

Lecture 3.Random Variables.

Def. A **random variable** (r.v.)  $X$  on the probability space  $(\Omega, \mathcal{F}, P)$  is a mapping  $X: \Omega \rightarrow \mathbb{R}$  s.t.  
 $\{\omega \in \Omega : X(\omega) = x\} \in \mathcal{F}$  for all  $x \in \mathbb{R}$ .

Similarly,

vector  $\begin{pmatrix} X \\ Y \end{pmatrix}: \Omega \rightarrow \mathbb{R}^2$  is a random vector.

Ex. Toss a coin.

$$\Omega = \left\{ \begin{matrix} H \\ \omega_1 \end{matrix}, \begin{matrix} T \\ \omega_2 \end{matrix} \right\}$$

$$X = \{\#\text{ of heads}\} \in \{0, 1\}$$

$$X(\omega_1) = X(H) = 1, X(\omega_2) = X(T) = 0$$

Toss a coin 10 times

$$X \in \{0, 1, \dots, 10\}$$

Ex.  $X = \text{mark for this course}$

$$\Omega = [0, 100], X(\omega) = \omega \in [0, 100]$$

Discrete Random Variable.

Recall:  $\Omega$  is finite, then for any  $A \in \mathcal{F}$   
 $A = \bigcup_{\omega_i \in A} \{\omega_i\}$ ,  $P(A) = \sum_{\omega \in A} P(\{\omega\})$

Ex. Roll a die 6 times.  $A = \{\text{all outcomes same}\}$

$$P(A) = P(\{1, 1, 1, 1, 1, 1\}) + P(\{2, 2, 2, 2, 2, 2\}) + \dots$$

Formal definition of a Discrete Probability Space  $(\Omega, \mathcal{F}, P)$ :

$\Omega$  - countable

$$P: \Omega \rightarrow [0, 1] : \underbrace{P(\omega) \geq 0, \omega \in \Omega}_{\substack{\sum_{\omega \in \Omega} P(\omega) = 1}} \xrightarrow{\substack{\text{probability} \\ \text{mass function}}}$$

Def. A r.v.  $X$  is said to be **discrete** if it can take only a finite or countably infinite number of distinct values.

$X: \Omega \rightarrow$  a countable set

$$P_X(x_c) = P(X=x_c) \geq 0, \sum_c P_X(x_c) = 1$$

Def. The probability distribution of a discrete r.v.  $X$  is represented by a formula, a table or a graph which provides the list of all possible values that  $X$  can take and the pmf for each value.

Ex. we roll a die.

$$X = \# \text{ comes up}, \quad X(\omega) = \omega \in \{1, \dots, 6\}$$

$$P(X=1) = P(X=2) = \dots = P(X=6) = \frac{1}{6}$$

$$P_X(\omega) = \frac{1}{6}, \quad \omega = 1, 2, \dots, 6$$

$X$  has a uniform dist'n.

### Bernoulli Distribution

Ex. Roll a die.

$$X = \begin{cases} 1, & 6 \text{ comes up} \\ 0, & \text{ow} \end{cases}$$

$$P(X=1) = \frac{1}{6}, \quad P(X=0) = \frac{5}{6}$$

pmf :  $P_X(\omega) = \begin{cases} \frac{1}{6}, & \omega=1 \\ \frac{5}{6}, & \omega=0 \end{cases}$

$x$	$P$
0	$5/6$
1	$1/6$

$$X = \begin{cases} 1, \text{with prob. } p \\ 0, \text{with prob. } q \end{cases} \sim \text{Bernoulli Dist'n}$$

$p + q = 1$

## Binomial Distribution.

Ex: Roll a die  $n$  times and count the number of times 6 came up.

$X = \# \text{ of times we have '6'}$

$$X \in \{0, 1, 2, \dots, n\}$$

$$P(X=3) = \binom{n}{3} \left(\frac{1}{6}\right)^3 \left(\frac{5}{6}\right)^{n-3}$$

In general, if identical Bernoulli trials are repeated  $n$  times independently and  $X$  is a r.v. that counts the number of successes in the  $n$  trials then

$$\boxed{p_X(x) = P(X=x) = \binom{n}{x} p^x (1-p)^{n-x}, x \in \{0, 1, \dots, n\}}$$

$p$  = prob. of success

$X \sim \text{Binomial dist'n with parameters } n \text{ and } p. \quad X \sim \text{Bin}(n, p)$

Q: Is  $p_X(x)$  a valid pmf?

$$p_X(x) \geq 0$$

$$\sum_{x=0}^n p_X(x) = \sum_{x=0}^n \binom{n}{x} p^x (1-p)^{n-x}$$

$$= (p + (1-p))^n = 1$$

Binomial formula

## Geometric Distribution

Ex. we roll a die until the first 6 comes up.

