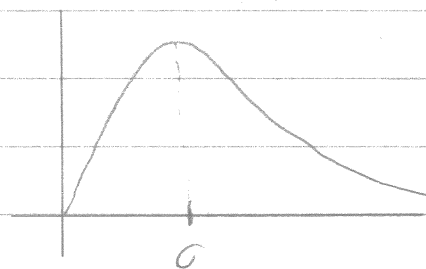


Rayleigh Peak values (envelope) of Gaussian random voltage or current are Rayleigh distributed. Peak values of the sum of sine waves with different frequencies. Error associated with aiming: suppose the origin of a rectangular coordinate system is the aim and the errors along the axes is $X \sim N(0, \sigma^2)$ and $Y \sim N(0, \sigma^2)$ respectively, then the miss distance is $R = \sqrt{X^2 + Y^2}$ and $R \sim \text{Rayleigh}(\sigma)$

$$f(r) = \frac{r}{\sigma^2} \exp\left(-\frac{r^2}{2\sigma^2}\right), \quad r \geq 0$$



$$E[R] = \int_0^{\infty} r f(r) dr = \int_0^{\infty} \frac{r^2}{\sigma^2} \exp\left(-\frac{r^2}{2\sigma^2}\right) dr$$

$$= \left| x = \frac{r}{\sigma} \quad dx = \frac{dr}{\sigma} \right| = \sigma \int_0^{\infty} x^2 e^{-\frac{x^2}{2}} dx = \sigma \sqrt{2\pi} \int_0^{\infty} \frac{1}{\sqrt{2\pi}} x^2 e^{-\frac{x^2}{2}} dx = \sigma \sqrt{2\pi} \frac{E[Z^2]}{2}$$

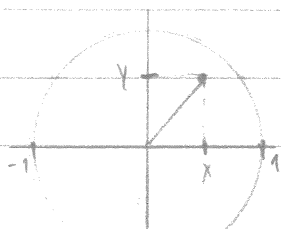
$$\text{Now } \text{var}(Z) = 1 \Leftrightarrow E[Z^2] - E[Z]^2 = 1 \Leftrightarrow E[Z^2] = 1 \Rightarrow$$

$$E[R] = \sigma \sqrt{2\pi} \frac{1}{2} = \sqrt{\frac{\pi}{2}} \sigma$$

$$\text{Var}(R) = \left(2 - \frac{\pi}{2}\right) \sigma^2$$

$$F(r) = \int_0^r \frac{u}{\sigma^2} \exp\left(-\frac{u^2}{2\sigma^2}\right) du = 1 - \exp\left(-\frac{r^2}{2\sigma^2}\right), \quad r \geq 0$$

Example.



$$X \sim N(0, \frac{1}{4}) \quad Y \sim N(0, \frac{1}{4})$$

$$\Rightarrow R \sim \text{Rayleigh}\left(\frac{1}{4}\right) \Rightarrow f(r) = 16r \exp(-8r^2)$$

The mean value of the miss distance is thus

$$E[R] = \sqrt{\frac{\pi}{2}} \cdot \frac{1}{4} = 0.313$$

The standard deviation is

$$\sqrt{\text{Var}(R)} = \sqrt{\left(2 - \frac{\pi}{2}\right) \sigma^2} = \sqrt{0.429 \cdot \frac{1}{4}} = 0.164$$

The probability of missing completely is

$$P_r(R > 1) = 1 - P_r(R \leq 1) = 1 - F(1) = 1 - \left[1 - \exp\left(-\frac{1^2}{2(0.25)^2}\right) \right] = e^{-8} = 3.35 \cdot 10^{-4}$$

If the bulls-eye is 2 in in diameter (1 ft = 12 in) then

$$P_r(\text{bulls-eye}) = P_r(R \leq \frac{1}{12}) = F\left(\frac{1}{12}\right) = 1 - \exp\left(-\frac{8}{144}\right) = 0.0540$$

← check again example in Expected values section

Maxwell Suppose we want to determine the probability that a given gas molecule will have more than twice the mean kinetic energy for all molecules

The kinetic energy is given by $e = \frac{1}{2}mv^2 \Rightarrow v = \sqrt{\frac{2e}{m}}$. The mean kinetic energy is $\frac{3}{2}m\sigma^2$, then the velocity of a molecule with twice the mean kinetic energy is $\sqrt{2 \cdot \left(\frac{3}{2}m\sigma^2\right)} / m = \sqrt{6}\sigma$

$$P_r(V > \sqrt{6}\sigma) = \int_{\sqrt{6}\sigma/\sqrt{\pi}}^{\infty} \frac{2}{\sigma^2} \frac{v^2}{\pi} \exp\left(-\frac{v^2}{2\sigma^2}\right) dv = 0.1126$$

↑ numerically

Chi-square Suppose $X \sim N(0,1) \Rightarrow Y = X^2 \sim \chi^2_{(1)}$

$$f(y) = \frac{y^{-\frac{1}{2}}}{\sqrt{2\pi}} \exp\left(-\frac{y}{2}\right), \quad y > 0$$

The power W dissipated in a resistor is given by $W = RI^2$. If the current I is assumed to be Gaussian with mean zero and variance $\sigma^2 \Rightarrow RI \sim N(0, R\sigma^2)$
 $\Rightarrow W \sim \chi^2_{(1)}$ (non-centered)

$$f(w) = \frac{1}{\sqrt{2\pi R w \sigma^2}} \exp\left(-\frac{w}{2R\sigma^2}\right) \quad w > 0$$

