

Let  $-\infty < \alpha < \beta < \infty$ . A continuous type r.v.  $X$  is said to have a uniform (or rectangular) dist<sup>n</sup> over the interval  $(\alpha, \beta)$  written as  $X \sim U(\alpha, \beta)$  if its p.d.f is given by

$$f(x|\alpha, \beta) = \begin{cases} \frac{1}{\beta - \alpha}, & \alpha < x < \beta \\ 0, & \text{o/w} \end{cases}$$

$\{U(\alpha, \beta) : -\infty < \alpha < \beta < \infty\}$  is a family of dist<sup>n</sup> with parameter  $\alpha$  &  $\beta$ .

$$\mu'_r = E(x^r) = \int_{\alpha}^{\beta} \frac{x^r}{\beta - \alpha} dx = \frac{\beta^{r+1} - \alpha^{r+1}}{(r+1)(\beta - \alpha)}$$

$$E(x) = \frac{\alpha + \beta}{2} = \mu$$

$$\mu_r = E(x - \mu)^2 = \int_{\alpha}^{\beta} \left(x - \frac{\alpha + \beta}{2}\right)^2 \frac{1}{\beta - \alpha} dx$$

$$= \int_{-\frac{\beta - \alpha}{2}}^{\frac{\beta - \alpha}{2}} \frac{t^r}{\beta - \alpha} dt = \begin{cases} 0, & r = 1, 3, 5, \dots \\ \frac{(\beta - \alpha)^r}{2^r (r+1)}, & r = 2, 4, 6, \dots \end{cases}$$

$$\text{Var}(x) = \frac{(\beta - \alpha)^2}{12}$$

c. d. f

$$F_X(x) = \begin{cases} 0, & x < \alpha \\ \frac{x - \alpha}{\beta - \alpha}, & \alpha \leq x < \beta \\ 1, & x \geq \beta \end{cases}$$

$$\begin{aligned} \text{m.g.f. } M_X(t) &= E(e^{tx}) = \int_{\alpha}^{\beta} \frac{e^{tx}}{\beta - \alpha} dx \\ &= \begin{cases} \frac{e^{t\beta} - e^{t\alpha}}{t(\beta - \alpha)}, & t \neq 0 \\ 1, & t = 0 \end{cases} \end{aligned}$$

Exponential distribution:

Consider a Poisson process  $X(t)$  with rate  $\lambda (> 0)$ .

Let  $T$  be the time of the first occurrence

We want prob. dist of  $T$



$$P(T > t) = P(\text{upto to time } t \text{ no occurrence})$$

$$= P(X(t) = 0) = \begin{cases} e^{-\lambda t}, & t > 0 \\ 1, & t \leq 0 \end{cases}$$

$$F_T(t) = P(T \leq t) = 1 - P(T > t)$$

$$= \begin{cases} 0, & t \leq 0 \\ 1 - e^{-\lambda t}, & t > 0 \end{cases}$$

So the pdf of  $T$  is

$$f_T(t) = \begin{cases} \lambda e^{-\lambda t}, & t > 0 \\ 0, & \text{o/w} \end{cases}$$

Def<sup>n</sup>: A random  $T$  variable said to follow a exponential dist<sup>n</sup> written as  $T \sim \text{Exp}(\lambda)$  if  $T$  has pdf.

$$f_T(t) = \begin{cases} \lambda e^{-\lambda t}, & t > 0 \\ 0, & \text{o/w} \end{cases}$$

The family of dist<sup>n</sup>  $\{\text{Exp}(\lambda) : \lambda > 0\}$  is a one parameter family of dist<sup>n</sup>.

