

Math 621
(Probability)

Tue - Thu

8:00pm - 9:15pm K: 258

Prof: Kapelner

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Lecture 1

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

$p(x) := P(X=x)$ and cumulative distribution function (CDF)

$F(x) := P(X \leq x)$. The r.v. X has "support"

$\text{supp}(X) := \{x \in \mathbb{R} : p(x) > 0\}$

since X is discrete $|\text{supp}(X)| \leq |\mathbb{N}|$

support and PMF are vectors, i.e.

$$\sum_{x \in \text{supp}(X)} p(x) = 1$$

The most fundamental discrete random variable is the **Bernoulli**

$$X \sim \text{Bern}(p) := \begin{cases} 1 & \text{w.p. } p \\ 0 & \text{w.p. } 1-p \end{cases}$$

what is p ?

p is a parameter, parameters have parameter spaces e.g. $p \in (0, 1)$

$p \neq 0$, and $p \neq 1$

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

"degenerate" indicator function

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

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

Independence

X_1, X_2 independent $\Leftrightarrow \forall s \quad X_1, X_2 \stackrel{\text{ind}}{\sim}$
 "X₁ and X₂ are independent"

By definition \rightarrow joint mass function

$$P(X_1, X_2) = P_{X_1}(x_1) P_{X_2}(x_2) \quad \forall x_1, x_2 \text{ in their support}$$

$X_1 \stackrel{d}{=} X_2$ the r.v.'s X_1, X_2 are equal in distribution if $P_{X_1}(x) = P_{X_2}(x)$

$X_1, X_2 \stackrel{\text{iid}}{\sim}$ the r.v.'s X_1, X_2 are independent and identically distributed

Def: $X_1, X_2 \stackrel{\text{iid}}{\sim}$ and $X_1 \stackrel{d}{=} X_2$

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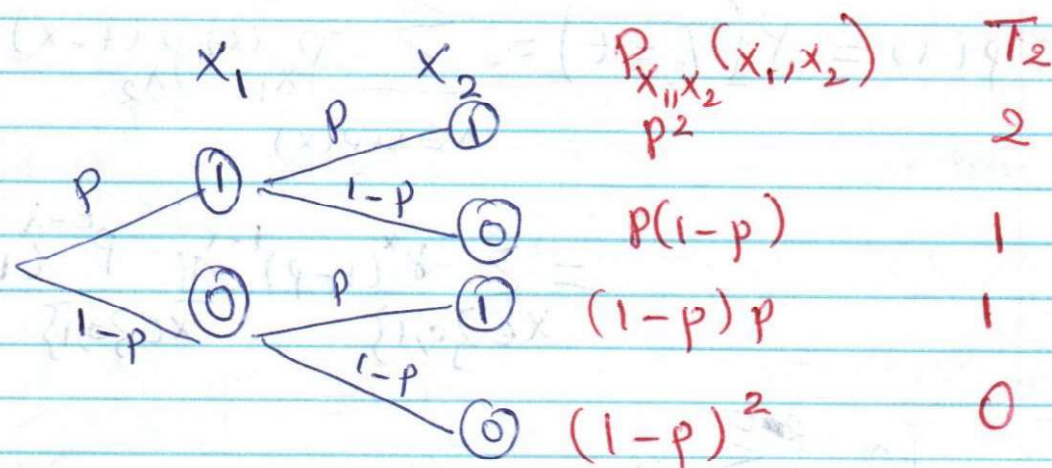
les $T_2 = X_1 + X_2$ where $X_1, X_2 \stackrel{iid}{\sim} \text{Bern}(p)$

