Exponential Distribution

Let 'X' be a positive continuous random variable with interval $(0,\infty)$ is said to be Exponential distribution, having its p.d.f

$$f(x) = \frac{1}{\theta} e^{-\frac{x}{\theta}} \qquad \qquad 0 \le x \le \infty$$

It is only one parameter θ .

If $\frac{1}{\theta} = \theta$ then it is also exponential distribution

$$f(x) = \theta e^{-x\theta}$$

If $\theta = 1$ then it follow standard exponential distribution

$$f(x) = e^{-x}$$

This is also known as negative exponential distribution or single parameter exponential distribution.

Properties

- i) Exponential distribution is a continuous distribution.
- ii) The total area under the curve is unity.
- iii) The range of the distribution is 0 to ∞ .
- iv) It has one parameter θ .
- v) The mean of the exponential distribution is $E(x) = \theta$.
- vi) The variance of the exponential distribution is $Var(x) = \theta^2$.
- vii) The m.g.f of the exponential distribution is m.g.f = $(1 \theta)^{-1}$.

Prove that total area under the curve is unity

Proof: Let by definition:

$$Area = \int f(x)dx$$

As $x \approx \exp(\theta)$
$$f(x) = \frac{1}{\theta}e^{-\frac{x}{\theta}}$$

$$0 \le x \le \infty$$

$$Area = \int_{0}^{\infty} \frac{1}{\theta}e^{-\frac{x}{\theta}}dx$$

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As we know that gamma function is

$$\overline{)ab^a} = \int_0^\infty x^{a-1} e^{-x/b} dx \tag{B}$$

(A)

Comparing (A) & (B) and we get

$$a = 1 \quad \& b = \theta$$

$$\int ab^{a} = \int \overline{1}\theta^{1}$$
Put in (A)
Area = $\frac{1}{\theta}\theta = 1$
Hence Prove

Prepared by: Muhammad Riaz M.Sc: Statistics (Silver medalist) I.U.B (2006-08) Lecturer: Statistics NICAAS College (RYK) 03337368518

Find mean & variance

Solution: Let by definition:

$$E(x) = \int xf(x)dx = \frac{1}{\theta} \int_{0}^{\infty} x e^{\frac{-x}{\theta}} dx$$

$$E(x) = \frac{1}{\theta} \int_{0}^{\infty} x^{2^{-1}} e^{\frac{-x}{\theta}} dx$$
 (A)

As we know that gamma function is

$$\overline{ab^a} = \int_0^\infty x^{a-1} e^{-x/b} dx \tag{B}$$

Comparing (A) & (B) and we get

$$a = 2 \quad \& b = \theta$$

$$\int ab^{a} = \int 2\theta^{2}$$
Put in (A)
$$E(x) = \frac{1}{\theta} \int 2\theta^{2}$$

$$E(x) = \theta$$

$$Var(x) = E(x^{2}) - [E(x)]^{2}$$

$$E(x^{2}) = \int x^{2} f(x) dx = \frac{1}{\theta} \int_{0}^{\infty} x^{2} e^{\frac{-x}{\theta}} dx$$

$$E(x^{2}) = \frac{1}{\theta} \int_{0}^{\infty} x^{3-1} e^{\frac{-x}{\theta}} dx$$
(A)

As we know that gamma function is

$$\overline{)ab^a} = \int_0^\infty x^{a-1} e^{-\frac{x}{b}} dx$$
(B)

Comparing (A) & (B) and we get

a = 3 & b =
$$\theta$$

 $\overline{ab^{a}} = \overline{3\theta^{3}}$ Put in (A)
 $E(x^{2}) = \frac{1}{\theta}\overline{3\theta^{3}} = 2\theta^{2}$
 $Var(x) = E(x^{2}) - [E(x)]^{2} = 2\theta^{2} - (\theta)^{2} = \theta^{2}$
Find rth moments about origin. By use it fin

Find rth moments about origin. By use it finds mean & variance Solution: Let by definition r' ()

$$\mu_{r} = E(x^{r})$$

$$\mu_{r}' = \int x^{r} f(x) dx$$

$$\mu_{r}' = \frac{1}{\theta} \int_{0}^{\infty} x^{r} e^{-x/\theta} dx$$

$$\mu_{r}' = \frac{1}{\theta} \int_{0}^{\infty} x^{r+1-1} e^{-x/\theta} dx$$

(A)

