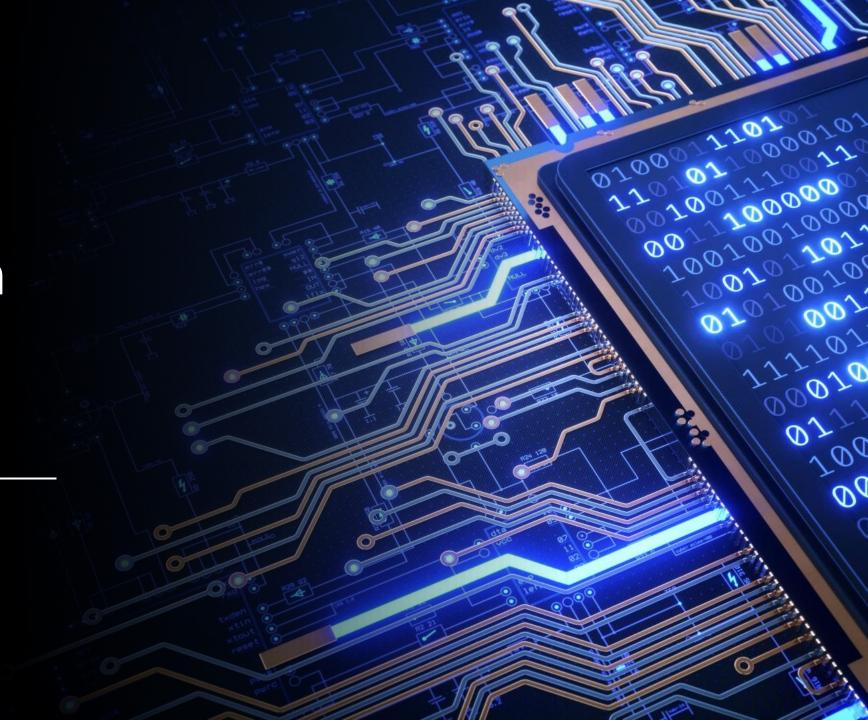
Simulations in Multivariate Setup

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Fixed Cost	Cost Per Unit	Average Sales Per Month	Average Price Per Unit	Profit
₹ 10,000	₹ 200	100	₹ 320	₹ 2,000

What is the probability that you have some profit?

Fixed Cost	Cost Per Unit	Average Sales Per Month	Average Price Per Unit	Profit
₹ 10,000	₹ 200	100	₹ 320	₹ 2,000
		Normal(100,10)		Normal(2000,10)

What is the probability that you have some profit?

Fixed Cost	Cost Per Unit	Average Sales Per Month	Average Price Per Unit	Profit
₹ 10,000	₹ 200	100	₹ 320	₹ 2,000
		Normal(100,10)	Normal(320,5)	Simulation

What is the probability that you have some profit?

Fixed Cost	Cost Per Unit	Average Sales Per Month	Average Price Per Unit	Profit
₹ 10,000	₹ 200	100	₹ 320	₹ 2,000
		Normal(100,10)	Normal(320,5)	??

Price	Sales	
317.92	103.78	
326.91	105.54	
326.78	102.76	
320.13	105.57	

What is the probability that you have some profit?

Fixed Cost	Cost Per Unit	Average Sales Per Month	Average Price Per Unit	Profit
₹ 10,000	₹ 200	100	₹ 320	₹ 2,000
		Normal(100,10)	Normal(320,5)	??
		Correlation = -0.7		, ,

What is the probability that you have some profit?

Fixed Cost	Cost Per Unit	Sales Per Month	Average Price Per Unit	Profit
₹ 10,000	₹ 200	100	₹ 320	₹ 2,000
		Hypergeometric	Normal(320,5)	??
		Correlation = -0.7		; ;

What is the probability that you have some profit?





Copula

The word copula is derived from the Latin noun for a link or a tie (as is the English word "couple")

Its purpose is to describe the dependence structure between two variables.

Sklar's theorem states that "Any multivariate joint distribution can be written in terms of univariate marginal distribution functions and a copula which describes the dependence structure between the two variables".

How Copula Works?

Let:

$$X_1$$
 and X_2 be RVs
 $U_1 = F(X_1)$; $U_2 = F(X_2)$

Then:

 $U_1 \sim uniform$ $U_2 \sim uniform$ How Copula Works?

Let:
$$U_1 = F(X_1); \ U_2 = F(X_2)$$

$$F(X_1, X_2) = P(X_1 \le x_1 \cap X_2 \le x_2)$$

$$= P(F^{-1}(U_1) \le x_1 \cap F^{-1}(U_2) \le x_2)$$

$$= P(U_1 \le F(x_1) \cap U_2 \le F(x_2))$$

$$= P(U_1 \le u_1 \cap U_2 \le u_2) = C(U_1, U_2)$$

Gaussian Copula - Simulation

For the normal copula, the input of the simulation is the correlation matrix Σ . The normal copula can be simulated by the following steps, in which $U=(U_1,\ldots,U_m)$ denotes one random draw from the copula:

- 1. Generate a multivariate normal vector $m{Z} \sim N(0, \Sigma)$, where Σ is an m-dimensional correlation matrix.
- 2. Transform the vector Z into $U = (\Phi(Z_1), \dots, \Phi(Z_m))^T$, where Φ is the distribution function of a univariate standard normal.

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Let $\Theta = \{(\nu, \Sigma) : \nu \in (1, \infty), \Sigma \in \mathbb{R}^{m \times m}\}$, and let t_{ν} be a univariate t distribution with ν degrees of freedom.

The Student's t copula can be written as

$$C_{\Theta}(u_1,u_2,\ldots,u_m) = oldsymbol{t}_{
u,\Sigma} \Big(t_{
u}^{-1}(u_1), t_{
u}^{-1}(u_2),\ldots,t_{
u}^{-1}(u_m) \Big)$$

where $t_{\nu,\Sigma}$ is the multivariate Student's t distribution that has a correlation matrix Σ with ν degrees of freedom.

The Student t Copula

The input parameters for the simulation are (ν, Σ) . The t copula can be simulated by the following steps:

- 1. Generate a multivariate vector $m{X} \sim t_m(\nu,0,\Sigma)$ that follows the centered t distribution with ν degrees of freedom and correlation matrix Σ .
- 2. Transform the vector \boldsymbol{X} into $\boldsymbol{U}=(t_{\nu}(X_1),\ldots,t_{\nu}(X_m))^T$, where t_{ν} is the distribution function of a univariate t distribution with ν degrees of freedom.

Gaussian Copula Vs t -Copula