$X = \# \text{ of rolls until we get 1st 6}$ .

$$P(X=4) = \left(\frac{5}{6}\right)^3 \frac{1}{6} =$$

$$x = 1, 2, 3, \dots$$

$X$  is a r.v. that counts the # of trials until the 1<sup>st</sup> success

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

$$X \sim \text{Geom}(p)$$

Question: Is this a valid pmf?

$$\sum_{x=1}^{\infty} p_X(x) = \sum_{x=1}^{\infty} (1-p)^{x-1} p = p + q p + q^2 p + \dots$$

$$\text{Calculate } P(X>k) = \underbrace{p(1+q+q^2+\dots)}_{=\frac{p}{1-q}} = \frac{p}{1-q} = 1$$

$$P(X>k) = P(X=k+1) + P(X=k+2) + \dots$$

$$= q^k p + q^{k+1} p + \dots = pq^k / (1+q+q^2+\dots)$$

$$= pq^k \frac{1}{1-q} = q^k$$

Memoryless property of geometric r.v.

13-6

$$\begin{aligned} P[X > m+n \mid X > m] &= P[X > n], \quad m, n \in \mathbb{Z} \\ \hookrightarrow &= \frac{P[X > m+n \text{ and } X > m]}{P[X > m]} = \frac{P[X > m+n]}{P[X > m]} \\ &= \frac{q^{m+n}}{q^m} = q^n = P[X > n] \end{aligned}$$

■

### Negative Binomial Distribution.

Ex. we roll a die until the second 6 comes up

$$P(X=4) = 3 \cdot \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^2$$

$X = \# \text{ of roll until we get two '6'}$

$X = \text{total \# of experiments when waiting for } r^{\text{th}} \text{ success}$

In general,

$$P_X(x) = \binom{x-1}{r-1} p^r (1-p)^{x-r}, \quad x = r, r+1, \dots$$

$$X \sim \text{neg. bin}(p, r)$$

Q: Is  $p_X(x)$  valid?

Show:  $\sum_{x=r}^{\infty} \binom{x-1}{r-1} p^r (1-p)^{x-r} = 1$

Hint:  $\sum_{k=0}^{\infty} \binom{k+r-1}{r-1} w^k = (1-w)^{-r}$

## Hypergeometric Distribution.

Ex. A hat contains 12 tickets: 7 black and 5 white.  
 Three tickets are drawn at random.  
 Let  $X = \#$  of black tickets drawn.

$$P(X=2) = \frac{\binom{7}{2} \binom{5}{1}}{\binom{12}{3}}$$

In general,  $N = \# \text{ of balls} = 12$   
 $r = \# \text{ of black balls} = 7$   
 $N-r = \# \text{ of white balls}$   
 $X = \# \text{ of black balls drawn}$   
 when taking  $n$  ball without replacement.

$$P(X=k) = \frac{\binom{r}{k} \binom{N-r}{n-k}}{\binom{N}{n}}$$

## Poisson Distribution.

- Model for the number of events occurring in a time (or space) interval where  $\lambda$  (a parameter of the distribution) is the rate of the occurrence of the events per one unit of time (or space).

$$X \sim \text{Poisson}(\lambda)$$

$$P_X(x) = P(X=x) = \frac{e^{-\lambda} \lambda^x}{x!}, \quad x=0, 1, 2, \dots$$

$X$  = # of events per one unit of time

Q. Is  $P_X(x)$  valid?

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

Ex. The number of cars that cross Spadina and Bloor intersection is a Poisson r.v. with  $\lambda = 15$  cars per minute. What is the probability that in a given minute exactly 8 cars will cross the intersection? less than 8?