As we know that gamma function is

$$\begin{aligned} & \overline{\rho}ab^{a} = \int_{0}^{\infty} x^{a-1}e^{\frac{-x}{b}}dx \end{aligned} \tag{B}$$

$$\begin{aligned} & \text{Comparing (A) \& (B) and we get} \\ & a = r+1 & \& b = \theta \\ & \overline{\rho}ab^{a} = \overline{\rho}r+1\theta^{r+1} \end{aligned} \qquad \text{Put in (A)} \end{aligned}$$

$$\begin{aligned} & \mu_{r}^{\ \prime} = \frac{1}{\theta}\overline{\rho}r+1.\theta^{r+1} \\ & \mu_{r}^{\ \prime} = \frac{1}{\theta}\overline{\rho}r+1.\theta^{r+1} \end{aligned} \qquad (C)$$

Use rth moments to find mean & variance

$$mean = \mu_1' = E(x)$$

$$\mu_r' = \overline{)r + 1}.\theta^r$$
Put r = 1 in eq (C)
$$\mu_1' = \overline{)1 + 1}.\theta^1 = \theta$$
Now, put r = 2 in eq.(C)
$$\mu_2' = \overline{)2 + 1}.\theta^2 = 2\overline{)2}.\theta^2 = 2\theta^2$$

$$\mu_2 = \mu_2' - (\mu_1')^2 = 2.\theta^2 - (\theta)^2 = \theta^2 = Var(X)$$

Find m.g.f of exponential distribution. Also find mean & variance by using m.g.f Solution: Let by definition

$$M_{x}(t) = m_{x}(t) = E(e^{tx}) = \int e^{tx} f(x) dx$$

As $x \approx \exp(\theta)$
$$f(x) = \frac{1}{\theta} e^{-\frac{x}{\theta}} \qquad 0 \le x \le \infty$$

$$M_{0}(t) = \frac{1}{\theta} \int_{0}^{\infty} e^{tx} e^{-\frac{x}{\theta}} dx$$

$$M_{0}(t) = \frac{1}{\theta} \int_{0}^{\infty} e^{tx^{-\frac{x}{\theta}}} dx$$

$$M_{0}(t) = \frac{1}{\theta} \int_{0}^{\infty} e^{-\frac{x}{\theta}(1-\theta t)} dx$$

$$M_{0}(t) = \frac{1}{\theta} \int_{0}^{\infty} x^{1-1} e^{-\frac{x}{\theta}(1-\theta t)^{-1}} dx$$

$$M(t) = \frac{1}{\theta} \int_{0}^{\infty} x^{1-1} e^{-\frac{x}{\theta}(1-\theta t)^{-1}} dx$$

As we know that gamma function is

$$\int \overline{a}b^a = \int_0^\infty x^{a-1} e^{-x/b} dx$$
(B)

(A)

