

# Introduction about Copulas.

**Duc T. Nguyen**

**Ulm University**

November 29, 2020

The talk is based on two following references:

① **An Introduction to Copulas.**

(2006) Roger B. Nelsen, Springer.

② **Elements of Copula Modeling with R.**

(2018) M. Hofert, I.Kojadinovic, M. Machler, J. Yan, Springer.

- First part: **Motivating Example.**
- Second part: **Analysis Approach to Copula.**
  - Subcopula and Copula
  - Sklar's Theorem.
- Third part: **Copulas and Random variables.**
  - Sklar's Theorem.
  - Fréchet-Hoeffding Bounds
- Forth part: **Survival Copula and Symmetric Copula.**

## PART 1: Motivating Example.

# The data set.

► Let  $(\mathbf{X}_1, \mathbf{X}_2)^{(n)} = (\mathbf{X}_{i_1}, \mathbf{X}_{i_2})_{i=1}^n$ ,  $(\mathbf{Y}_1, \mathbf{Y}_2)^{(n)} = (\mathbf{Y}_{i_1}, \mathbf{Y}_{i_2})_{i=1}^n$  be two given data set.

**Goal:** Comparing the dependence between  $\mathbf{X}_1$  and  $\mathbf{X}_2$  with  $\mathbf{Y}_1$  and  $\mathbf{Y}_2$ .

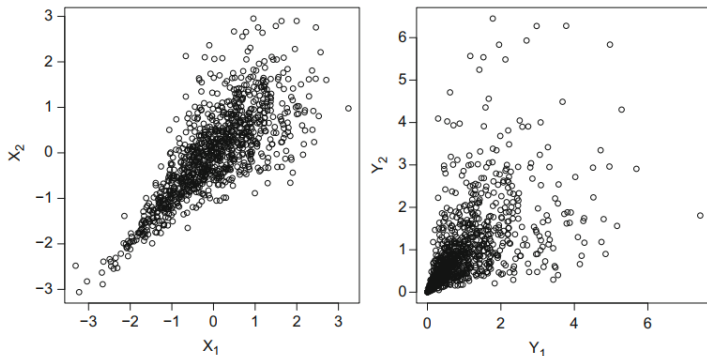


Figure:  $n = 1000$  independent observations of  $(\mathbf{X}_1, \mathbf{X}_2)$  and  $(\mathbf{Y}_1, \mathbf{Y}_2)$ .

# Kernel Estimation.

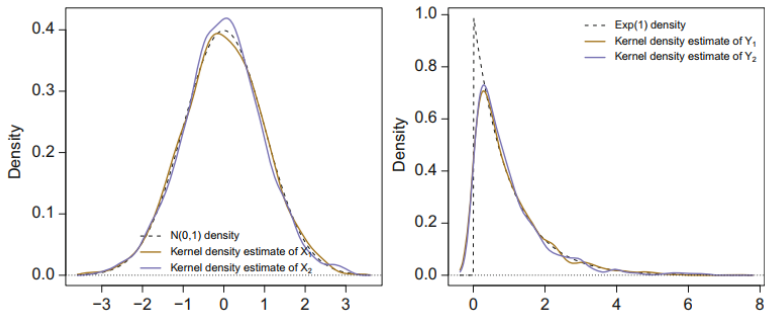


Figure: Kernel density estimates of the densities of  $(X_1, X_2)$  and  $(Y_1, Y_2)$ .

- $X_1$  and  $X_2$  are likely to follow **standard normal distribution**.
- $Y_1$  and  $Y_2$  are likely to follow **standard exponential distribution**.

