

## Bernoulli Random Variable

1. 
$$X = \begin{cases} 1 & \text{if success} \\ 0 & \text{if failure} \end{cases}$$
 || This is R.V.

2. P.M.F  $P(X=1) = p$   $(0 \leq p \leq 1)$   
 $P(X=0) = 1-p$

3.  $E(X) = p$

4.  $V(X) = p(1-p)$

5. its distribution function is not relevant as only 2 values are possible.  $X=0$  or  $X=1$  ... so,  $P(X \leq x)$  is of no use...

$X \sim (n, p)$  Binomial Random Variable  $(n, p)$   $\rightarrow$  called parameters

1.  $X$  = number of successes in  $n$  independent trials, each trial having  $p$  as probability of success.

2. P.M.F  $P(X=i) = {}^n C_i (p)^i (1-p)^{n-i}$   
 $\approx$   $i$  successes in  $n$  trials.

3.  $E(X) = np$

4.  $V(X) = np(1-p)$

5. Distribution -  $P(X \leq i) = \sum_{k=0}^i {}^n C_k p^k (1-p)^{n-k}$   
 $i = 0, 1, \dots, n$

Note  $\left[ \begin{array}{l} X_1 \sim (n_1, p), X_2 \sim (n_2, p) \\ \Rightarrow (X_1 + X_2) \sim (n_1 + n_2, p) \end{array} \right]$  Something like seg. sampling  
 $\Rightarrow$  we can  $\uparrow$  no. of indep. trials if prob. remains same

## Poisson Random Variable (parameter)

1.  $X$  = defined via p.m.f.  
 i.e a variable that follows  
 p.m.f given in 2. is a P.R.V.

2. p.m.f  $P(X=i) = \frac{e^{-\lambda} \lambda^i}{i!}; i=0,1,\dots$

where  $\lambda (>0)$  is parameter.

3.  $E(X) = \lambda$   $\frac{P(X=i+1)}{P(X=i)} = \frac{\lambda}{i+1}$  : just a result

4.  $V(X) = \lambda$

5. Distribution:  $P(X \leq i) = \sum_{k=0}^i e^{-\lambda} \frac{\lambda^k}{k!}; i=0,1,\dots$

Note → when  $n$  is large and  $p$  is small,  
 binomial R.V goes close to Poisson...  
 mathematically, it is easy to calculate  
 Poisson. So, we use it.

Example - (i) no. of failures of bank transaction  
 (ii) no. of meteors reaching earth  
 (iii) no. of failed downloads over  
 a website  $\hat{=}$

Poisson reproductive property  $\rightarrow$

$$\left\{ \begin{array}{l} X_1 \sim \text{Poisson } (\lambda_1) \quad \& \quad X_1, X_2 \text{ are independent} \\ X_2 \sim \text{Poisson } (\lambda_2) \\ \Rightarrow X_1 + X_2 \sim \text{Poisson } (\lambda_1 + \lambda_2) \end{array} \right. \begin{array}{l} \\ \\ \text{i.e. basically avg adds} \end{array}$$

Very useful in factory production analysis.

Ex- Daily defect items  $\sim$  Poisson (4)

Find possibilities of defects over a week.

$\Rightarrow X_1 = \text{defects on day 1} \sim \text{Poisson } (4)$

$\vdots$   
 $X_7 = \text{, , , , , , } \sim \text{ " } (4)$

Total weekly  $(X_1 + \dots + X_7) = \text{Poisson } (28)$   
so, instantly we can analyze weekly thing.

Note - There are 2 more R.V's given in book - Hypergeometric  $\rightarrow$  I think not needed.  
and uniform  $\rightarrow$  very simple

## Normal Random Variable

$\mathcal{N}(\mu, \sigma^2)$

continuous

1.  $X \sim \mathcal{N}(\mu, \sigma^2)$  is said to be normally distributed.

$$- \frac{(x-\mu)^2}{2\sigma^2}$$

$$e^{-\frac{(x-\mu)^2}{2\sigma^2}} : (x \in \mathbb{R})$$

2. P.d.f  $f(x) = \frac{1}{\sigma \sqrt{2\pi}}$

↓

Cdf can be found using integration (Note -  $\frac{1}{\sqrt{2\pi}} = 0.399$ )

3.  $E(X) = \mu \quad \Leftarrow \quad V(X) = \sigma^2$

$[E(ax+b) = a\mu + b] \quad [V(ax+b) = a^2\sigma^2]$

Property - each  $\mathcal{N}(\mu, \sigma^2)$  can be reduced to  $\mathcal{N}(0, 1)$ . This eases the maths...

standard Cdf =  $\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-y^2/2} dy \quad (x \in \mathbb{R})$

Note - if  $X \sim \mathcal{N}(\mu, \sigma^2) \Rightarrow \left(\frac{x-\mu}{\sigma}\right) \sim \mathcal{N}(0, 1)$

$$P(X \leq b) = \Phi\left(\frac{b-\mu}{\sigma}\right)$$

$$P(a \leq X \leq b) = \Phi\left(\frac{b-\mu}{\sigma}\right) - \Phi\left(\frac{a-\mu}{\sigma}\right)$$