

## Properties of the expectation operator $E(\cdot)$

Let  $X$  and  $Y$  be random variables — they can be discrete or continuous. The following properties can be verified:

$$E(\alpha) = \alpha, \text{ where } \alpha \text{ is a constant.}$$

$$E(\alpha X) = \alpha E(X)$$

$$E(g(X) + h(X)) = E(g(X)) + E(h(X)), \text{ where}$$

$g(\cdot)$  and  $h(\cdot)$  are  
functions of  $X$ .

$$E(X + Y) = E(X) + E(Y).$$

## Properties of the variance operator $\text{Var}(\cdot)$

$$\text{Var}(\alpha) = 0$$

$$\text{Var}(\alpha X) = \alpha^2 \text{Var} X$$

What about  $\text{Var}(X + Y)$ ?

It turns out that a bit of care is needed.

In the case where  $X$  and  $Y$  are so-called 'independent' random variables (which implies that  $\text{Cov}(X, Y) = 0$ ), then we can write

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y).$$

But this is not the general case. Let us examine that.

$$\begin{aligned}\text{Var}(X+Y) &= E(X+Y - E(X+Y))^2 \\ &= E(X+Y - E(X) - E(Y))^2\end{aligned}$$

Let us expand this, to give 16 terms:

$$\begin{aligned}\text{Var}(X+Y) &= E(\cancel{X^2} + \cancel{XY} - \cancel{XE(X)} - \cancel{XE(Y)} \\ &\quad + \cancel{Y^2} + \cancel{XY} - \cancel{YE(X)} - \cancel{YE(Y)} \\ &\quad - E(X)X - E(X)Y + (E(X))^2 + E(X)E(Y) \\ &\quad - E(Y)X - E(Y)Y + E(Y)E(X) + E(Y)^2) \\ &= E(X^2 + Y^2 + (E(X))^2 + (E(Y))^2 \\ &\quad + 2XY - 2E(X)X - 2E(Y)Y - 2E(X)Y \\ &\quad - 2E(Y)X + 2E(X)E(Y))\end{aligned}$$

(Sorry, that was a bit tedious...)

Now, given that this involves an expectation of a sum, and noting that  $E(X)$  and  $E(Y)$  are constants, which we can write  $\mu_x$  and  $\mu_y$ , then we have:

$$\begin{aligned}\text{Var}(X+Y) &= E(X^2) + E(Y^2) + \mu_x^2 + \mu_y^2 \\ &\quad + 2E(XY) - 2\mu_x E(X) - 2\mu_y E(Y) - 2\mu_x E(Y) \\ &\quad - 2\mu_y E(X) + 2\mu_x \mu_y \\ &= E(X^2) + E(Y^2) + \mu_x^2 + \mu_y^2 \\ &\quad + 2E(XY) - 2\mu_x^2 - 2\mu_y^2 - 2\mu_x \mu_y \\ &\quad - 2\mu_y \mu_x + 2\mu_x \mu_y \\ &= E(X^2) - \mu_x^2 + E(Y^2) - \mu_y^2 \\ &\quad + 2E(XY) - 2\mu_x \mu_y\end{aligned}$$

But we know we can write the variance of  $X$  as:

$$\begin{aligned}\text{Var}(X) &= E(X - E(X))^2 \\ &= E(X - \mu_x)^2 \\ &= E(X^2 - 2\mu_x X + \mu_x^2) \\ &= E(X^2) - 2\mu_x E(X) + \mu_x^2 \\ &= E(X^2) - \mu_x^2\end{aligned}$$

and similarly

$$\text{Var}(Y) = E(Y^2) - \mu_y^2$$

then we have

$$\begin{aligned}\text{Var}(X + Y) &= \text{Var}(X) + \text{Var}(Y) \\ &\quad + 2(E(XY) - \mu_x \mu_y)\end{aligned}$$

What about  $E(XY) - \mu_x \mu_y$  ?

Consider

$$\begin{aligned}\text{Cov}(X, Y) &\equiv E((X - \mu_x)(Y - \mu_y)) \\ &= E(XY - \mu_x Y - \mu_y X + \mu_x \mu_y) \\ &= E(XY) - \mu_x E(Y) - \mu_y E(X) + \mu_x \mu_y \\ &= E(XY) - \mu_x \mu_y - \mu_x \mu_y + \mu_x \mu_y \\ &= E(XY) - \mu_x \mu_y\end{aligned}$$

Therefore, in general, we have

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y).$$