3.7 Exponential and Gamma Distributions

Definition. (Exponential Distribution)

A random variable X is said to have an exponential distribution with parameter $\lambda > 0$, if its density is given by

$$f_X(x) = \lambda e^{-\lambda x}$$
 for all $x \ge 0$ and $\lambda > 0$.

We write it as $X \sim Expo(\lambda)$.

Definition. (Gamma Distribution)

A random variable X is said to have a gamma distribution if its density is given by

$$f_X(x) = \frac{\lambda}{\Gamma(r)} (\lambda x)^{r-1} e^{-\lambda x} \text{ for all } x \ge 0,$$

where r > 0 and $\lambda > 0$ are called the parameters of the gamma distribution. We write $X - gam(\lambda; r)$ or $X - G(\lambda; r)$.

Remark 1. $\Gamma(r)$ is called gamma function and is defined as

$$\Gamma(r) = \int_{0}^{\infty} x^{r-1} e^{-x} dx, r > 0.$$

It is easy to verify that

It is easy to verify that
$$\Gamma(1) = 1$$
, $\Gamma(a+1) = a \Gamma(a)$, $\Gamma(n) = (n-1)$!, n a positive integer.

Remark 2. Taking r = 1, we see that gamma density becomes exponential density ($:: \Gamma(1) = 1$).

Theorem 3.7.1. If X has an exponential distribution, then

$$E\left[X\right] = \frac{1}{\lambda}$$
, $var\left[X\right] = \frac{1}{\lambda^{2}}$ and $m_{X}\left(t\right) = \frac{\lambda}{\lambda - t}$ for $t < \lambda$.

Proof. We have

$$E[X] = \int_{0}^{\infty} x f_X(x) dx = \int_{0}^{\infty} x \lambda e^{-\lambda x} dx, \ \lambda > 0.$$

Integrating by parts, we get

$$E[X] = \lambda \left| x \left(\frac{e^{-\lambda x}}{-\lambda} \right) \right|_0^\infty + \int_0^\infty 1 \cdot e^{-\lambda x} dx$$
$$= 0 - \frac{1}{\lambda} \left| e^{-\lambda x} \right|_0^\infty = -\frac{1}{\lambda} (0 - 1) = \frac{1}{\lambda}. \tag{1}$$

Now
$$E[X^{2}] = \int_{0}^{\infty} x^{2} (\lambda e^{-\lambda x}) dx$$
$$= \lambda \left| x^{2} \left(\frac{e^{-\lambda x}}{-\lambda} \right) \right|_{0}^{\infty} + 2 \int_{0}^{\infty} x e^{-\lambda x} dx$$

$$= 0 + \frac{2}{\lambda} \int_{0}^{\infty} x \lambda e^{-\lambda x} dx = \frac{2}{\lambda} \left(\frac{1}{\lambda} \right) = \frac{2}{\lambda^{2}}, \text{ by (1)}.$$

$$\operatorname{var}[X] = E[X^2] - (E[X])^2 = \frac{2}{\lambda^2} - \frac{1}{\lambda^2} = \frac{1}{\lambda^2}.$$

The m.g.f. of X is given by

Hence

$$m_X(t) = E\left[e^{tX}\right] = \int_0^\infty e^{tx} \left(\lambda e^{-\lambda x}\right) dx$$

$$= \lambda \int_0^\infty e^{-(\lambda - t)x} dx, \lambda > t$$

$$= -\frac{\lambda}{\lambda - t} \left| e^{-(\lambda - t)x} \right|_0^\infty = \frac{\lambda}{t - \lambda} (0 - 1)$$

$$= \frac{\lambda}{\lambda - t}, \text{ for } t < \lambda$$

$$m_X(t) = \left(1 - \frac{t}{\lambda}\right)^{-1}, \frac{t}{\lambda} < 1.$$

Theorem 3.7.6. If the random variable X has an exponential distribution with parameter $\lambda > 0$, then

