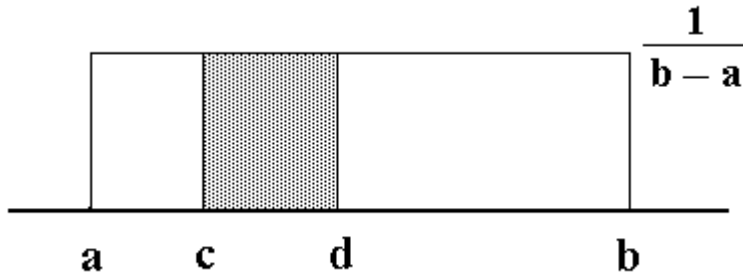


Uniform Distribution over an interval $[a, b]$:



For Uniform distribution,

$$P(c \leq X \leq d) = \frac{d-c}{b-a}, \quad a \leq c \leq d \leq b.$$

$$E(X) = \frac{a+b}{2}, \quad \text{Var}(X) = \frac{(b-a)^2}{12}.$$

0.25. Let X be a random variable distributed uniformly over the interval $[a, b]$. Find the moment-generating function of X .

$$\begin{aligned} M_X(t) &= E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} \cdot f(x) dx = \int_a^b e^{tx} \cdot \frac{1}{b-a} dx \\ &= \frac{1}{b-a} \cdot \left(\frac{e^{tx}}{t} \right) \Big|_a^b = \frac{e^{tb} - e^{ta}}{t(b-a)}, \quad t \neq 0. \end{aligned}$$

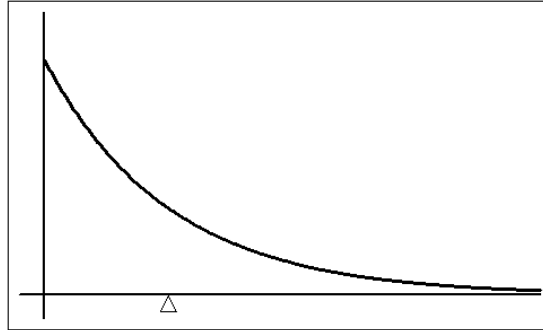
$$M_X(0) = 1.$$

Exponential Distribution:

$$f(x) = \begin{cases} \frac{1}{\theta} e^{-x/\theta} & \text{for } x \geq 0 \\ 0 & \text{otherwise} \end{cases} \quad f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{for } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

$$E(X) = \theta,$$

$$\text{Var}(X) = \theta^2.$$



$$E(X) = 1/\lambda,$$

$$\text{Var}(X) = 1/\lambda^2.$$

0.50. Find the moment generating function of an exponential random variable.

$$\begin{aligned} M_X(t) &= E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} \cdot f(x) dx = \int_0^{\infty} e^{tx} \cdot \frac{1}{\theta} e^{-x/\theta} dx \\ &= \frac{1}{\theta} \cdot \int_0^{\infty} e^{x(t-1/\theta)} dx = \frac{1}{\theta} \cdot \left(\frac{e^{x(t-1/\theta)}}{t-1/\theta} \right) \Big|_0^{\infty} = \frac{1}{1-t\theta}, \quad t < 1/\theta. \end{aligned}$$

OR

$$\begin{aligned} M_X(t) &= E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} \cdot f(x) dx = \int_0^{\infty} e^{tx} \cdot \lambda e^{-\lambda x} dx \\ &= \lambda \cdot \int_0^{\infty} e^{x(t-\lambda)} dx = \lambda \cdot \left(\frac{e^{x(t-\lambda)}}{t-\lambda} \right) \Big|_0^{\infty} = \frac{\lambda}{\lambda-t}, \quad t < \lambda. \end{aligned}$$

0.75. Suppose X has Exponential distribution with mean θ .

$$\Rightarrow f_X(x) = \frac{1}{\theta} e^{-x/\theta}, \quad x > 0.$$

a) Find $P(X > t)$ for $t > 0$.

$$P(X > t) = \int_t^{\infty} \frac{1}{\theta} e^{-x/\theta} dx = \left(-e^{-x/\theta} \right) \Big|_t^{\infty} = e^{-t/\theta}, \quad t > 0.$$

b) Show that for positive t and s ,

$$P(X > t + s \mid X > t) = P(X > s)$$

(*memoryless* property).

$$\begin{aligned} P(X > t + s \mid X > t) &= \frac{P(X > t + s \cap X > t)}{P(X > t)} = \frac{P(X > t + s)}{P(X > t)} \\ &= \frac{e^{-(t+s)/\theta}}{e^{-t/\theta}} = e^{-s/\theta} = P(X > s). \end{aligned}$$

Compare to Geometric distribution:

a) For Geometric(p), $P(X > a) = (1 - p)^a$, $a = 0, 1, 2, 3, \dots$.

b) For positive integers a and b ,

$$P(X > a + b \mid X > a) = P(X > b)$$

(*memoryless* property).

