

Lecture 3

Proof by induction

$X_1, \dots, X_n \stackrel{iid}{\sim} \text{Bern}(p)$

$T = \sum_{i=1}^n X_i \sim \text{neg Bin}(r, p) = \binom{t+r-1}{r-1} (1-p)^r p^{t-r}$

Diagram illustrating the negative binomial distribution as a sum of Bernoulli trials. A sequence of trials is shown with 1's and 0's. The first trial is 1, followed by two 0's, then a 1, and so on. A bracket under the first three trials (1, 0, 1) is labeled "t-trials". An arrow points from the first trial to the binomial coefficient $\binom{t+r-1}{r-1}$, and another arrow points from the first trial to the probability p^{t-r} .

$$X \sim \text{Bin}(n, p) := \binom{n}{x} p^x (1-p)^{n-x} = \binom{n}{x} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x}$$

Let n be large and p be small. $\lim_{n \rightarrow \infty} p(x) \Rightarrow$
 s.t. n, p are relevant v.d.

$$\lambda = np \Rightarrow p = \frac{\lambda}{n}$$

$$\lim_{n \rightarrow \infty} p(x) = \lim_{n \rightarrow \infty} \frac{n!}{x!(n-x)!} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x} = \frac{\lambda^x}{x!} \lim_{n \rightarrow \infty} \frac{n!}{(n-x)!} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x}$$

$$\lim_{n \rightarrow \infty} \frac{n!}{(n-x)!} = \lim_{n \rightarrow \infty} \frac{n(n-1)(n-2) \dots (n-x+1)}{1} = \lim_{n \rightarrow \infty} \frac{n^x}{1} = \lambda^x$$

$$\lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^n = e^{-\lambda}$$

$$\lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^{-x} = 1$$

$$\lim_{n \rightarrow \infty} p(x) = \frac{\lambda^x e^{-\lambda}}{x!} = \text{Poisson}(\lambda)$$

Recall convolution formula $P(t) = \sum_{x \in \text{Supp}[X]} P_{X_1}(x) P_{X_2}(t-x)$ \parallel $t-x \in \text{Supp}[X]$

$X_1, X_2 \stackrel{iid}{\sim} \text{Poisson}(\lambda)$

$$T = X_1 + X_2 \sim \sum_{x \in \{0,1,\dots\}} \left(\frac{\lambda^x e^{-\lambda}}{x!} \right) \left(\frac{\lambda^{t-x} e^{-\lambda}}{(t-x)!} \right) \parallel t-x \in \{0,1,\dots\}$$

$$= \lambda^t e^{-2\lambda} \sum_{x \in \{0,1,\dots\}} \frac{1}{x!(t-x)!} \parallel x \leq t = \lambda^t e^{-2\lambda} \sum_{x=0}^t \frac{1}{x!(t-x)!} \cdot \frac{t!}{t!}$$

$$= \frac{\lambda^t e^{-\lambda}}{\lambda!} \sum_{x=0}^t \binom{t}{x} = \frac{(2\lambda)^t e^{-2\lambda}}{t!} = \text{Poisson}(2\lambda)$$

Ex: $X, Y \stackrel{iid}{\sim} \text{Geom}(p)$ $E\left[\frac{1}{X+Y}\right] = P(X \geq Y)$

$$P(X \geq Y) = \frac{1}{2}$$

$$1 = P(X > Y) + P(X = Y) + P(X < Y) = 2P(X > Y) + \underbrace{P(X = Y)}_{> 0}$$

$$P(X > Y) = P(X < Y) \Rightarrow \text{bc they are arbitrary}$$

Idea

$P_{X,Y}(X,Y)$

	0	1	2	3
0	mm	mm	mm	mmmm
1		m	m	mmmm
2			m	mmmm $\rightarrow X > Y$
3				mmmm

$$\Rightarrow \frac{1}{1-(1-p)^2} \dots \text{Interpreting}$$

Recall Geometric Series

$$p_x(x) p_y(y) \quad \text{for } q \in (0,1) = \sum_{x=0}^{\infty} q^x = \frac{1}{1-q}$$

general Idea

$$P(X > Y) = \sum_{y \in \mathbb{R}} \sum_{x \in \mathbb{R}} p_{x,y}(x,y) \mathbb{1}_{x > y} = \sum_{y \in \mathbb{R}} \sum_{x \in (y, \infty)} ((1-p)^x p) \mathbb{1}_{x \in \{0,1,\dots\}} \Rightarrow$$

