

## Discrete Random Variables (II)

Suppose  $X$  is a discrete random variable.

$X = \text{set of values of } X$ .

- if finite
- if countably infinite.

Today's Goal:

Study some discrete r.v.'s which have  $X$  to be infinite.

### ① Poisson Distribution

$X$  is said to have Poisson Distribution with parameter:  $\lambda > 0$

if  $X = \{0, 1, 2, 3, \dots\}$  for  $x \in X$

$$p_X(x) = \frac{e^{-\lambda} \cdot \lambda^x}{x!}$$

$$\begin{aligned} \text{Note: } ① \quad e^\lambda &= 1 + \lambda + \frac{\lambda^2}{2!} + \frac{\lambda^3}{3!} + \dots \\ &= \sum_{x=0}^{\infty} \frac{\lambda^x}{x!} \end{aligned}$$

$$\Rightarrow 1 = e^{-\lambda} \cdot \sum_{x=0}^{\infty} \frac{\lambda^x}{x!} = \sum_{x=0}^{\infty} e^{-\lambda} \cdot \frac{\lambda^x}{x!} = \sum_{x=0}^{\infty} p_X(x)$$

$$\text{Also, } p_X(x) = e^{-\lambda} \cdot \frac{\lambda^x}{x!} \geq 0 \quad \forall x$$

$p_X$  is a legitimate probability Mass Function.

② If  $X$  has the Poisson Distribution with parameter  $\lambda$

$$\rightarrow E(X) = \lambda$$

$$V(X) = \sigma_X^2 = \lambda$$

$$\therefore \sigma_X = \sqrt{\lambda}$$

③ The Poisson distribution is often used to model phenomena where we are waiting for events to happen.

For example: In a "Poisson Process"  $\rightarrow$  the number of occurrences in a given time interval  $t$ .

### ② Negative Binomial Distribution

$X$  has the Negative Binomial Distribution with parameters:  $r, p$

if  $X = \{0, 1, 2, 3, \dots\}$  for  $x \in X$

$$p_X(x) = \binom{x+r-1}{r-1} p^r (1-p)^x$$

Note:

① The experiment for this distribution  $\rightarrow$  Perform Bernoulli trials with  $P(\text{success}) = p$  repeatedly until we get exactly  $r$ -successes.

$X = \# \text{ of failures that precede the } r^{\text{th}} \text{ success.}$

Then

$P(X=x) = P(\text{exactly } x \text{ failures before the } r^{\text{th}} \text{ success})$

Total ways we can have exactly  $x$  failures  $\rightarrow$  choosing  $x$  spots in  $x+r-1$  possible places (since the last spot has to be a success)

$$\text{can be done in } \binom{x+r-1}{x} = \binom{x+r-1}{r-1} \text{ ways.}$$

$$\therefore P(X=x) = \binom{x+r-1}{r-1} p^r (1-p)^x \quad \text{since } x \text{ failures}$$

since  $r$  successes.

② If  $X$  has negative Binomial Distribution with parameters:  $r, p$

$$\rightarrow E(X) = \lambda = \frac{r(1-p)}{p}$$

$$V(X) = \sigma_X^2 = \frac{r(1-p)}{p^2}$$

③ Called negative binomial

$$\binom{x+r-1}{x} = (-1)^x \binom{-r}{x} = (-1)^x \frac{(-r)(-r-1)(-r-2)\dots(-r-x+1)}{x(x-1)(x-2)\dots3 \cdot 1}$$

so that

$$p_X(x) = (-1)^x \binom{-r}{x} p^r (1-p)^x$$

very similar to the pmf of Binomial Distribution.

### ③ Geometric Distribution

$X$  has Geometric Distribution with parameter:  $p$

if  $X = \{1, 2, 3, \dots\}$  for  $x \in X$

$$p_X(x) = p(1-p)^{x-1}$$

Note:

① Geometric distribution  $\rightarrow$  special case of Negative Binomial pmf

When  $r=1$  and  $x=1, 2, 3, \dots$

② Is the simplest of the "waiting time" distributions

$X$  can be interpreted as the trial at which first success occurs  
"waiting for a success"

③ If  $X$  has Geometric distribution with "success" parameter:  $p$

$\rightarrow E(X) = \lambda = \frac{1}{p}$

$$V(X) = \sigma_X^2 = \frac{(1-p)}{p^2}$$

④ Sometimes used to model "lifetime" or "time until failure" of components.

