

Copula

Suppose (X, Y) is a random vector. There are many situations where X and Y will be dependent on each other.

Ex 1 $(X, Y) = (\text{Oil price}, \text{Car price})$

Ex 2 $(X, Y) = (\text{Oil price}, \text{Food price})$

A copula can be used to model the dependence among 2 or more variables.

For 2 variables, a copula is a function

$$C : \boxed{[0, 1] \times [0, 1]} \rightarrow \boxed{[0, 1]}$$

Unit square Unit interval

with

$$\left. \begin{aligned} C(u, 1) &= u \\ C(1, v) &= v \end{aligned} \right\} \begin{array}{l} \text{marginal} \\ \text{CDFs of} \\ C \text{ are} \\ \text{Uniform}[0, 1] \end{array}$$

Suppose (X, Y) has joint CDF $F_{X, Y}(x, y)$ and marginal CDFs $F_X(x), F_Y(y)$.

Statement 1 For every $F_{X, Y}$ there is a corresponding copula.

Proof : We have

$$\begin{aligned} & F_{X, Y}(x, y) \\ &= P(X \leq x, Y \leq y) \\ &= P(F_X(X) \leq F_X(x), F_Y(Y) \leq F_Y(y)) \end{aligned}$$

$$\begin{aligned} & \stackrel{\textcircled{=}}{=} P(\text{Uni}[0, 1] \leq F_X(x), \text{Uni}[0, 1] \leq F_Y(y)) \\ & \quad \swarrow \left[\text{Probability Integral Transform} \right] \end{aligned}$$

$$= C_1(F_X(x), F_Y(y))$$

The proof is complete.

Statement 2 For every G there is a corresponding $F_{X, Y}$.

Proof We have

$$\begin{aligned} G(u, v) &= P(\text{Uni}[0, 1] \leq u, \text{Uni}[0, 1] \leq v) \\ &= P(F_X^{-1}(\text{Uni}[0, 1]) \leq F_X^{-1}(u), \\ &\quad F_Y^{-1}(\text{Uni}[0, 1]) \leq F_Y^{-1}(v)) \end{aligned}$$

$$\Leftrightarrow P(X \leq F_X^{-1}(u), Y \leq F_Y^{-1}(v))$$

Probability Integral Transform

$$= F_{X, Y}(F_X^{-1}(u), F_Y^{-1}(v)).$$

The proof is complete.

Two simple copulas

1) Independence copula is

$$C(u, v) = uv$$

$$\Leftrightarrow P(\text{Uni}[0, 1] \leq u, \text{Uni}[0, 1] \leq v) = uv$$

$$\Leftrightarrow P(F_X^{-1}(\text{Uni}[0, 1]) \leq F_X^{-1}(u), \\ F_Y^{-1}(\text{Uni}[0, 1]) \leq F_Y^{-1}(v)) = uv$$

$$\Leftrightarrow P(X \leq F_X^{-1}(u), Y \leq F_Y^{-1}(v)) = uv$$

Probability Integral Transform

$$\text{Set } x = F_X^{-1}(u) \Rightarrow u = F_X(x)$$

$$" \quad y = F_Y^{-1}(v) \Rightarrow v = F_Y(y)$$

$$\Leftrightarrow P(X \leq x, Y \leq y) = F_X(x) F_Y(y)$$

$$\Leftrightarrow F_{X, Y}(x, y) = F_X(x) F_Y(y)$$

usual definition of independence between X and Y .

2) Completely dependent copula is

$$C^1(u, v) = \min(u, v)$$

$$\Leftrightarrow P(\text{Uni}[0, 1] \leq u, \text{Uni}[0, 1] \leq v) = \min(u, v)$$

$$\Leftrightarrow P(F_X^{-1}(\text{Uni}[0, 1]) \leq F_X^{-1}(u), F_Y^{-1}(\text{Uni}[0, 1]) \leq F_Y^{-1}(v)) = \min(u, v)$$

$$\Leftrightarrow P(X \leq F_X^{-1}(u), Y \leq F_Y^{-1}(v)) = \min(u, v)$$

Probability Integral Transform

$$\text{Set } x = F_X^{-1}(u) \Rightarrow u = F_X(x)$$

$$" \quad y = F_Y^{-1}(v) \Rightarrow v = F_Y(y)$$

$$\Leftrightarrow P(X \leq x, Y \leq y) = \min(F_X(x), F_Y(y))$$

$$\Leftrightarrow F_{X, Y}(x, y) = \min(F_X(x), F_Y(y))$$

usual definition of complete dependence between X and Y .