1. Suppose the lifetime of a particular brand of light bulbs is exponentially distributed with mean of 400 hours.

$$\text{Mean} = 400 \Rightarrow \theta = 400, \lambda = 1/400.$$

- a) Find the probability that a randomly selected light bulb would last over 500 hours.

$$\text{Exponential, } \theta = 400. \quad P(T > 500) = e^{-500/400} = e^{-1.25} \approx \mathbf{0.2865}.$$

$$\text{OR} \quad P(T > 500) = \int_{500}^{\infty} \frac{1}{400} e^{-x/400} dx = \dots$$

- b) Find the probability that 3 out of 7 randomly selected light bulbs would last over 500 hours.

$$\text{Binomial, } n = 7, p = 0.2865.$$

$$P(X = 3) = \binom{7}{3} \cdot 0.2865^3 \cdot 0.7135^4 \approx \mathbf{0.2133}.$$

- c) Find the probability that a randomly selected light bulb would last between 300 hours and 800 hours.

$$\text{Exponential, } \theta = 400.$$

$$\begin{aligned} P(300 < T < 800) &= P(T > 300) - P(T > 800) = e^{-300/400} - e^{-800/400} \\ &= e^{-0.75} - e^{-2.0} \approx \mathbf{0.337}. \end{aligned}$$

$$\text{OR} \quad P(300 < T < 800) = \int_{300}^{800} \frac{1}{400} e^{-x/400} dx = \dots$$

2. An insurance policy reimburses a loss up to a benefit limit of C and has a deductible of d . The policyholder's loss, X , follows a distribution with density function $f(x)$. Find the expected value of the benefit paid under the insurance policy?

$$\text{Benefit Paid} = \begin{cases} 0 & x < d \\ x - d & d \leq x < C + d \\ C & x \geq C + d \end{cases}$$

$$E(\text{Benefit Paid}) = \int_0^d 0 \cdot f_X(x) dx + \int_d^{C+d} (x-d) \cdot f_X(x) dx + \int_{C+d}^{\infty} C \cdot f_X(x) dx.$$

For example, if X has an Exponential distribution with mean θ ,

$$\begin{aligned} E(\text{Benefit Paid}) &= \int_d^{C+d} (x-d) \cdot f_X(x) dx + \int_{C+d}^{\infty} C \cdot f_X(x) dx \\ &= \int_d^{C+d} (x-d) \cdot \frac{1}{\theta} e^{-x/\theta} dx + \int_{C+d}^{\infty} C \cdot \frac{1}{\theta} e^{-x/\theta} dx \\ &= \left(-(x-d) \cdot e^{-x/\theta} - \theta e^{-x/\theta} \right) \Big|_d^{C+d} + \left(-C e^{-x/\theta} \right) \Big|_{C+d}^{\infty} \\ &= \theta e^{-d/\theta} - \theta e^{-(C+d)/\theta}. \end{aligned}$$

For example, if $d = 2$, $C = 10$, X has an Exponential distribution with mean $\theta = 5$,

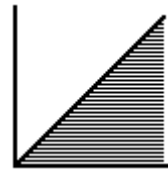
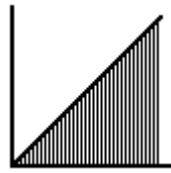
$$\begin{aligned} E(\text{Benefit Paid}) &= \int_2^{12} (x-2) \cdot f_X(x) dx + \int_{12}^{\infty} 10 \cdot f_X(x) dx \\ &= \int_2^{12} (x-2) \cdot \frac{1}{5} e^{-x/5} dx + \int_{12}^{\infty} 10 \cdot \frac{1}{5} e^{-x/5} dx \\ &= \left(-(x-2) \cdot e^{-x/5} - 5 e^{-x/5} \right) \Big|_2^{12} + \left(-10 e^{-x/5} \right) \Big|_{12}^{\infty} \\ &= 5 e^{-2/5} - 5 e^{-12/5} = 5 e^{-0.4} - 5 e^{-2.4} \approx \mathbf{2.898}. \end{aligned}$$

