

# Essentials of Applied Data Analysis

## IPSA-USP Summer School 2017

### Handout - Covariance and Correlation

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### Functions of random variables

We can sum random variables. For example:  $Z = X + Y$

What is the expected value of  $Z$ ?

$$E[Z] = E[X + Y] = E[X] + E[Y]$$

We can multiply random variables. For example:  $Z = X * Y$

What is the expected value of  $Z$ ? It depends. If  $X$  and  $Y$  are independent of each other, then

$$E[Z] = E[X * Y] = E[XY] = E[X] * E[Y]$$

What if they are not independent?

### Covariance

The covariance of a variable is defined as

$$Cov(X, Y) = E[(X - E[X]) * (Y - E[Y])]$$

But wait, this looks familiar!

$$\begin{aligned} Cov(X, X) &= E[(X - E[X]) * (X - E[X])] = \\ &= E[(X - E[X])^2] = E[X^2] - (E[X])^2 = Var[X] \end{aligned}$$

Another way to look at the covariance:

$$Cov(X, Y) = E[XY] - E[X] * E[Y]$$

Wait again! Isn't  $E[XY]$  the expected value of the multiplication of  $X$  and  $Y$ ? Yes it is, so we can rearrange the formula to obtain:

$$E[XY] = E[X] * E[Y] + Cov(X, Y)$$

If  $X$  and  $Y$  are independent of each other, then

$$E[XY] = E[X] * E[Y]$$

Or we can simply say that

$$Cov(X, Y) = 0$$

So the covariance is a measure of how two variables are related to each other. One problem with the covariance is that it can be any number, hence, we can't compare two covariances. But there is a way to "standardize" covariances..

## Correlation

The correlation of two variables is defined as:

$$\rho_{xy} = Corr(X, Y) = \frac{Cov(X, Y)}{\sigma_x * \sigma_y}$$

Where  $\sigma_x$  and  $\sigma_y$  are the standard deviations of  $X$  and  $Y$ .

One important property of the correlation is that, differently from the covariance, it is bounded:

$$-1 \leq Corr(X, Y) \leq 1$$

As we will see in the future, correlation is a measure of linear relation among two random variables.

## Variance of the sum

Finally, when we sum two variables, for example:  $Z = X + Y$ , the variance of the sum is:

$$Var[Z] = Var[X + Y] = Var[X] + Var[Y] + 2 * Cov(X, Y)$$

Since  $Cov(X, Y) = 0$  under indepenence, if  $X$  and  $Y$  are indepent we can simplify the variance to:

$$Var[Z] = Var[X + Y] = Var[X] + Var[Y]$$