$$P[X>a+b \mid X>a] = P[X>b], \text{ for } a>0, b>0.$$

$$P[X>a+b \mid X>a] = \frac{P[X>a+b \text{ and } X>a]}{P[X>a]} = \frac{P[X>a+b]}{P[X>a]}$$

$$[\because a+b>a \text{ as } b>0]$$

$$= \int_{a+b}^{\infty} \lambda e^{-\lambda x} dx + \int_{a}^{\infty} \lambda e^{-\lambda x} dx = \frac{e^{-\lambda(a+b)}}{e^{-\lambda a}}$$

$$= e^{-\lambda h} = \int_{a}^{\infty} \lambda e^{-\lambda x} dx = P[X>b].$$

EXAMPLES

Example 3.7.1. If $X - Expo(\lambda)$ with $P[X \le 1] = P[X > 1]$, find var[X].

Solution. We have

$$P[X \le 1] = P[X > 1] = 1 - P[X \le 1]$$

$$\Rightarrow 2P[X \le 1] = 1 \text{ or } P[X \le 1] = \frac{1}{2},$$

$$\therefore \frac{1}{2} = \int_{0}^{1} \lambda e^{-\lambda x} dx = -\left| e^{-\lambda x} \right|_{0}^{1} = -(e^{-\lambda} - 1) = 1 - e^{-\lambda}$$

$$\Rightarrow e^{-\lambda} = 1 - \frac{1}{2} = \frac{1}{2} \Rightarrow e^{\lambda} = 2 \Rightarrow \lambda = \log_{e} 2.$$
Since $X \sim \text{Expo}(\lambda)$, $\text{var}[X] = \frac{1}{\lambda^{2}}$ [Theorem 3.7.1.]

Hence var $[X] = \frac{1}{(\log 2)^2}$.

Example 3.7.2. If X has exponential distribution with mean 2, find P[X < 1 | X < 2].

Solution. We are given that $\frac{1}{\lambda} = 2$ *i.e.*, $\lambda = \frac{1}{2}$.

Now
$$P[X < 1 \mid X < 2] = \frac{P[X < 1]}{P[X < 2]} = \frac{0}{2}$$

$$= \frac{-\left|e^{-\lambda x}\right|_{0}^{1}}{-\left|e^{-\lambda x}\right|_{0}^{2}} = \frac{1 - e^{-\lambda}}{1 - e^{-2\lambda}}$$

$$= \frac{1}{1 + e^{-\lambda}} = \frac{1}{1 + e^{-1/2}}.$$

Example 3.7.3. If $X \sim U(0, 2)$ and $Y \sim Expo(\lambda)$, find the value of λ , such that

$$P[X < 1] = P[Y < 1].$$

Solution. We have

$$f(x) = \frac{1}{2-0} = \frac{1}{2}$$
, $0 < x < 2$.

By the given hypothesis, we obtain

$$\int_{0}^{1} \frac{1}{2} dx = \int_{0}^{1} \lambda e^{-\lambda x} dx$$

$$\Rightarrow \frac{1}{2} = - \left| e^{-\lambda x} \right|_0^1 = - \left(e^{-\lambda} - 1 \right) = 1 - e^{-\lambda}$$

$$\Rightarrow e^{-\lambda} = \frac{1}{2} \Rightarrow -\lambda = \log\left(\frac{1}{2}\right) = -\log 2.$$

Hence $\lambda = \log 2$.

Example 3.7.5. Show that $Y = -(1/\lambda) \log F(x)$ is Expo (λ) . Solution. The m.g.f. of Y is

$$m_Y(t) = E\left[e^{tY}\right] = E \exp\left(tY\right) = E \exp\left[-\frac{t}{\lambda}\log F(x)\right]$$

$$= E \exp\left[\log\left\{F(x)\right\}^{-t/\lambda}\right]$$

$$= E\left\{\left\{F(x)\right\}^{-t/\lambda}\right\}$$

$$= E\left\{Z^{-t/\lambda}\right\}, Z = F(x) - U(0, 1)$$

$$= \int_0^1 z^{-t/\lambda} dz$$

$$= \left(1 - \frac{t}{\lambda}\right)^{-1}, \text{ which is m.g.f. of } Expo(\lambda).$$
[See Theorem 3.7.1.]

