

Last day, we started  $\text{Unif} [a, b]$  dist $\tilde{\sim}$ .

$$f_x(x) = \frac{1}{b-a} \quad \text{if} \quad a \leq x \leq b$$

$$= 0 \quad , \quad \text{o.w.}$$

$$E(x) = \int_a^b x \cdot \frac{1}{b-a} dx = \frac{b+a}{2} \quad (\text{Remark!- you can derive from the expression of MGF as well})$$

$$\text{Var}(x) = E(x^2) - \{E(x)\}^2 \stackrel{\text{(check)}}{=} \frac{(b-a)^2}{12} \rightarrow \text{That also, you can derive from MGF.}$$

S.p. case:- If  $a = 0$  &  $b = 1$ , we have well-known

uniform  $[0, 1]$  dist $\tilde{\sim}$ .

Why  $\text{Unif}[0,1]$  is so special ?? (because it has numerous theoretical applications).

### Applications :-

1. In the limiting sense, the discrete uniform dist<sup>n</sup>.  
(prob. mass  $\frac{1}{n}$  on each of the mass points  $(\frac{1}{n}, \frac{2}{n}, \dots, \frac{n-1}{n}, \frac{n}{n})$ )  
will converge to  $\text{Unif}[0,1]$ . Already, we know that discrete uniform enables us to construct sample version of the moments, which is useful real data analyses.
2. If  $X$  is a cont<sup>n</sup>. r.v. with CDF  $F_X$ , Then  $F_X$  will follow uniform  $(0,1)$  dist<sup>n</sup>.

This helps us to generate a random from any cont<sup>n.</sup> dist<sup>n.</sup>

Example:-

Suppose, we want to generate random number

from Cauchy (0, 1) dist<sup>n.</sup>  $\rightarrow f_x(x) = \frac{1}{\pi(1+x^2)}, x \in \mathbb{R}$ ,  
p.d.f.

Since  $F_x$  follows wif (0, 1), choose one number

CDF of  
Cauchy (0, 1)

random from (0, 1). (say, 0.89).

If  $x \sim \text{Cauchy}(0, 1)$ , then  $F_x(x) = \int_{-\infty}^x f_x(y) dy$

$$= \int_{-\infty}^x \frac{1}{\pi(1+y^2)} dy$$

Here, the random number from Cauch (0, 1) will be  $\check{=} \frac{1}{2} + \frac{1}{\pi} \tan^{-1} x$   
 $\frac{1}{2} + \frac{1}{\pi} \tan^{-1} x = 0.89 \Leftrightarrow x = ??$

### 3. Monte-Carlo integration

W.L.G, we consider  $\int_0^1 f(x) dx = \int_0^1 f(x) \times 1 dx = E_x [f(x)]$ , where  $x \sim \text{Unif}(0,1)$ .  
 instead of  $\int_a^b f(x) dx$

Now, based on WLLN, one can approximate  $E_x [f(x)]$ .

Algorithm:-

Step 1:- Let  $x_1, \dots, x_n$  be random numbers from  $\text{Unif}(0,1)$ .

Step 2:- Compute  $f(x_1), f(x_2), \dots, f(x_n)$ .

Step 3:-  $\int_0^1 f(x) dx = E_x [f(x)] \approx \frac{1}{n} \sum_{i=1}^n f(x_i)$

Rough idea:-

$$n^{\delta} \left[ \int_0^1 f(x) dx - \frac{1}{n} \sum_{i=1}^n f(x_i) \right], \quad \delta > 0$$

## 2. Exponential dist<sup>n=</sup> :-

$X$ : Random variable with exp. dist<sup>n=</sup> with mean =  $\frac{1}{\lambda}$ ,

The p.d.f. of  $X$  will be

$$f_X(x) = \lambda \cdot e^{-\lambda x}, \quad x \geq 0$$

$$= 0, \quad \text{otherwise}$$

The CDF of  $X$  will be

$$P[\omega: X(\omega) \leq x] = F_X(x) = \begin{cases} \checkmark & 1 - e^{-\lambda x}, \quad x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

$$M_X(t) = \frac{1}{1 - \frac{t}{\lambda}} \quad \text{for } t > 0$$

$$\text{MGF} \quad \mathbb{E}(x) = \frac{1}{\lambda} \quad \& \quad \text{Var}(x) = \mathbb{E}(x^2) - \{\mathbb{E}(x)\}^2 = \frac{1}{\lambda^2}$$

Memory loss property of Exp. dist $\stackrel{?}{=}$ .

Want to establish that

$$P[X > s+t | X > s] = P[X > t].$$

Observe that  $P[X > s+t | X > s]$

$$\begin{aligned} &= \frac{P[X > s+t \cap X > s]}{P[X > s]} = \frac{P[X > s+t]}{P[X > s]} \\ &= \frac{e^{-\lambda(s+t)}}{e^{-\lambda s}} = e^{-\lambda t} = P[X > t]. \end{aligned}$$

Remark :- A person aged 10, the survival chance till age 20 will NOT be the same as a person aged 80, the survival chance till age 90 - And hence, the lifetime of human cannot be fitted by exp. dist $\stackrel{?}{=}$ .

3. Gamma dist<sup>n</sup> :-

X : Random variable associated with Gamma ( $\alpha, \beta^2$ ) dist<sup>n</sup>.

The p.d.f is

$$f_X(x) = \frac{1}{\Gamma(\alpha) \beta^\alpha} e^{-\frac{x}{\beta}} x^{\alpha-1}, x \geq 0.$$

$\downarrow$   
gamma f<sup>n</sup>

Here,  $\alpha$  is called the shape parameter &  $\beta$  is the scale parameter.

$\Gamma_x$  : Gamma f<sup>n</sup>.

Properties :-

$$\text{i)} \quad \Gamma_x = (x-1) \Gamma_{x-1}$$

ii) If  $x$  is an integer, then  $\Gamma_x = (x-1)!$

$$\text{iii)} \quad \Gamma_2 = \sqrt{\pi}$$

$$M_x(t) = E[e^{tx}] = \frac{1}{\beta^\alpha \Gamma(\alpha)} \int_0^\infty e^{tx} e^{-\frac{x}{\beta}} x^{\alpha-1} dx$$

*check  
carefully*

$$= \frac{1}{(1-\beta t)^\alpha}$$

And  $E(x^K) = \frac{1}{\beta^\alpha \Gamma(\alpha)} \int_0^\infty x^K x^{\alpha-1} e^{-\frac{x}{\beta}} dx$

*check  
carefully*

$$= \frac{\Gamma(\alpha+K)}{\Gamma(\alpha)} \beta^K$$

Properties of Gamma function

$$E(x) = \frac{\Gamma(\alpha+1)}{\Gamma(\alpha)} \quad \beta = \frac{\alpha \Gamma(\alpha)}{\Gamma(\alpha)} \quad \beta = \alpha \beta^{\alpha-1}$$

$E(x^2) = \frac{\Gamma(\alpha+2)}{\Gamma(\alpha)} \quad \beta^2 = \alpha(\alpha+1) \beta^2$ , and hence,  $\text{Var}(x) = E(x^2) - \{E(x)\}^2$

$$= \alpha \beta^2.$$

Remark:- Put  $\alpha = 1$  &  $\beta = \frac{1}{\lambda}$ , that will be  
exp. dist<sup>n</sup> with mean =  $\frac{1}{\lambda}$ .  
(So, everything like  $M_x(t)$  or  $E(x^k)$  of exp. dist<sup>n</sup>-  
can directly be computed from Gamma dist<sup>n</sup>).

Application:- Earthquake data follow Gamma dist<sup>n</sup>

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