# Transform to the uniform distributions.

**Lemma:** Let  $X$  be continuous random variable with distribution function  $F$ , then  $F(X)$  is a standard uniform random variable, i.e.,  $F(X) \sim U(0, 1)$

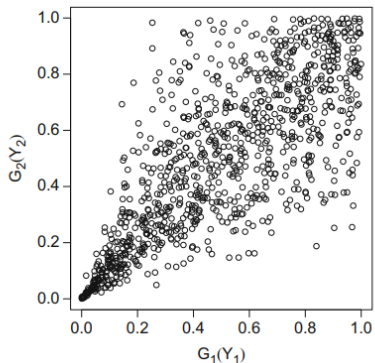
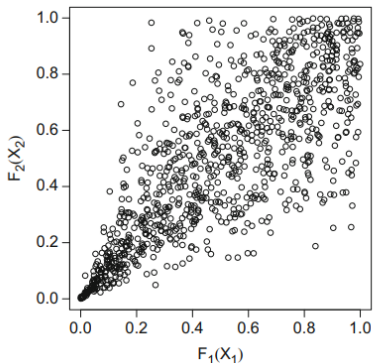


Figure: Scatter plots of  $(F(X_1), F(X_2))$  and  $(G(Y_1), G(Y_2))$

# Comments.

- ▶ The distribution of  $(F(X_1), F(X_2))$  and  $(G(Y_1), G(Y_2))$  seem to be identical.
- ▶ Two given data sets is indistinguishable in terms of **dependence** and only differ in terms of the underlying **marginal distribution function**.

**Question:** Is there any tool to measure the **dependence** between random variables?



# Structure of the talk

**Analysis approach to Copula.**



**Sklar's Theorem.**

+

**Fréchet-Hoeffding Bounds.**



**Copula and Random Variable.**

## PART 2: Analysis approach to Copula.

# Preliminaries.

- ▶ A rectangle  $K$  in  $\overline{\mathbb{R}}^2$  is denoted by:

$$K = [x_1, x_2] \times [y_1, y_2].$$

- ▶ For any function  $H : \overline{\mathbb{R}}^2 \rightarrow \mathbb{R}$ ,  $DomH$  is its domain and  $RanH$  is its range. Assume that  $DomH = S_1 \times S_2$ .
- ▶ Denote  $a_1, a_2$  be the least elements of  $S_1, S_2$  respectively and  $b_1, b_2$  be the greatest elements of  $S_1, S_2$  respectively
- ▶  $H$  – volume of  $K$ :

$$V_H(K) = H(x_2, y_2) - H(x_2, y_1) - H(x_1, y_2) + H(x_1, y_1)$$

- $H$  deduces a measure on  $\overline{\mathbb{R}}^2$ .

## 2 - increasing function.

- ▶ A 2 - place real-valued function  $H$  is **2 - increasing** function if  $V_H(K) \geq 0$  for all rectangles within  $DomH$ .
- ▶ Let  $a_1, a_2$  be the least elements of  $S_1, S_2$  respectively. The function  $H : S_1 \times S_2 \rightarrow \mathbb{R}$  is **grounded** if

$$H(x, a_2) = 0 = H(a_1, y) \text{ for all } (x, y) \in S_1 \times S_2.$$

### Remark:

$$\begin{aligned} 2 - \text{increasing} &\rightarrow \text{grounded} : H(x, y) = (2x - 1)(2y - 1) \\ \text{grounded} &\rightarrow 2 - \text{increasing} : H(x, y) = \max(x, y) \end{aligned}$$

**Lemma:** If  $H$  be a **grounded, 2 - increasing** function  $H$  is nondecreasing in each argument.

# Subcopula.

- ▶ A two - dimensional real-valued subcopula  $\mathbf{C}'$  is a function which has the following properties:
  - $\mathbf{DomC}'$  is an subset of square unit  $\mathbb{I}^2$  containing the points 0 and 1.
  - $\mathbf{C}'$  is **grounded** and **2 - increasing**.
  - For every  $(\mathbf{u}, \mathbf{v})$  in  $\mathbf{S}_1 \times \mathbf{S}_2$ ,  $\mathbf{C}'(\mathbf{u}, \mathbf{1}) = \mathbf{u}$ ,  $\mathbf{C}'(\mathbf{1}, \mathbf{v}) = \mathbf{v}$ .
  
