Logarithm of exponential and Cauchy random variables

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1. Introduction.

Exponential distribution is frequently used in reliability engineering. Lognormal distribution is also used. It is the distribution, the logaritm of which has a normal distibution. What if we take a logarithm of exponential distribution. We want to calculate the first moment and the second moment.

It can be done in two ways. One way is to integrate the transormation function, which is the logaritm in this case, multiplied by the density function. The other way is to find the density function of a new variable and use it to calculate the first moment. Both ways yield the same result.

The same idea is applied to the Cauchy distribution. The moments of the Cauchy distribution are not defined but the moments of the logarithm of its absolute value can be calculated.

2. Exponential distribution.

The density function of exponentially distributed random variable X is defined as $\mu e^{-\mu x}$ for $x \geq 0$, equal to zero for $x < 0$, provided $\mu > 0$. We will study the case when $\mu = 1$, that is, the density is just e^{-x}. We define the new variable Y as $Y = \ln(X)$.

1) The first way to define the first moment of Y is

$$E(Y) = \int_0^\infty e^{-x} \ln(x) dx = -\gamma$$

where γ is the Euler constant.

The proof may be found in Nahin (2015) p.173. It is interesting to find out that $\gamma = -0.5772\ldots$ appears also in statistics.

2) The second way is described in most textbook on probability theory. Let $F_X(x)$ denote the cdf of the random variable X and $f_X(x)$ its pdf. Assume $g(x)$ is the transformation of X giving us a random variable $Y = g(X)$. We denote by $F_Y(y)$ a cdf of Y and its pdf by $f_Y(y)$. We assume that $g(x)$ is differentiable and increasing, its inverse is denoted by $h(y) = g^{-1}(y)$.

Then we differentiate a composite function

$$f_Y(y) = \frac{d}{dy} F_Y(y) = \frac{d}{dy} F_X(h(y)) = f_X(h(y)) h'(y).$$

In our case we have

$$f_X(x) = e^{-x}, g(x) = \ln(x), h(y) = e^y$$

and the density function of Y is

$$f_Y(y) = e^{-e^y} e^y = e^{y - e^y}.$$

We just verify that

$$\int_{-\infty}^{\infty} e^{y-e^y} dy = -e^{-e^y}\Big|_{-\infty}^{\infty} = -(0-1) = 1.$$

We write the expectation of Y in both ways as

$$E(Y) = \int_{-\infty}^{\infty} y e^{y-e^y} dy = \int_0^{\infty} e^{-x} \ln(x) dx = -\gamma.$$

We have actually calculated the integral

$$\int_{-\infty}^{\infty} y e^{y-e^y} dy = -\gamma.$$

The second moment is

$$M_2(Y) = \int_0^{\infty} e^{-x}(\ln(x))^2 dx = \frac{\pi^2}{6} + \gamma^2.$$

The proof can be found at
https://www.youtube.com/watch?v=wUY6TJxfFTk

The second moment can also be written as

$$\int_{-\infty}^{\infty} y^2 e^{y-e^y} dy = \frac{\pi^2}{6} + \gamma^2.$$

The fact that the second moment is finite means, by the way, that we are free to use the central limit theorem.

3. Cauchy distribution.

The pdf of the standard Cauchy distribution X is

$$f(x) = \frac{1}{\pi(x^2+1)}$$

defined for any x. If consider the absolute value, the pdf of $|X|$ is

$$f_A(x) = \frac{2}{\pi(x^2+1)}$$

defined this way for $x \geq 0$, otherwise $f_A(x) = 0$ for $x < 0$.

We study the transformation $Y = \ln|X|$. First of all we calculate the expectation as

$$E(Y) = \int_0^{\infty} \ln(x) f_A(x) dx = \int_0^{\infty} \frac{2\ln(x)}{\pi(x^2+1)} dx = \frac{2G}{\pi}.$$

$G = 0.91596\ldots$ is the Catalan constant defined in Nahin (2015) p.xv. The proof of $\int_0^{\infty} \frac{\ln(x)}{(x^2+1)} dx = G$ is in Nahin (2015) p.151.

The other way to calculate the expectation requires the determination of the pdf of $Y = \ln|X|$ first. We have $h(y) = e^y$ and the result is

$$f_Y(y) = f_A(h(y))h'(y) = \frac{2e^y}{\pi(e^{2y}+1)}.$$

The expectation is

$$E(Y) = \int_{-\infty}^{\infty} \frac{2ye^y}{\pi(e^{2y}+1)} dy.$$

But it is equal to $2G/\pi$. We have obtained the integral

$$\int_{-\infty}^{\infty} \frac{ye^y}{(e^{2y}+1)} dy = G.$$

Let us find out what the second moment is. One way to do it is this

$$M_2(Y) = \int_0^{\infty} \frac{2(\ln(x))^2}{\pi(x^2+1)} dx = \frac{\pi^2}{4}.$$

The proof of $\int_0^{\infty} \frac{(\ln(x))^2}{(x^2+1)} dx = \pi^3/8$ is in Nahin (2015) p.160.

The second way is the use the pdf of Y, that is,

$$M_2(Y) = \int_{-\infty}^{\infty} \frac{2y^2e^y}{\pi(e^{2y}+1)} dy.$$

Two ways of defining $M_2(Y)$ give the same result which is $\pi^2/4$.

4. Substitution.

We assume that X is a random variable with a pdf f_X positive on some interval I_X and zero otherwise. We define a random variable $Y = g(X)$ taking on values in the interval obtained by mapping the interval I_X to the interval I_Y or $I_Y = g(I_X)$. In our examples we get $I_X = [0, \infty)$ and $I_Y = (-\infty, \infty)$. We assume that $g(x)$ is differentiable and increasing, its inverse is $h(y) = g^{-1}(y)$.

When we express the n-th moment as

$$\int_{I_X} g(x)^n f_X(x) dx$$

and use the substitution $y = g(x), x = h(y), dx = h'(y)dy$, we get

$$\int_{I_Y} y^n f_X(h(y))h'(y)dy.$$

This is exactly what we would get if we calculated the n-th moment with the help of pdf of f_Y.

5. Conclusion.

We have shown that certain integrals that are impossible to calculate with the help of antiderivatives may be applied in probability theory. The constants such as the Euler or Catalan constants may appear in such examples.

It may be difficult to find evaluations of certain non trivial integrals. The book by Nahin (2015) was very helpful, but unfortunatly, the regular editor and its search function did not make the task easy.

6. References.

Nahin, P.J., Inside Interesting Integrals, (2015), Springer