

Lectures of May 25th, 2004

Scribe: Alaa Farhat, Ara Tchobanian, Graham Daniel Cran

1 Understanding α and scaling

Example 1 *If the probability of getting calls is 1.3 on average over 2 hours. Whats the probability of getting 3 calls over a 5 hour interval?*

Given that the following assumptions are true:

- Independence along time
- Equal likelihood on each time-point

Solution

Over 2 hour interval:

Probability is: $P_X(k=3)$ with $\alpha = 1.3$

Over 5 hour interval:

Probability is: $P_X(k=3)$ but with a new adjusted α , so that:

$$\alpha = (1.3) \cdot \left(\frac{2}{5}\right)$$

2 Continuing the 'Big Example' from previous lecture

2.5-a

With the same assumption that every interval of length 2 has α phone calls on average, consider the continuous version of 2.3. i.e. When will the first phone call arrive?

Solution

Let the answer be X , which is a RV. Note: X is a RV.

In 2.3, for any positive integer k , we have:

$$P[X > k\Delta] = P[\text{no call hits the first } k \text{ intervals}] = (1 - P)^k$$

When Δ is small enough, any time t can be approximated by

$$t = k\Delta, \text{ for some } k$$

Then,

$$P[X > 1] = (1 - P)^{\left(\frac{1}{\Delta}\right)} = \left(1 - \left(\frac{\alpha}{n}\right)\right)^{\left(\frac{1}{\Delta}\right)} = \left(1 - \left(\frac{\alpha}{n}\right)\right)^{\left(\frac{tn}{2}\right)}$$

where, n = number of intervals of Δ in 2 hours distribution

$$\text{As } n \rightarrow \infty, (\exp^{-\alpha})^{\left(\frac{1}{\Delta}\right)} = (\exp)^{\left(-\frac{\alpha}{2}t\right)}$$

note: $\frac{\alpha}{2} := \lambda$, average number of calls arriving over any duration of 1 time unit (in this case 1 hour)

The CDF of X is then,

$$P_X(x) = P[x \leq k] = 1 - P[X > x] = 1 - \exp^{-\lambda x}, \text{ for } x \geq 0$$

The pdf of X is then,

$$f_X(x) = \lambda \exp^{-\lambda x}, \text{ for } x \geq 0 \\ = 0, \text{ otherwise}$$

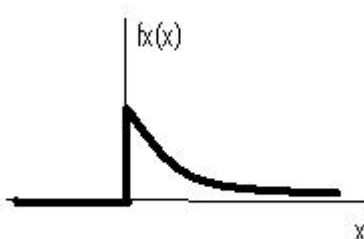


Figure 1. The PDF of X

This example shows that, with the 2 mentioned assumptions, we will have to end up with an exponential distribution of the RV in terms of time.

Remarks:

	Discrete-time (in terms of # of trials)	Continuous-time (in terms of # of trials)
Order of n trials [how many]	<i>Binomial RV</i>	<i>Poisson RV</i>
How long to wait until the first successful trial (ex. "Heads"/"call") [When]	<i>Geometric RV</i>	<i>Exponential RV</i>

2.5-b

Suppose that 5 hours have passed, and there were no calls, when will the first call arrive?

Solution:

The answer X is a RV having the following "memoryless" property:

$$P[x > 5 + b | x > 5] = P[x > b], \text{ for any positive } b$$

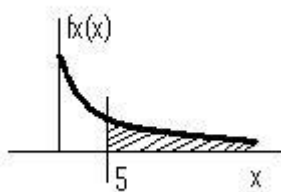


Figure 2. The PDF of X

The PDF is:

$$f_{x|x>5}(x) = \beta f_x(x - 5) \text{ for some } \beta : \int_{x>5} f_{x|x>5}(x) dx = 1$$

Note in this example the fact that no calls arrive until $x=5$ does not mean we will get one very soon (Independence). So if we scale the conditional PDF above so that the total area = 1, we will get the original PDF.

3 Continuing Types of Random Variables

6. Gaussian RV

→ Continuous RV

→ $S_x = (-\infty, +\infty)$

→ A Gaussian RV is specified by the following PDF:

$$f_x(x) = \frac{1}{\sqrt{2\pi}\sigma} \exp^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

Parametrized by:

- μ : The mean (average) of the RV
- σ : The standard deviation of the RV

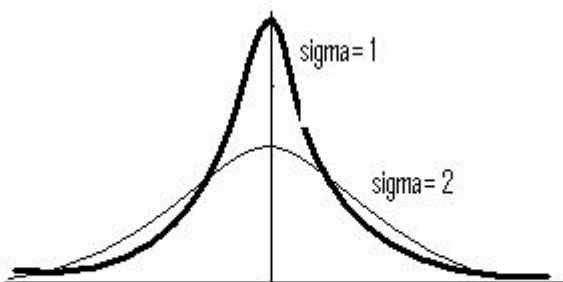


Figure 2. PDF of a Gaussian RV

The Gaussian RV is typically used to model the sum of many independent RV's or simply assumed for convenience.

