

Exponential r.v.s are used to model the time between occurrences of an event (arrivals, failures) and ~~or~~ the lifetime of devices.

In most situations, λ is in units event / (time unit) where the time unit is seconds, minutes, hours, etc.

\Rightarrow This means $\frac{1}{\lambda}$ is in units (time unit) / event. Sometimes $\frac{1}{\lambda}$ is referred to as the average arrival time, average lifetime, etc...

The exponential r.v. satisfies the memoryless property

$$\Pr(X > t_0 + t \mid X > t_0) = \Pr(X > t)$$

The L.H.S gives the probability of having to wait at least t after waiting t_0 given one has already been waiting t_0 . The R.H.S gives the probability of waiting at least t from the start. Thus the probability of waiting at least t is the same regardless of when you start.

$$\begin{aligned} \text{Pf: } \Pr(X > t_0 + t \mid X > t_0) &= \frac{\Pr(\{X > t_0 + t\} \cap \{X > t_0\})}{\Pr(X > t_0)} \\ &= \frac{\Pr(\{X > t_0 + t\})}{\Pr(X > t_0)} \\ &= \frac{e^{-\lambda(t_0 + t)}}{e^{-\lambda t_0}} \\ &= e^{-\lambda t} = \Pr(X > t) \end{aligned}$$

The memory less property can also written as

$$f_x(x | x > t_0) = f_x(x - t_0), t_0 > 0$$

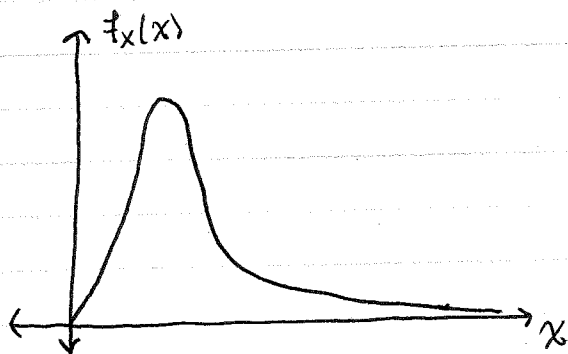
Note: The exponential r.v. is the only continuous r.v. that satisfies the memory less property.

Rayleigh Distribution

The Rayleigh distribution is often found to model the amplitude of received communication signals.

A Rayleigh r.v. X with parameter α has p.d.f.

$$f_x(x) = \frac{x}{\alpha^2} e^{-x^2/2\alpha^2} u(x), \alpha > 0$$



$$E[X] = \alpha \sqrt{\pi/2}$$

$$\text{Var}[X] = (2 - \pi/2) \alpha^2$$

Important Discrete Random Variables

Bernoulli R.V.

Let A be an event. The Bernoulli r.v. is the value of $\mathbb{1}_A$ (indicator function of A).

A Bernoulli r.v. with $\Pr(A) = p$ has pmf

$$P_X(x_i) = \begin{cases} 1-p, & x_i = 0 \\ p, & x_i = 1 \end{cases}$$

$$E[X] = p, \quad \text{Var}[X] = p(1-p)$$

Bernoulli random variables are useful in the analyses of sequences of independent random experiments.

Binomial R.V.

Used to model the number of occurrences of an event A in n independent trials of a random experiment. A binomial r.v. X with $\Pr(A) = p$ has pmf

$$P_X(k) = \binom{n}{k} p^k (1-p)^{n-k}, \quad k = 0, 1, \dots, n$$

$$E[X] = np, \quad \text{Var}[X] = np(1-p)$$

Geometric R.V.

Used to model the number of trials (or failures) until the occurrence of an event A , in independent trials of a random experiment.

A geometric r.v. X with $\Pr(A) = p$ has pmf

1st version (trials)

$$P_X(m) = (1-p)^{m-1} p, \quad m = 1, 2, \dots$$

$$E[X] = 1/p \quad \text{Var}[X] = (1-p)/p^2$$

2nd version (failures)

$$P_X(m) = (1-p)^m p, \quad m = 0, 1, \dots$$

$$E[X] = 1/p \quad \text{Var}[X] = (1-p)/p^2$$

The geometric r.v. satisfies the memory less property

$$\Pr(X \geq m_0 + m \mid X > m_0) = \Pr(X \geq m)$$

Interpretation is similar to that of the exponential r.v., i.e., the probability of having to perform at least m trials is the same regardless of how many trials have already been performed.

Pf. Left as exercise

Uniform R.V.

Used to model equally likely outcomes.

A uniform r.v. X on the set $S_X = \{x_1, x_2, \dots, x_n\}$ has pmf

$$P_X(x_i) = \frac{1}{n}, \quad x_i \in \{x_1, x_2, \dots, x_n\}$$

If $S_X = \{1, 2, \dots, n\}$

$$E[X] = \frac{n+1}{2}, \quad \text{Var}[X] = \frac{n^2-1}{12}$$

Characteristic Function

Transform methods are useful aids in the solution of equations with derivative and integrals.

We will discuss a Fourier-like transform of pdf's. This method will become very useful when discussing sums of random variables.

The characteristic function of r.v. X is defined as

$$\varphi_X(\omega) = E[e^{j\omega X}], \quad j = \sqrt{-1}$$

$$= \int_{-\infty}^{\infty} f_X(x) e^{j\omega x} dx$$

$\varphi_X(\omega)$ can be viewed as the expected value of $e^{j\omega X}$ or * as the "Fourier" transform of $f_X(x)$. We can find $f_X(x)$ from $\varphi_X(\omega)$ as

$$f_X(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \varphi_X(\omega) e^{-j\omega x} d\omega$$