

Integrating like a statistician

Take the constant function

$$g(x) = 1$$

And integrate over $[a, b]$

$$\int_a^b dx = [x]_a^b = (b - a) \neq 1$$

\implies Not a pdf! But $\frac{b-a}{b-a} = 1$, so scale $g(x)$ by $\frac{1}{b-a}$:

$$f(x) = \frac{1}{b-a}g(x) = \frac{1}{b-a}$$

and integrate over $[a, b]$

$$\frac{1}{b-a} \int_a^b dx = \frac{1}{b-a} [x]_a^b = \frac{b-a}{b-a} = 1$$

So $f(x) = \frac{1}{b-a}$ over $[a, b]$ is a pdf.

Scaling variance / expectation value

The property of variance

$$\begin{aligned}\mathbb{V}[aX] &= \mathbb{E}[(aX - \mathbb{E}[aX])(aX - \mathbb{E}[aX])] \\ &= \mathbb{E}[(aX - a\mathbb{E}[X])(aX - a\mathbb{E}[X])] \\ &= \mathbb{E}[a(X - \mathbb{E}[X])a(X - \mathbb{E}[X])] \\ &= \mathbb{E}[a^2(X - \mathbb{E}[X])(X - \mathbb{E}[X])] \\ &= a^2 \mathbb{E}[(X - \mathbb{E}[X])(X - \mathbb{E}[X])] \\ &= a^2 \mathbb{V}[X]\end{aligned}$$

means we can manipulate the parameters defining pdf's of random variables. For example, if $X \sim N(0, \sigma^2)$, then

$$\mathbb{V} X = \mathbb{E}[(X - \mathbb{E} X)^2] = \sigma^2$$

And if we now want the variance to equal 1, we can work backwards to find a

scaling magnitude that achieves that

$$\begin{aligned}
 1 &= \frac{\sigma^2}{\sigma^2} \\
 &= \frac{1}{\sigma^2} \sigma^2 \\
 &= \frac{1}{\sigma^2} \mathbb{E}[(X - \mathbb{E} X)^2] \\
 &= \mathbb{E}\left[\left(\frac{1}{\sigma^2}\right)(X - \mathbb{E} X)^2\right] \\
 &= \mathbb{E}\left[\left(\frac{1}{\sigma}\right)(X - \mathbb{E} X)\left(\frac{1}{\sigma}\right)(X - \mathbb{E} X)\right] \\
 &= \mathbb{E}\left[\left(\frac{1}{\sigma}X - \frac{1}{\sigma}\mathbb{E} X\right)\left(\frac{1}{\sigma}X - \frac{1}{\sigma}\mathbb{E} X\right)\right] \\
 &= \mathbb{E}\left[\left(\frac{1}{\sigma}X - \mathbb{E}\frac{1}{\sigma}X\right)\left(\frac{1}{\sigma}X - \mathbb{E}\frac{1}{\sigma}X\right)\right] \\
 &= \mathbb{V}\left[\frac{1}{\sigma}X\right]
 \end{aligned}$$

So scaling X by $\frac{1}{\sigma}$ makes the resulting random variable to have a variance of one.

Scaling variance with matrix notation

Let $\mathbb{V} \mathbf{x} = \mathbf{\Omega}$. Then

$$\begin{aligned}
 \mathbb{V}[\mathbf{Ax}] &= \mathbb{E}[(\mathbf{Ax} - \mathbb{E}[\mathbf{Ax}])(\mathbf{Ax} - \mathbb{E}[\mathbf{Ax}])'] \\
 &= \mathbb{E}[(\mathbf{Ax} - \mathbf{A}\mathbb{E}[\mathbf{x}])(\mathbf{Ax} - \mathbb{E}[\mathbf{x}]\mathbf{A})'] \\
 &= \mathbb{E}[\mathbf{A}(\mathbf{x} - \mathbb{E}[\mathbf{x}])(\mathbf{x} - \mathbb{E}[\mathbf{x}])'\mathbf{A}'] \\
 &= \mathbf{A}\mathbb{E}[(\mathbf{x} - \mathbb{E}[\mathbf{x}])(\mathbf{x} - \mathbb{E}[\mathbf{x}])']\mathbf{A}' \\
 &= \mathbf{A}\mathbb{V} \mathbf{x} \mathbf{A}' \\
 &= \mathbf{A}\mathbf{\Omega}\mathbf{A}'
 \end{aligned}$$