Comparing (A) & (B) and we get

$$\begin{aligned} a = 1 \quad \& b = o^{(1-a)^{-1}} \\ \overline{ab}^{a} = \overline{j1} \left\{ o^{(1-a)^{-1}} \right\} & \text{Put in (A)} \\ m_{x}(t) = \frac{1}{a} \overline{j1} \left\{ o^{(1-a)^{-1}} \right\} = (1-a)^{-1} & \text{Required m.g.f.} \\ \text{Use it to find mean & variance} \\ E(x) = \mu_{1}^{'} = \left[\frac{d}{dx} m_{x}(t) \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x) = \left[\frac{d}{dx} (1-a)^{-1} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x) = \left[\frac{d}{dx} (1-a)^{-2} \right]_{t=0}^{t=0} & \text{Required mean} \\ Again differentiate eq(e) w.r.t to 't' \\ E(x^{2}) = \mu_{2}^{'} = \left[\frac{d^{2}}{dt^{2}} M_{x}(t) \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{dt} \left[\theta(1-a)^{-2} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{dt} \left[\theta(1-a)^{-2} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{dt} \left[\theta(1-a)^{-2} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{dt} \left[\theta(1-a)^{-2} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{dt} \left[\theta(1-a)^{-2} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{dt} \left[\theta(1-a)^{-2} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{dt} \left[\theta(1-a)^{-2} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{2} \left[-2\theta(1-a)^{-3} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{2} \left[-2\theta(1-a)^{-1} \right]_{t=0}^{t=0} & \text{Required mean} \\ E(x^{2}) = \frac{d}{2} \left[-2\theta(1-a)^{-1} \right]_{t=0}^{t=0} & \text{Required mean} \\ \text{Solution: Let by definition} & k(t) = \log \left[(1-a)^{-1} \right]_{t=0} & \text{Required mean} \\ k(t) = \log \left[(1-a)^{-1} \right] \\ k(t) = -\log (1-a)^{t=0} & \text{Required mean} \\ \text{Required mean} & \text{Required mean} \\ k(t) = -\log (1-a)^{2} & \frac{1}{2} - \frac{x^{3}}{3} - \frac{x^{4}}{4} \dots \\ \text{Required mean} & \frac{1}{2} & \frac{1}{2} - \frac{x^{2}}{2} - \frac{x^{3}}{3} - \frac{x^{4}}{4} \dots \\ k(t) = -\left[-\left(a\right) - \frac{(a)^{2}}{2} - \frac{(a)^{3}}{3} - \frac{(a)^{4}}{4} + \dots \right] \\ k(t) = \left[\theta + \theta^{2} \frac{t^{2}}{2} + \theta^{3} \frac{t^{3}}{3} + \theta^{4} \frac{t^{4}}{4} \dots \right] \\ k(t) = \left[\theta + \frac{1}{1!} + \theta^{2} \frac{t^{2}}{2!} + 2\theta^{3} \frac{t^{3}}{3!} + \theta^{4} \frac{t^{4}}{4!} \dots \right]$$
 (A)

As we know general expression of cummulent

$$k(t) = \left[k_1 \frac{t}{1!} + k^2 \frac{t^2}{2} + k_3 \frac{t^3}{3} + k_4 \frac{t^4}{4} \dots\right]$$
(B)

By comparing (A) & (B) and we get

$$k_1 = \mu_1' = \theta$$

©

$$k_{2} = \mu_{2} = \theta^{2} = Var(X)$$

$$k_{3} = \mu_{3} = 2\theta^{3}$$

$$k_{4} = \mu_{4}' = 6\theta^{4}$$

$$\mu_{4} = k_{4} + 3k_{2}^{2} = 6\theta^{4} + 3(\theta^{2})^{2}$$

$$\mu_{4} = 6\theta^{4} + 3\theta^{4} = 9\theta^{4}$$

Moment Ratio

$$\beta_1 = \frac{{\mu_3}^2}{{\mu_2}^3} = \frac{\left(2\theta^3\right)^2}{\left(\theta^2\right)^3} = \frac{4\theta^6}{\theta^6} = 4$$

 $\mu_3 = 2\theta^3$ It is positive so distribution positively skewed

$$\beta_{2} = \frac{\mu_{4}}{{\mu_{2}}^{2}} = \frac{9\theta^{4}}{(\theta^{2})^{2}} = \frac{9\theta^{4}}{\theta^{4}} = 9$$

As $\beta_2 > 3$ so the distribution is leptokurtic.

Find mode of exponential distribution

Solution: Let by definition If following two conditions are satisfied then mode exists.

$$f(x') = 0 \quad \text{or} \quad \frac{d}{dx} \log f(x) = 0$$

$$f(x') < 0 \quad \text{or} \quad \frac{d^2}{dx^2} \log f(x) < 0$$
As $x \approx \exp(\theta)$

$$f(x) = \frac{1}{\theta} e^{-\frac{x}{\theta}} \qquad 0 \le x \le \infty$$
Taking log on both sides
$$\log f(x) = \log\left[\frac{1}{\theta}e^{-\frac{x}{\theta}}\right]$$

$$\log f(x) = \left[\log\left(\frac{1}{\theta}\right) - \frac{x}{\theta}\log e\right] \qquad :-\log_e = 1$$

$$\log f(x) = \left[\log\left(\frac{1}{\theta}\right) - \frac{x}{\theta}\right]$$
Differentiate w.r.t to 'x'
$$\frac{d}{dx}\log f(x) = \frac{d}{dx}\left[\log\left(\frac{1}{\theta}\right) - \frac{x}{\theta}\right]$$

$$\frac{d}{dx}\log f(x) = \left[-\frac{1}{\theta}\right]$$

It means that mode of exponential distribution does not exist.