In general, if $X_1, X_2, \dots, X_n$ are independent Gaussian random variables with mean 0 and variance 1 then $Y = \sum_{i=1}^n X_i^2 \sim \chi^2(n)$

$$f(y) = \frac{y^{\frac{n}{2}-1}}{2^{\frac{n}{2}} \Gamma(\frac{n}{2})} \exp\left(-\frac{y}{2}\right), \quad y \geq 0$$

$$E[Y] = n \quad \text{and} \quad \text{Var}(Y) = 2n$$

This distribution arises statistics problems like hypothesis testing (for example, to determine if a sampled voltage is just noise or contains a signal)

Example. Back to the distribution of the power in a resistor

$$\begin{aligned} P_r(W \leq w) &= P_r(RI^2 \leq w) = P_r(I^2 \leq \frac{w}{R}) = P_r(|I| \leq \sqrt{\frac{w}{R}}) \\ &= P_r\left(-\sqrt{\frac{w}{R}} \leq I \leq \sqrt{\frac{w}{R}}\right) = P_r\left(\frac{-\sqrt{w/R}}{\sigma} \leq \frac{I}{\sigma} \leq \frac{\sqrt{w/R}}{\sigma}\right) = \Phi\left(\frac{\sqrt{w/R}}{\sigma}\right) - \Phi\left(\frac{-\sqrt{w/R}}{\sigma}\right) \\ &= \Phi\left(\frac{\sqrt{w/R}}{\sigma}\right) - \left[1 - \Phi\left(\frac{\sqrt{w/R}}{\sigma}\right)\right] = 2\Phi\left(\frac{\sqrt{w/R}}{\sigma}\right) - 1 \end{aligned}$$

Consider a speaker with resistance of 4 Ω rated for a max power of 25 W

If we assume a Gaussian input current with mean 4 W, what's the probability that the maximum power level of the speaker will be exceeded? Since 4 W dissipated in 4 Ω implies $\sigma^2 = 1$ it follows that:

$$\begin{aligned} P_r(W > 25) &= 1 - P_r(W \leq 25) = 1 - \left[2\Phi\left(\frac{\sqrt{25/4}}{1}\right) - 1\right] = 2 - 2\Phi(2.5) = 2 - 2(0.9938) \\ &= 0.0124 \end{aligned}$$

Log-normal Let $Y \sim N(\mu, \sigma^2)$ and $X \sim e^Y \Rightarrow X \sim \text{LogNormal}(\mu, \sigma^2)$

In communication systems the attenuation A of a signal power is computed as $A = \ln\left(\frac{W_{\text{out}}}{W_{\text{in}}}\right)$ where W_{out} and W_{in} are the output and input

signal powers respectively. A is very often quite close to a Gaussian random variable $\Rightarrow X = \frac{W_{\text{out}}}{W_{\text{in}}} \sim \text{Log-Normal}(\mu, \sigma^2)$

$$f(x) = \frac{1}{\sqrt{2\pi\sigma}x} \exp\left\{-\frac{(\ln(x)-\mu)^2}{2\sigma^2}\right\}, \quad x > 0.$$

$$E[X] = \exp\left\{\mu + \frac{1}{2}\sigma^2\right\} \quad \text{Var}(X) = (e^{\sigma^2} - 1)e^{2\mu + \sigma^2}$$

Functions of a random variable

Suppose that a random variable X has density function $f_X(x)$. We often need to find the density function of a new random variable defined as $Y = g(X)$. If g is differentiable, strictly monotonic in some interval I and $f_X(x) = 0$ if $x \notin I$ then

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{d}{dy} g^{-1}(y) \right|$$

$$\text{Proof. } F_Y(y) = P_r(Y \leq y) = P_r(g(X) \leq y) \stackrel{g \text{ monotonically increasing}}{\downarrow} P_r(X \leq g^{-1}(y)) = F_X(g^{-1}(y))$$

$$f_Y(y) = \frac{d}{dy} F_Y(y) = \frac{d}{dx} F_X(g^{-1}(y)) = f_X(g^{-1}(y)) \frac{d}{dy} g^{-1}(y)$$

By considering g monotonically decreasing and putting the results together we conclude the proof.

Example $X \sim N(0,1)$, $Y = X^2 \Rightarrow g(x) = x^2 \Rightarrow g^{-1}(y) = \pm\sqrt{y}$, $y \geq 0$

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{d}{dy} g^{-1}(y) \right| = \frac{1}{\sqrt{2\pi}} \left[e^{-\frac{(-\sqrt{y})^2}{2}} + e^{-\frac{(\sqrt{y})^2}{2}} \right] \left| \frac{1}{2\sqrt{y}} \right| = \frac{y^{-\frac{1}{2}}}{\sqrt{2\pi}} e^{-\frac{y}{2}}$$

Alternatively: $F_Y(y) = P(Y \leq y) = P(X^2 \leq y) = P(|X| \leq \sqrt{y}) = P(-\sqrt{y} \leq X \leq \sqrt{y}) = \Phi(\sqrt{y}) - \Phi(-\sqrt{y})$

$$f_Y(y) = \frac{1}{2} y^{-\frac{1}{2}} \phi(\sqrt{y}) + \frac{1}{2} y^{-\frac{1}{2}} \phi(-\sqrt{y}) = y^{-\frac{1}{2}} \phi(\sqrt{y}) = \frac{y^{-\frac{1}{2}}}{\sqrt{2\pi}} e^{-\frac{y}{2}}$$

auk! 