

# Multivariate Monte Carlo

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## Problem

Given a  $2 \times 2$  standardized variance-covariance matrix, also known as a correlation matrix,  $R$ , where  $\rho$  is the coefficient of correlation,

$$R = \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix}$$

We start with uncorrelated variates  $x$ . Our job is to transform the uncorrelated variates to produce correlated variates  $z$  with the same expected variance-covariance matrix as  $R$ . This will put informational structure into the otherwise independently occurring  $x$ .

When we say transform we mean this mathematically.

$$z = Lx$$

such that

$$R = LL^T$$

## Solution

We can decompose  $R$  into the product of upper and lower triangular symmetrical matrices. This is a standard trick of matrix algebra.

$$R = \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix} = \begin{bmatrix} \ell_{11} & 0 \\ \ell_{21} & \ell_{22} \end{bmatrix} \begin{bmatrix} \ell_{11} & \ell_{21} \\ 0 & \ell_{22} \end{bmatrix}$$

When we multiply the upper and lower triangular matrices we get this interesting matrix.

$$R = \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix} = \begin{bmatrix} \ell_{11}^2 & \ell_{11}\ell_{21} \\ \ell_{11}\ell_{21} & \ell_{22}^2 \end{bmatrix}$$

Now it is a matter of matching elements of  $R$  and our new matrix. Here we go.

- $\ell_{11}^2 = 1$  implies  $\ell_{11} = 1$
- $\ell_{11}\ell_{21} = \rho$  then implies  $\ell_{21} = \rho$
- $\ell_{11}^2 + \ell_{22}^2 = 1$  then implies that  $\ell_{22} = \sqrt{1 - \rho^2}$

That wasn't as back as we might have thought when we started. We then let  $L$  be the new matched to  $R$  matrix.

$$L = \begin{bmatrix} \ell_{11} & 0 \\ \ell_{21} & \ell_{22} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ \rho & \sqrt{1 - \rho^2} \end{bmatrix}$$

## Now we can simulate

We generate correlated  $z = Lx$  building on the uncorrelated  $x$ . We first generate a random  $x_1$  and, independently, a random  $x_2$ . We can use the `=RAND()` function in Excel to perform this task.

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

Then we generate a  $z_1$  and a  $z_2$  using the  $x$  vector of random numbers, but transformed by pre-multiplying  $x$  with  $L$ .

$$z = \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ \rho & \sqrt{1 - \rho^2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

We do remember that By definition, if  $x_1$  is not correlated with  $x_2$ , then  $\rho_{12} = 0$ . We can check our maths with this calculation.

$$xx^T = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \begin{bmatrix} x_1 & x_2 \end{bmatrix} = \begin{bmatrix} x_1^2 & x_1x_2 \\ x_1x_2 & x_2^2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

The multiplication of a column vector of independently drawn  $x$  with its transpose, the row vector of those same  $x$  random numbers will always return the identity matrix  $I$ . This shows that variates are perfectly correlated with themselves but not each other.

Back to the main business at hand. We now calculate

$$zz^T = (Lx)(Lx)^T = Lxx^TL^T = LIL^T = LL^T$$

But,  $R = LL^T$  so that  $R = zz^T$ .

Thus we have sketched out these steps.

- Generate uncorrelated  $x$ .
- Generate  $z = Lx$ , where  $L$  reflects the desired correlation structure.
- Find that  $L$  generates the same correlation structure as the correlations in  $z$ .