$$P(X=8) = \frac{e^{-15} \cdot 15^8}{8!} =$$

$$P(X < 8) = P(0) + P(1) + \dots + P(7) \\ = \frac{e^{-15} 15^0}{0!} + \dots + \frac{e^{-15} 15^7}{7!}$$

$$P(X > 8) = 1 - P(X \leq 8)$$

# Distribution Function of Random Variables.

3.9

Def. A cumulative distribution function (cdf) of a r.v.  $X$  is a mapping  $F: \mathbb{R} \rightarrow [0, 1]$  defined by

$$F_X(x) = P(\{\omega \in \Omega : X(\omega) \leq x\}) = P(X \leq x)$$

$X$  is discrete with pmf  $p_X(x)$ ,  $x = 0, 1, 2, \dots$

$$F_X(s) = \begin{cases} 0, & s < 0 \\ p_X(0) + p_X(1) + \dots + p_X(\lfloor s \rfloor), & s \geq 0 \end{cases}$$

$\lfloor s \rfloor = \text{greatest integer} \leq s$

Ex.  $X \sim \text{Bin}(2, \frac{1}{2})$ . Find  $F_X(x)$ .

$$p_X(x) = \binom{2}{x} \left(\frac{1}{2}\right)^x \left(\frac{1}{2}\right)^{2-x}, \quad x = 0, 1, 2$$

$$p_X(0) = \frac{1}{4}, \quad p_X(1) = \frac{1}{2}, \quad p_X(2) = \frac{1}{4}$$

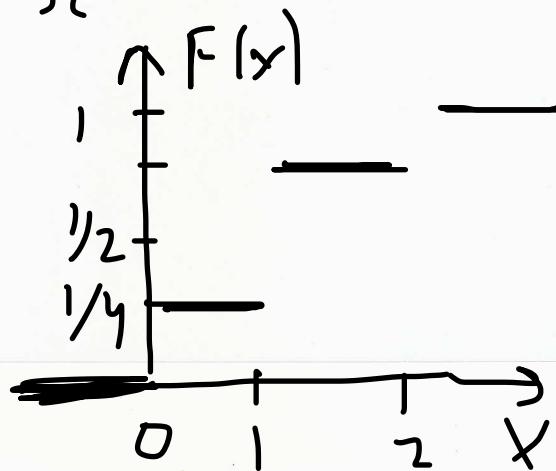
$$F(-2) = P(X \leq -2) = 0$$

$$F(0.5) = P(X \leq 0.5) = p_X(0) = \frac{1}{4}$$

$$F(1.5) = P(X \leq 1.5) = \frac{1}{4} + \frac{1}{2} = \frac{3}{4}$$

$$F(3) = P(X \leq 3) = \sum_{x=0}^3 p_X(x) = 1$$

$$F_X(x) = \begin{cases} 0, & x < 0 \\ \frac{1}{4}, & 0 \leq x < 1 \\ \frac{3}{4}, & 1 \leq x < 2 \\ 1, & x \geq 2 \end{cases}$$



Step function

## Properties of cdf:

- $F$  is monotone, non-decreasing
- $F(x) \rightarrow 0$  as  $x \rightarrow -\infty$
- $F(x) \rightarrow 1$  as  $x \rightarrow \infty$
- $F(x)$  is continuous from the right
- For  $a < b$   $P(a < X \leq b) = F_x(b) - F_x(a)$

why?

$$\begin{aligned}
 P(a < X \leq b) &= P((a < X) \cap \{X \leq b\}) \stackrel{R}{=} \\
 &= P(a < X) + P(X \leq b) - P((a < X) \cup (X \leq b)) \\
 &= P(a < X) + P(X \leq b) - 1 \\
 &= P(X \leq b) - P(X \leq a) = F_x(b) - F_x(a)
 \end{aligned}$$

## Relation between Binomial and Poisson Distributions,

$$P_X(x) = \binom{n}{x} p^x (1-p)^{n-x}$$

$p$  - prob. of success  $\rightarrow 0$

$n \rightarrow \infty$

Choose these limits so that

$$ph = \lambda$$

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

$\lambda = ph$

$$\lim_{h \rightarrow \infty} \binom{h}{x} p^x (1-p)^{h-x} =$$

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

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

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

$$= \frac{\lambda^x}{x!} \lim_{h \rightarrow \infty} 1 \cdot \left(1 - \frac{1}{h}\right)^1 \cdots \left(1 - \frac{x-1}{h}\right)^1 \left(1 - \frac{x}{h}\right)^{-x}$$

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

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

$$= \frac{h \cdot (h-1) \cdot (h-2) \cdots (h-x+1)}{(h-x)! (h-1) \cdots (h-(x-1))}$$