Q Functions

For any x :

$$Q(x) = \int \frac{1}{\sqrt{2\pi}} \exp^{-\frac{x^2}{2}}$$

That is, $Q(x)$ is the rite tail of the Gaussian PDF with $\sigma = 1$

Example 2 Suppose we chose the meter(m) to be the unit of distance. Suppose the position of a bug on the x -axiz is a Gaussing RV with $\mu = 0$ and $\sigma^2 = 1$ m²

i.e.

$$f_x(x) = \frac{1}{\sqrt{2\pi}} \exp^{-\frac{x^2}{2}}$$

Find the probability that the bug is more than 2 meters to the rite of the origin

Answer is $Q(2)$

Now suppose that we change the unit of distance to mm. Answer the same question aain.

Clearly the answer is still $Q(2)$ but the distribution is different.

Noe: the PDF expressed using the 1 mm is different:

$$f_x(x) = \frac{1}{\sqrt{2\pi}100} \exp^{-\frac{x^2}{2 \cdot 10000}} dx$$

4 Mean and Variance of a RV

Suppose $f_x(x)$ is the PDF of a RV X , then:

- The mean (also called the expected value) of x is defined as:

$$E[x] = \int_{-\infty}^{+\infty} x f_X(x) dx$$

- The variance of X is defined as:

$$VAR[X] = E[X^2] - E[X]^2 = \int_{-\infty}^{+\infty} (x - E[x])^2 f_X(x) dx$$

- The standard deviation of X is defined as:

$$STD[X] = \sqrt{VAR[X]}$$

Example 3

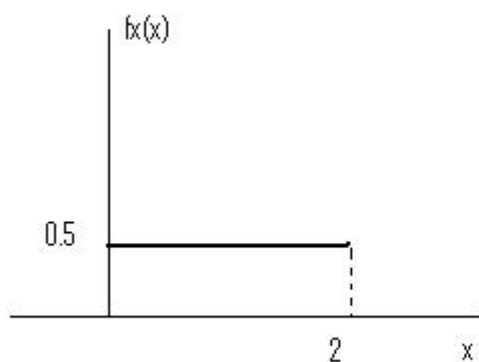


Figure 4. The PDF of X

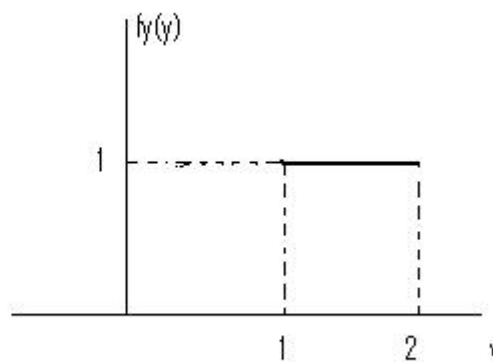


Figure 5. The PDF of Y

Let X and Y be two RV's with PDF given in the figures above. Now we will generate RV Z as follows:

Toss a coin with $P[H] = 0.3$

if the outcome is H , draw a number X and let $z = x$ if the outcome is T , draw a number Y and let $z = y$

Find the mean of Z .

$$P[Z = z] = P[Z = z|C = H].P[C = H] + P[Z = z|C = T].P[C = T]$$

$$F_Z(z) = f_x(z)$$

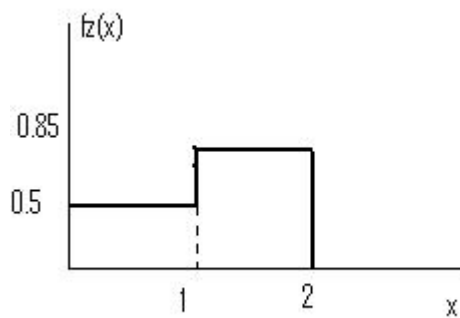


Figure 5. The combined PDF of X and Y with total area = 1

$$E[x] = \int_{-\infty}^{+\infty} f_z(z) dz = \int_0^1 0.15z dz + \int_1^2 0.85z dz$$

$$STD[X] = \int_{-\infty}^{+\infty} [X - E[x]]^2 f_x dx$$