$$\mu'_k = E(T^k) = \int_0^{\infty} t^k e^{-\lambda t} dt = \lambda \frac{\Gamma(k+1)}{\lambda^{k+1}} = \frac{\lambda k!}{\lambda^{k+1}}$$

$$= \frac{k!}{\lambda^k}, \quad k = 1, 2, \dots$$

$$\mu_1' = E(T) = \frac{1}{\lambda}, \quad \mu_2' = \frac{2}{\lambda^2}, \quad \mu_3' = \frac{1}{\lambda^3}$$

$$\mu_3' = \frac{6}{\lambda^3}, \quad \mu_4' = \frac{24}{\lambda^4}, \quad \mu_3 = \frac{2}{\lambda^3}, \quad \mu_4 = \frac{9}{\lambda^4}$$

$$\beta_1 = \frac{2}{\lambda^3} / \frac{1}{\lambda^3} = 2 > 0 \quad \text{Positively skewed.}$$

$$\beta_1 = \frac{9}{\lambda^4} / \frac{1}{\lambda^4} - 3 = 6 > 0 \quad \text{always have high peak.}$$

Consider

$$P(T > a) = e^{-\lambda a}$$

$$P(\underbrace{T > a+b}_{E_1} | \underbrace{T > b}_{E_2}) = \frac{P(E_1 \cap E_2)}{P(E_2)} = \frac{P(T > a+b)}{P(T > b)}$$

$$= e^{-\lambda a} = P(T > a)$$

This is called the lack of memory property of exponential dist<sup>n</sup>.

Let  $T$  denote that waiting time for the failure of a system

$$P(T > a+b | T > b) = P(\text{given that the system not fail till time } b, \text{ the system will not fail an additional time } a)$$

$P(T > a) = P(\text{system will not fail till time } a).$

### Shifted Exponential Dist<sup>n</sup>

Def<sup>n</sup>: A random variable  $X$  of continuous type is said to follow a shifted exponential dist<sup>n</sup> written as  $X \sim \text{Exp}(\mu, \sigma)$  if its probability density function is

given by

$$f_X(x) = \begin{cases} \frac{1}{\sigma} e^{-\frac{(x-\mu)}{\sigma}}, & x > \mu \\ 0, & \text{o/w} \end{cases} \quad \begin{matrix} \sigma > 0 \\ \mu \in \mathbb{R} \end{matrix}$$

$$E(X-\mu)^k = \int_{\mu}^{\infty} (x-\mu)^k \frac{1}{\sigma} e^{-\frac{x-\mu}{\sigma}} dx$$

$$= \int_0^{\infty} \sigma^k t^k e^{-t} dt = \Gamma(k+1) \sigma^k = k! \sigma^k$$

$$E(X-\mu) = \sigma \Rightarrow E(X) = (\mu + \sigma)$$

$$E(X-\mu)^2 = 2\sigma^2 \Rightarrow E(X^2) = 2\sigma^2 + 2\mu\sigma + \mu^2$$

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

m.g.f.

$$M_X(t) = E(e^{tx}) = \int_{\mu}^{\infty} \frac{1}{\sigma} e^{tx} e^{-\frac{x-\mu}{\sigma}} dx$$

Take  $\frac{x-\mu}{\sigma} = y$

$$M_X(t) = \int_0^{\infty} e^{t(\sigma y + \mu)} e^{-y} dy = \int_0^{\infty} e^{t\mu} e^{y(\sigma t - 1)} dy$$

$$= e^{\mu t} \left. \frac{e^{y(\sigma t - 1)}}{(\sigma t - 1)} \right|_0^{\infty} \quad \sigma t < 1 = t < \frac{1}{\sigma}$$

$$= \frac{e^{\mu t}}{1 - \sigma t}$$

$$M_X(t) = \frac{e^{\mu t}}{1 - \sigma t}$$

Take  $\mu = 0$ ,  $M_X(t) = \frac{1}{1 - \sigma t}$

Take  $\frac{1}{\sigma} = \lambda \Rightarrow \sigma = \frac{1}{\lambda}$

$$M_X(t) = \frac{\lambda}{\lambda - t} \leftarrow \text{m.g.f. of exponential dist}^n$$

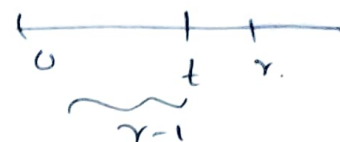
Result:  $X \sim \text{Exp}(\mu, \sigma)$ ,  $Y = ax + b \sim \text{Exp}(a\mu + b, a\sigma)$

Proof: H.W. use m.g.f. of  $X$  find m.g.f. of  $Y$ .