### Relationships Between Discrete distributions (in the limit)

#### ① Binomial $\hookrightarrow$ hypergeometric

Let  $N, M, n \in \mathbb{N}$ ,  $p = \frac{M}{N}$

(i)  $\text{bin}(x; n, p) \rightarrow$  pmf for binomial with params:  $n$  = sample size  $p$  = prob of success

(ii)  $\text{hyper}(x; N, M, n) \rightarrow$  pmf for hypergeometric distribution with params:  $N$  = population size  $M$  = # of successes  $n$  = sample size

If  $N, M \rightarrow \infty$   
st  $\frac{M}{N} \rightarrow p$   $\rightarrow$   $\text{hyper}(x; N, M, n) \rightarrow \text{bin}(x, n, p)$

if sample size  $n$  is small compared to population size  $N$ , can assume that samples are "approximately" independent.

#### ② Binomial $\hookleftarrow$ Poisson

Suppose

$\text{bin}(x; n, p) \rightarrow$  pmf of Binomial with params  $n$  = sample size  $p$  = probability of success

$\text{pois}(x; \lambda) \rightarrow$  pmf of Poisson Distribution with parameter  $\lambda > 0$ .

If  $n \rightarrow \infty$  and  $p \rightarrow 0$   
such that  $n \cdot p \rightarrow \lambda$   $\rightarrow$   $\text{bin}(x; n, p) \rightarrow \text{pois}(x; \lambda)$

If  $n$  is large, and  $p$  very small  
 $\text{bin}(x; n, p) \approx \text{pois}(x; \lambda = np)$

"The Poisson Distribution is approximately Binomial for rare events."

$$\text{bin}(x; 100, 0.00) \quad n=20 ; \quad P(X \geq x)$$

$$\text{bin}(20; 100, 0.00) = \binom{100}{20} \cdot (0.00)^{20} \cdot (1-0.00)^{80}$$

$$\rightarrow \frac{100 \times 99 \times 98 \times \dots \times 81}{20!} \cdot ( )$$

$$\text{pois}(x; \lambda) = e^{-\lambda} \cdot \frac{\lambda^x}{x!} ;$$

$$\text{pois}(20; 0.1) = e^{-0.1} \cdot \frac{(0.1)^{20}}{20!}$$

$$X = \{0, 1, 2, \dots, \infty\}$$

$X \sim \text{pois}(\lambda) \iff \lambda > 0$

$$p_X(x) = P(X=x) = e^{-\lambda} \frac{\lambda^x}{x!}$$

$$\left\{ \begin{array}{l} p_X(x) \geq 0 \quad \forall x \in X \\ e^{-\lambda} \frac{\lambda^x}{x!} \geq 0 \end{array} \right.$$

$$e^{-\lambda} = 1 + \lambda + \frac{\lambda^2}{2!} + \frac{\lambda^3}{3!} + \dots$$

$$= \sum_{x=0}^{\infty} \frac{\lambda^x}{x!} \quad (\text{multiply both sides by } e^{-\lambda})$$

$$1 = e^{-\lambda} \cdot \sum_{x=0}^{\infty} \frac{\lambda^x}{x!} \Rightarrow 1 = \sum_{x=0}^{\infty} \left( e^{-\lambda} \frac{\lambda^x}{x!} \right)$$

→  $p_X(x)$  for Poisson distribution.

$x$	0	1	2	3	4	$\dots$
$p_X(x)$	$e^{-\lambda}$	$e^{-\lambda} \lambda$	$e^{-\lambda} \frac{\lambda^2}{2!}$	$e^{-\lambda} \frac{\lambda^3}{3!}$		

$$p_X(0) = e^{-\lambda} \frac{\lambda^0}{0!}, \quad p_X(1) = e^{-\lambda} \lambda$$

$$p_X(2) = e^{-\lambda} \frac{\lambda^2}{2!}$$

$$p_X(x) = e^{-\lambda} \frac{\lambda^x}{x!}$$

$$E(X) = 0 \cdot p_X(0) + 1 \cdot p_X(1) + 2 \cdot p_X(2) + \dots = \lambda$$

$$\text{Var}(X) = E(X^2) - (E(X))^2 = \lambda$$

Typically used to calculate the chance/probs of getting a certain # events in a given.

