most "fundamental" rv is the Bernoulli:

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

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

Let's define the "indicator function"

$$\mathbb{1}_A := \begin{cases} 1 & \text{if } A \\ 0 & \text{if } A^c \end{cases}$$

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

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

• What if $p=1$?

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

$$\Downarrow$$

$$X \sim \text{Deg}(1) = \{1 \text{ w/probability } 1\}$$

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

• What if $X \sim \text{Bern}(0) = \text{Deg}(0)$.

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 rv 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 then define the "joint mass function" (jmf) as:

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

$$\text{and if } \sum_{\vec{x} \in \mathbb{R}^n} P(\vec{x}) = 1$$

If $X_1, X_2, X_3, \dots, X_n$ ~~What~~ are independent then,

$$P(\vec{x}) = P_{X_1}(x_1) P_{X_2}(x_2) \dots P_{X_n}(x_n) = \prod_{i=1}^n P_{X_i}(x_i)$$

↑
"the multiplication rule"

If $X_1 \stackrel{d}{=} X_2 \stackrel{d}{=} \dots \stackrel{d}{=} X_n$ this means "equally distributed" meaning their pmf are the same. However, this offers no simplification of the jmf unless...

$$P_{X_1}(x_1) = P_{X_2}(x_2) = \dots = P_{X_n}(x_n) \quad \forall x$$

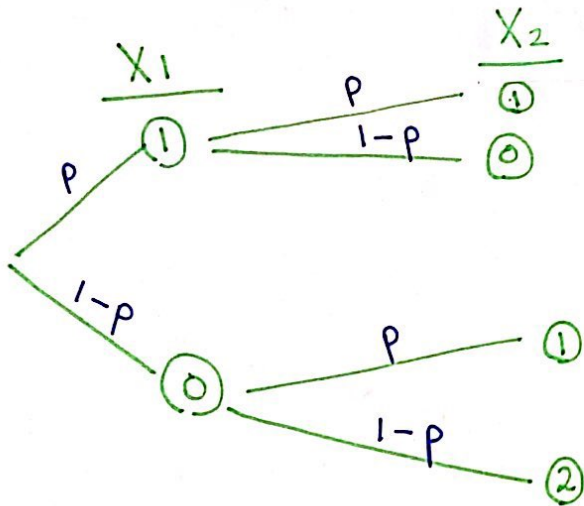
$X_1, X_2, X_3, \dots, X_n$ iid that means indep. & identically distributed

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

Let $X_1, X_2 \stackrel{iid}{\sim} \text{Bern}(p)$ and let $T_2 = f(X_1, X_2) = X_1 + X_2 \sim ?$
 denoted $P_{T_2}(t) = P_{X_1}^{(x)} * P_{X_2}^{(x)}$ $P_T(t)$

"called the convolution operator"

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



$$\frac{P_{X_1, X_2}(X_1, X_2)}{p^2}$$

$$\begin{aligned} &p(1-p) \\ &(1-p)p \\ &(1-p)^2 \end{aligned}$$