Fact: Let X be a nonnegative continuous random variable with p.d.f. $f(x)$ and c.d.f. $F(x)$. Then

$$E(X) = \int_0^{\infty} (1 - F(x)) dx.$$

Proof:

$$E(X) = \int_0^{\infty} x f(x) dx = \int_0^{\infty} \left(\int_0^x dy \right) f(x) dx = \int_0^{\infty} \left(\int_0^x f(x) dy \right) dx$$



$$\int_0^{\infty} \left(\int_0^x f(x) dy \right) dx = \int_0^{\infty} \left(\int_y^{\infty} f(x) dx \right) dy$$

$$\Rightarrow E(X) = \int_0^{\infty} \left(\int_y^{\infty} f(x) dx \right) dy = \int_0^{\infty} P(X > y) dy = \int_0^{\infty} (1 - F(y)) dy.$$

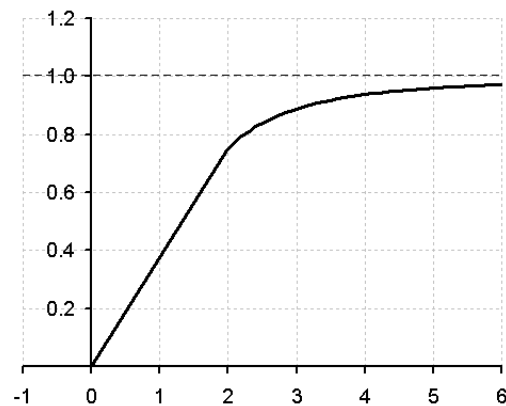
Example: (2. from Examples for 3.3)

Let X be a continuous random variable with the cumulative distribution function

$$F(x) = 0, \quad x < 0,$$

$$F(x) = \frac{3}{8} \cdot x, \quad 0 \leq x \leq 2,$$

$$F(x) = 1 - \frac{1}{x^2}, \quad x > 2.$$



$$\begin{aligned} \text{c) } E(X) &= \int_0^{\infty} (1 - F(x)) dx = \int_0^2 \left(1 - \frac{3}{8} \cdot x \right) dx + \int_2^{\infty} \left(\frac{1}{x^2} \right) dx \\ &= 2 - \frac{3}{8} \cdot \left(\frac{x^2}{2} \right) \Big|_0^2 + \left(-\frac{1}{x} \right) \Big|_2^{\infty} = 2 - \frac{3}{4} + \frac{1}{2} = \frac{7}{4} = \mathbf{1.75}. \end{aligned}$$

Example: Find the expected value of an Exponential (θ) distribution.

For Exponential (θ), $1 - F(x) = P(X > x) = e^{-x/\theta}, \quad t > 0.$

$$E(X) = \int_0^{\infty} (1 - F(x)) dx = \int_0^{\infty} e^{-x/\theta} dx = \theta.$$

Fact: Let X be a random variable of the discrete type with pmf $p(x)$ that is positive on the nonnegative integers and is equal to zero elsewhere. Then

$$E(X) = \sum_{x=0}^{\infty} [1 - F(x)],$$

where $F(x)$ is the cdf of X .

“Proof”:

$$1 - F(x) = P(X > x) = p(x+1) + p(x+2) + p(x+3) + p(x+4) + \dots$$

$$1 - F(0) \quad p(1) + p(2) + p(3) + p(4) + p(5) + p(6) + p(7) + \dots$$

$$1 - F(1) \quad p(2) + p(3) + p(4) + p(5) + p(6) + p(7) + \dots$$

$$1 - F(2) \quad p(3) + p(4) + p(5) + p(6) + p(7) + \dots$$

$$1 - F(3) \quad p(4) + p(5) + p(6) + p(7) + \dots$$

$$1 - F(4) \quad p(5) + p(6) + p(7) + \dots$$

... ..

$$\Rightarrow \sum_{x=0}^{\infty} [1 - F(x)] = 1 \times p(1) + 2 \times p(2) + 3 \times p(3) + 4 \times p(4) + \dots = E(X).$$

Example: Find the expected value of a Geometric (p) distribution.

For Geometric (p), $1 - F(x) = P(X > x) = (1 - p)^x, \quad x = 0, 1, 2, \dots$

$$E(X) = \sum_{x=0}^{\infty} [1 - F(x)] = \sum_{x=0}^{\infty} [1 - p]^x = \frac{1}{1 - [1 - p]} = \frac{1}{p}.$$