(Check out textbook for Poisson process with rate  $\alpha$ )

### Example:

Telephone operator on avg 5 calls over a 3 minute period.

What is the probability that there will be no calls in the next minute?

Assuming that the telephone calls follow a Poisson Process

$$P_k(t) = P(\text{getting } k \text{ calls over a time period } t)$$

Poisson distribution with rate  $\lambda = \alpha \cdot t$

# of occurrences in unit time.

5 calls in 3 minutes

$$\Rightarrow \frac{5}{3} \text{ calls in a minute} \iff \alpha = \frac{5}{3}$$

$$\lambda = \# \underset{\sum}{\text{of calls in a minute}}$$

Poisson dist with  $\lambda = \frac{5}{3}$

$$P(X=0) = P(\text{no calls in a minute}) = e^{-\frac{5}{3}}$$

$$P(\text{at least 2 calls}) = P(X \geq 2) \quad \text{in one min period}$$

$$= \sum_{x=2}^{\infty} p_X(x) = \sum_{x=2}^{\infty} p_X(x)$$

$$1 = \sum_{x=0}^{\infty} p_x(x) = p_x(0) + p_x(1) + \sum_{x=2}^{\infty} p_x(x)$$

$$p_x(x) = e^{-\lambda} \frac{\lambda^x}{x!} \quad \lambda = 5/3$$

$$\begin{aligned} P(X \geq 2) &= \sum_{x=2}^{\infty} p_x(x) = 1 - p_x(0) - p_x(1) \\ &= 1 - e^{-5/3} - e^{-5/3} \cdot \frac{5}{3} \\ &= 1 - e^{-5/3} \left( 1 + \frac{5}{3} \right) \end{aligned}$$

$Y = \# \text{ calls in 2 minutes.}$

$$\alpha = \frac{5}{3} \quad t = 2$$

$$\therefore \lambda = 2 \cdot \alpha = \frac{10}{3}$$

$$\begin{aligned} P(\text{0 calls in 2 minutes}) &= P(Y=0) \\ &= e^{-10/3} \end{aligned}$$

$$e^{-5/3} > e^{-10/3}$$

### Negative Binomial Distribution

$X$  has Neg Bin dist

$r = \# \text{ successes}$

$p = \text{prob of getting a success}$

$$X = \{0, 1, 2, \dots\}$$

$$p_x(x) = \binom{r+x-1}{x} \cdot p^r \cdot (1-p)^x$$

### Inverse Binomial Sampling

Population  $\rightarrow$  sampling from this population

{successes, failures}  $\rightarrow$  we are aware of  $P(S) = p$

$X = \# \text{ failures until the } r^{\text{th}} \text{ success.}$

$P(X=N) = P(\text{of exactly } N \text{ failures before the } r^{\text{th}} \text{ success})$

$P(X > N) = P(\text{at least } N \text{ failures before the } r^{\text{th}} \text{ success})$

$$r, p \rightsquigarrow \sum_{x=N}^{\infty} p_x(x)$$

### Geometric Distribution

$X = \# \text{ trials until a success}$

$$\begin{cases} 1, 2, 3, 4, \dots \end{cases}$$

$$p_x(x) = p \cdot (1-p)^{x-1}$$

$\xrightarrow{x-1 \text{ failures}}$

first success.

$$\text{nbm}(r, r, p) \longrightarrow \text{pois}(r; \lambda)$$

$$\begin{cases} r \rightarrow \alpha \\ p \rightarrow 1 \\ 1-p \rightarrow 0 \end{cases} \quad \begin{cases} r \cdot (1-p) \rightarrow \lambda \\ (1-p) \rightarrow 0 \end{cases}$$

If  $r$  is large  $p \rightarrow 1$ ;  $(1-p) \rightarrow 0$ .

$$P(\text{getting } k \text{ failures until the } r^{\text{th}} \text{ success}) \sim e^{-r} \cdot \frac{r^k}{k!}$$

$$= e^{-r(1-p)} \cdot \frac{(r(1-p))^k}{k!}$$