Hence Y is $Expo(\lambda)$.

EXERCISES

1. A random variable X has a uniform distribution over (-3, 3), find P[|X| < 2].

[Hint.
$$P[|X| < 2] = \int_{-2}^{2} \frac{1}{6} dx = \frac{2}{3}$$
.]

2. If X is uniformly distributed over (-a, a), determine a if

$$P[|X|>2] = \frac{3}{4}$$
 [Ans. $a=8$]

[Hint.
$$\frac{1}{4} = 1 - P[|X| \le 2] = 1 - P[-2 \le X \le 2] = 1 - \int_{-2}^{2} \frac{1}{2a} dx$$
]

3. For the rectangular distribution :

$$dF = k dx, 1 \le x \le 2$$

show that : A.M. > G.M. > H.M.

4. If X is uniformly distributed with mean 2 and variance 3, find $P\{X < 2\}.$

[Hint, Similar to Example 3.6.1.]

5. If X has the distribution

$$f(x) = \frac{5}{A}, \frac{-A}{10} \le x < \frac{A}{10}; f(x) = 0$$
 otherwise,

determine the constant A when

$$P[|X| \le 2] = 2 P[|X| > 2].$$
 [Ans. $A = 60$]
[Hint. We have $P[X \le 2] = 2 [1 - P[\{|X| \le 2\}]]$ or $\frac{2}{3} = P[X \le 2]$

or $\frac{2}{3} = \int_{0}^{2} (5/A) dx$.

If X ~ Expo (λ), find the value of k such that

$$\frac{P[X > k]}{P[X \le k]} = a. \qquad \left[\text{Ans. } k = \frac{1}{\lambda} \log (1 + a^{-1}) \right]$$
[Hint. $P[X \le k] = \int_{0}^{k} \lambda e^{-\lambda x} dx = 1 - e^{-\lambda k}, P[X > k] = e^{-\lambda k}$]

Find the median of the exponential distribution. [Ans. log 2/k]
 [Hint, If m be the median, then

$$\frac{1}{2} = \int_{0}^{m} \lambda e^{-\lambda x} dx = 1 - e^{-\lambda m}.$$

8. If $X \sim U(-1, 3)$ and $Y \sim \text{Expo}(\lambda)$. Find λ such that $\sigma_X^2 = \sigma_Y^2 \ i.e., \text{ var } [X] = \text{var } [Y]. \qquad [\text{Ans. } \lambda = 2/\sqrt{3}]$ $\left[\text{Hint. var } [X] = \frac{1}{12} (b - a)^2 = \frac{1}{12} (3 + 1)^2 = \frac{4}{3}, \text{ var } [Y] = \frac{1}{\lambda^2} \right]$

9. Show that the mean and variance of the simple gamma distribution

$$f(x) = \frac{e^{-x} x^{r-1}}{\Gamma(r)}, r > 0, 0 < x < \infty$$

is each r. Further show that $m_X(t) = (1-t)^{-r}$, |t| < 1.

[Take $\lambda = 1$ in the gamma density of Sec. 3.7.]

If X₁, X₂ are independent random variables having exponential distributions with parameters λ₁, λ₂; respectively, show that Z = min (X₁, X₂) has exponential distributions with parameter λ₁ + λ₂.

[Hint. Let G be c.d.f. of Z. Then

$$G(z) = P[Z \le z] = 1 - P[Z > z] = 1 - P[X_1 > z]P[X_2 > z]$$

$$=1-\left[\int\limits_{z}^{\infty}\lambda_{1}\,e^{-\lambda_{1}x}\,dx\right]\left[\int\limits_{z}^{\infty}\lambda_{2}\,e^{-\lambda_{2}x}\,dx\right]$$