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

$$= \text{sup}[X_1] + \text{sup}[X_2]$$

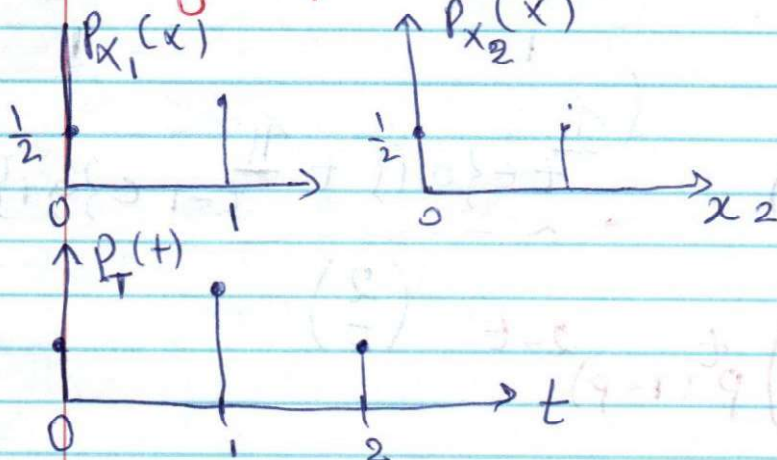
$$A+B = \{a+b : a \in A, b \in B\}$$

Probability tree



$$\Rightarrow P_{T_2} \cdot T_2 \left\{ \begin{array}{l} 2 \text{ up } p^2 \\ 0 \text{ up } (1-p)^2 \\ 1 \text{ up } 2p(1-p) \end{array} \right.$$

Imagine $p = 1/2$



$$P_{T_2}(t) =$$

$$P(T_2=t) = \sum_{x \in \text{sup}(X)} P_{X_1}(x) P_{X_2}(t-x)$$

$$T_2 = X_1 + X_2$$

Example: For 2 Bern

$$p(t) = P(T_2=t) = \sum_{x \in \text{sup}(x)} P_{X_1}(x) P_{X_2}(t-x)$$

$$= \sum_{x \in \{0,1\}} p^x (1-p)^{1-x} \prod_{x \in \{0,1\}} p^{t-x} (1-p)^{1-t+x} \prod_{(t-x) \in \{0,1\}}$$

$$= p^t (1-p)^{2-t} \sum_{x \in \{0,1\}} \prod_{x \in \{0,1\}} \prod_{t-x \in \{0,1\}}$$

$$= p^t (1-p)^{2-t} \left(\prod_{0 \in \{0,1\}} \prod_{t-0 \in \{0,1\}} + \prod_{1 \in \{0,1\}} \prod_{t-1 \in \{0,1\}} \right)$$

$$P(0) = (1-p)^2$$

$$P(2) = p^2$$

$$P(1) = 2p(1-p)$$

$$\left(\prod_{t \in \{0,1\}} + \prod_{t-1 \in \{0,1\}} \right)$$

$$\Rightarrow P(t) = \binom{2}{t} p^t (1-p)^{2-t} \quad \binom{2}{t}$$

(3)

$$X \sim \text{Bern}(p) = \text{Bin}(1, p) = \binom{1}{x} p^x (1-p)^{1-x}$$

note: $\binom{n}{k}$ is only valid if $k \leq n$ otherwise 0

$$\sum_{x \in \text{sup}[x_1]} p_{x_1}(x) p_{x_2}(x) = \sum_{x \in \{0,1\}} \binom{1}{x} p^x (1-p)^{1-x} \binom{1}{t-x} p^{t-x} (1-p)^{1-t+x}$$

$$p^t (1-p)^{2-t} \sum_{x \in \{0,1\}} \binom{1}{x} \binom{1}{t-x} = \binom{1}{0} \cdot \binom{1}{t} + \binom{1}{1} \cdot \binom{1}{t-1}$$

$$= \binom{2}{t} p^t (1-p)^{2-t}$$

$$P(t) = P(T_2 = t) = p_{x_1}(x) * p_{x_2}(x) := \sum_{x \in \text{sup}(x_1)} p_{x_1}(x) p_{x_2}(t-x)$$

$x_1, x_2, x_3 \stackrel{\text{iid}}{\sim} \text{Bern}(p)$

Convolution of two independent PMF's

$$T_3 = x_1 + x_2 + x_3 = x_3 + T_2 = p_{x_3}(x) * p_{T_2}(x)$$

$$\sum_{x \in \text{sup}[x_3]} p_{x_3}(x) p_{T_2}(t-x) =$$

$$\sum_{x \in \{0,1\}} \binom{1}{x} p^x (1-p)^{1-x} \binom{2}{t-x} p^{t-x} (1-p)^{2-t+x} =$$

$$= p^t (1-p)^{3-t} \sum_{x \in \{0,1\}} \binom{1}{x} \binom{2}{t-x}$$

$$= p^t (1-p)^{3-t} \left(\binom{2}{t} + \binom{2}{t-1} \right)$$

$$= \binom{3}{t} p^t (1-p)^{3-t}$$