$$((1-p)^y p \mathbb{1}_{y \in \{0,1,\dots\}}) = p \sum_{y \in \{0,1,\dots\}} (1-p)^y \sum_{x \in \{y+1, y+2, \dots\}} (1-p)^x$$

let's re-index $x' = x - (y+1) \Rightarrow x = x' + y + 1$

$$\Rightarrow p^2 \sum_{y \in \{0,1,\dots\}} (1-p)^y \sum_{x' \in \{0,1,\dots\}} (1-p)^{x'+y+1} \Rightarrow (1-p)^{x'} (1-p)^y (1-p)$$

$$\Rightarrow p^2 (1-p) \sum_{y \in \{0,1,\dots\}} (1-p)^{2y} \sum_{x' \in \{0,1,\dots\}} (1-p)^{x'} = p(1-p) \sum_{y \in \{0,1,\dots\}} (1-p)^{2y}$$

$\frac{1}{1-(1-p)} \Rightarrow$ Geometric Formula

Rewrite

$$\Rightarrow p(1-p) \sum_{y \in \{0,1,\dots\}} ((1-p)^2)^y$$

$$\Rightarrow \frac{1}{1-(1-p^2)} \Rightarrow \frac{1}{1-(1-2p+p^2)} \Rightarrow \frac{1}{2p-p^2} \Rightarrow \frac{1}{p(2-p)}$$

$$\Rightarrow \frac{p(1-p)}{p(2-p)} = \boxed{\frac{1-p}{2-p}} \quad p > 0$$

Recall $a^{bc} = (a^b)^c \neq a^{(b^c)}$

Note $\sum_{x \in \mathbb{R}}$ can also be represented as $\sum_{n=-\infty}^{\infty}$

Expectation:

X, y discrete $E[X] = \sum_{x \in \mathbb{R}} x p(x)$

Prob weighted avg.

$$E[g(x)] = \sum_{x \in \mathbb{R}} g(x) p(x)$$

$$E[g(x, y)] = \sum_{x \in \mathbb{R}} \sum_{y \in \mathbb{R}} g(x, y) p(x, y) \Rightarrow$$

Consider:

$$E\left[\frac{1}{g(x)} \cdot g(x)\right] \Rightarrow$$

$$\Rightarrow \sum_{x \in \mathbb{R}} \sum_{x \in A} p(x) = \sum_{x \in A} p(x) = P(X \in A)$$

#6 R.V. Multinomial Coefficient.

p_1 = Prob of Apple in one draw from bag
 p_2 = " " banana " "
 $\Rightarrow p_1 + p_2 = 1$

Draw n fruits w/ replacement

X_1 = # Apple

X_2 = # Bananas

binomial

$$X_1 \sim \text{Bin}(n, p_1) = \binom{n}{x_1} p_1^{x_1} (1-p_1)^{n-x_1}$$

$$X_2 \sim \text{Bin}(n, p_2) = \binom{n}{x_2} p_2^{x_2} (1-p_2)^{n-x_2}$$

Note: $X_1 + X_2 = n$

Let $\vec{X} := \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \sim p_{X_1, X_2}(x_1, x_2)$

$n = 8$ fruits

$x_1 = 5$ Apple

$x_2 = 3$ Bananas

$$p_1^{x_1} p_2^{x_2} = \frac{n!}{x_1! x_2!} \prod_{x_1+x_2=n} \prod_{x_1 \in \{0,1,\dots,n\}} \prod_{x_2 \in \{0,1,\dots,n\}}$$

Apple Bananas

$$\binom{n}{x_1} = \frac{n!}{x_1! (n-x_1)!}$$

NEXT CLASS EXX

$$= \frac{n!}{x_1! x_2!}$$

Let us define multinomial Coefficient

$$\binom{n}{x_1, x_2} := \frac{n!}{x_1! x_2!} \prod_{x_1+x_2=n} \prod_{x_1 \in \{0, 1, \dots, n\}} \prod_{x_2 \in \{0, 1, \dots, n\}}$$

Rewriting the prob we get

$$= \binom{n}{x_1, x_2} p^{x_1} p^{x_2} = \text{Multinomial}(n, \vec{p})$$