

## COVARIANCE

# Why

1

### Definition

The *covariance* between two random variables which are each integrable and whose product is integrable is the expectation of their product less the product of their expectation.

#### Notation

Let f and g be two integrable random variables with fg integrable. Denote the covariance of f with g by  $\mathbf{cov}(f,g)$ . We defined it:

$$cov(f, g) = E(fg) - E(f) E(g).$$

## **Properties**

Prop. 1. Covariance is symmetric and billinear.<sup>2</sup>

**Prop. 2.** The covariance of a random variable with itself is its variance.

Proof. Let f be a square-integrable real-valued random variable, then

$$\operatorname{cov}(f,f) = \operatorname{E}(ff) - \operatorname{E}(f)\operatorname{E}(f) = \operatorname{E}(f^2) - (\operatorname{E}(f))^2 = \operatorname{var}(f).$$

<sup>1</sup>Future editions will include this.

<sup>&</sup>lt;sup>2</sup>Future editions will include an account.

**Prop.** 3. The variance of a sum of integrable real-valued random variables whose pairwise products are integrable is the double sum of the pairwise covariances.

*Proof.* Let  $f_1, \ldots, f_n$  be integrable random variables with  $f_i f_j$  integrable for all  $i, j = 1, \ldots, n$ . Using the billinearity,

$$\mathsf{var}igg(\sum_{i=1}^n f_iigg) = \mathsf{cov}igg(\sum_{i=1}^n f_i, \sum_{i=1}^n f_iigg)$$
 
$$= \sum_{i=1}^n \sum_{j=1}^n \mathsf{cov}(f_i, f_j)$$