Consider a Poisson process  $X(t)$  with rate  $\lambda$ .

Let  $T_r$  denotes the time of  $r^{\text{th}}$  occurrence.

$$P(T_r > t) = P(X(t) \leq r-1), \quad t > 0$$



$$[P(T_r > t) = P(\text{r}^{\text{th}} \text{ occurrence not take place till time } t)]$$

$$P(T_r > t) = \begin{cases} P(X(t) \leq r-1), & t > 0 \\ 1, & t \leq 0. \end{cases}$$

$$= \begin{cases} \sum_{j=0}^{r-1} \frac{e^{-\lambda t} (\lambda t)^j}{j!}, & t > 0 \\ 1, & t \leq 0. \end{cases}$$

$$F_{T_r}(t) = 1 - P(T > t) = \begin{cases} 0, & t \leq 0 \\ 1 - \sum_{j=0}^{r-1} \frac{e^{-\lambda t} (\lambda t)^j}{j!}, & t > 0. \end{cases}$$

So the pdf  $T_r$  is



$$f_{T_r}(t) = \begin{cases} \frac{\lambda^r t^{r-1} e^{-\lambda t}}{\Gamma(r)}, & t > 0, \quad \lambda > 0 \\ 0 & \text{o/w.} \end{cases}$$

This gives the pdf of the waiting time for the  $r$ th occurrence in a Poisson process.

Def<sup>n</sup>: A continuous type r.v.  $X$  is said to follow a gamma distribution as written as

$X \sim \text{Gamma}(r, \lambda)$  if its pdf is given as

$$f_X(x) = \begin{cases} \frac{\lambda^r x^{r-1} e^{-\lambda x}}{\Gamma(r)}, & x > 0, \quad \lambda > 0, r > 0 \\ 0 & \text{o/w} \end{cases}$$

$$\mu'_k = E(X^k) = \int_0^{\infty} \frac{x^k \lambda^r x^{r-1} e^{-\lambda x}}{\Gamma(r)} dx$$

$$= \frac{\lambda^r}{\Gamma(r)} \int_0^{\infty} x^{k+r-1} e^{-\lambda x} dx = \frac{\lambda^r}{\Gamma(r)} \frac{\Gamma(k+r)}{\lambda^{k+r}}$$

$$= \frac{\Gamma(k+r)}{\Gamma(r)} \frac{1}{\lambda^k}$$



$$E(X) = \frac{\Gamma(r+1)}{\Gamma(r)} \cdot \frac{1}{\lambda} = \frac{r}{\lambda}$$

$$E(X^2) = \mu'_2 = \frac{r(r+1)}{\lambda^2}, \quad \text{Var}(X) = \frac{r(r+1)}{\lambda^2} - \frac{r^2}{\lambda^2} = \frac{r}{\lambda^2}$$

$$M_X(t) = \int_0^{\infty} e^{tx} \frac{\lambda^r e^{-\lambda x} x^{r-1}}{\Gamma(r)} dx$$

$$= \frac{\lambda^r}{\Gamma(r)} \int_0^{\infty} e^{-x(\lambda-t)} x^{r-1} dx$$

$$= \frac{\Gamma(r)}{(\lambda-t)^r} \cdot \frac{\lambda^r}{\Gamma(r)}, \quad \lambda > t.$$

$$= \left( \frac{\lambda}{\lambda-t} \right)^r, \quad \lambda > t.$$

Note:  $\{G(r, \lambda): r > 0, \lambda > 0\}$  is a family of dist<sup>n</sup>.

Note: If we take  $r=1$ ,  $X \sim \exp(\lambda)$ .

Beta Distribution: Let  $X$  be a r.v. of continuous type and let  $a > 0$  &  $b > 0$  be constants. The r.v.  $X$  is said to follow the dist<sup>n</sup> with parameter  $(a, b)$  written as  $X \sim \text{Be}(a, b)$  if its prob. density fun<sup>n</sup> is given by