State & prove Memory less property of Exponential Distribution Statement:

If 'x' follows the negative exponential distribution with p.d.f $f(x) = \theta e^{-x\theta}$ $0 \le x \le \infty$ then by definition

$$P\left\lfloor\frac{x > (a+b)}{x > a}\right\rfloor = P(x > b) \qquad \text{by law of complement}$$

$$\frac{1 - P\left[x \le (a+b)\right]}{1 - P(x \le a)} = \frac{1 - F(a+b)}{1 - F(a)} \qquad (i)$$

Let 'x' be continuous random variable having the p.d.f $f(x) = \theta e^{-x\theta}$ its c.d.f is

$$\begin{split} F(x) &= \int_{0}^{x} \theta \, e^{-x\theta} \, dx \\ F(x) &= \theta \, \frac{e^{-x\theta}}{-\theta} \Big|_{0}^{x} \\ F(x) &= 1 - e^{-x\theta} \\ F(x) &= 1 - e^{-x\theta} \\ F(a) &= 1 - e^{-a\theta} \\ F(a+b) &= 1 - e^{-(a+b)\theta} \\ F(b) &= 1 - e^{-b\theta} \\ F(b) &= 1 - e^{-b\theta} \\ Put \text{ in (i)} \\ \frac{1 - F(a+b)}{1 - F(a)} &= \frac{1 - \left(1 - e^{-(a+b)\theta}\right)}{1 - \left(1 - e^{-a\theta}\right)} \\ \frac{1 - F(a+b)}{1 - F(a)} &= \frac{1 - 1 + e^{-(a+b)\theta}}{1 - 1 + e^{-a\theta}} \\ \frac{1 - F(a+b)}{1 - F(a)} &= e^{-a\theta - b\theta + a\theta} = e^{-b\theta} \\ \frac{1 - F(a+b)}{1 - F(a)} &= 1 - \left(1 - e^{-b\theta}\right) = 1 - F(b) = 1 - P(x \le b) = P(x > b) \\ \frac{2^{nd} \text{ Method}}{2^{nd} \text{ Method}} \end{split}$$

Suppose A be the event such that (x>a) & B is another event such that x>(a+b). i.e B is a subset of A. Then we consider

$$P\left[\frac{B}{A}\right] = \frac{P(A \cap B)}{P(A)} = \frac{P(B)}{P(A)}$$

Because B is a subset of A, replacing A = x>a & B = x> a+b we get
$$P\left[\frac{x > (a+b)}{x > a}\right] = \frac{P(x > a+b)}{P(x > a)}$$

Then c.d.f of Negative Exp.Distribution

$$P(X \le x) = \int_{-\infty}^{\infty} f(x) dx = \int_{0}^{x} \theta e^{-x\theta} dx = \theta \frac{e^{-x\theta}}{-\theta} \Big|_{0}^{x} = 1 - e^{-x\theta}$$

Then by compliment we know that
$$P(X \le x) = 1 - e^{-x\theta}$$
$$P(x > a) = 1 - P(X \le x) = 1 - 1 - e^{-a\theta} = e^{-a\theta}$$

Similarly
$$P(x > a + b) = e^{-(a+b)\theta}$$
$$\frac{P(x > a + b)}{P(x > a)} = \frac{e^{-(a+b)\theta}}{e^{-a\theta}} = e^{-a\theta - b\theta + a\theta} = e^{-b\theta} = P(x > b)$$

Hence, Memory less property is proved.

Prepared by: Muhammad Riaz M.Sc: Statistics (Silver medalist) I.U.B (2006-08) Lecturer: Statistics NICAAS College (RYK) 03337368518