- ▶ A two - dimensional real-valued **copula** is a **subcopula** whose domain is  $\mathbb{I}^2$ .

# Example of copula.

- ▶ Considering the **Frank** Copula

$$C^F(u, v) = \frac{1}{9} \log \left( 1 + \frac{(e^{9u} - 1)(e^{9v} - 1)}{e^9 - 1} \right), (u, v) \in [0, 1]^2$$

One can check it satisfies three conditions to be a copula.

# Important Lemmas

**Lemma:** Let  $C'$  be a subcopula. Then for every  $(u_1, v_1), (u_2, v_2)$  in  $\text{Dom}C'$ ,

$$|C'(u_2, v_2) - C'(u_1, v_1)| \leq |u_2 - u_1| + |v_2 - v_1|.$$

Hence  $C$  is **uniformly continuous** on its domain.

**Lemma:** Let  $C'$  be a subcopula. Then for any  $(u, v)$  in  $\text{Dom}C'$ ,

$$\max(u + v - 1, 0) \leq C'(u, v) \leq \min(u, v).$$

**Lemma:** Let  $C$  be a copula. For any  $v$  in  $\mathbb{I}$ , the partial derivative  $\partial C(u, v)/\partial u$  exist for almost all  $u$ , and for such  $v$  and  $u$ ,

$$0 \leq \frac{\partial}{\partial u} C(u, v) \leq 1.$$

# Distribution Function.

- ▶ A distribution function is a function  $F$  with domain  $\overline{\mathbb{R}}$  such that
  - $F$  is nondecreasing
  - $F(-\infty) = 0$  and  $F(\infty) = 1$
- ▶ A joint distribution function is a function  $H$  whose domain  $\overline{\mathbb{R}}^2$  such that
  - $H$  is 2 - increasing
  - $H(x, -\infty) = 0 = H(-\infty, y)$  and  $H(\infty, \infty) = 1$ .
- ▶ The margins  $F$  and  $G$  of  $H$  are given by and .
  - $F(x) = H(x, \infty)$
  - $G(y) = H(\infty, y)$



## Sklar's Theorem

Let  $H$  be a joint distribution function with margins  $F$  and  $G$ . Then there exists a **copula**  $C$  such that for all  $x, y$  in  $\overline{\mathbb{R}}$ ,

$$H(x, y) = C(F(x), G(y)). \quad (1)$$

If  $F$  and  $G$  are continuous, then  $C$  is **unique**; otherwise,  $C$  is **uniquely determined** on  $\text{Ran}F \times \text{Ran}G$ .

Conversely, if  $C$  is a copula and  $F, G$  are distributions function, then the function  $H$  defined by (1) is a joint distribution function with margins  $F$  and  $G$ .

# Procedure of the proof.

There exists a unique subcopula  $C'$ ;  $H(x, y) = C'(F(x), G(y))$  .

+

There exists a copula  $C$ , it is coincide with  $C'$  over the  $DomC'$  .

+

If  $F$  and  $G$  are continuous, then  $RanF = RanG = \mathbb{I} \rightarrow C \equiv C'$  .

↓

**Theorem is proved.**

## Lemma 1

Let  $H$  be the joint distribution function with margins  $F$  and  $G$ . Then there exists a unique subcopula  $C'$  such that

- $DomC' = RanF \times RanG$
- For all  $x, y$  in  $\overline{\mathbb{R}}$ ,  $H(x, y) = C'(F(x), G(y))$ .

**Proof:** For any  $(x_1, y_1), (x_2, y_2) \in \overline{\mathbb{R}}^2$ , one has

$$|H(x_2, y_2) - H(x_1, y_1)| \leq |F(x_2) - F(x_1)| + |G(y_2) - G(y_1)|.$$

$H(x_1, y_1) = H(x_2, y_2)$  when  $(F(x_1), G(y_1)) = (F(x_2), G(y_2))$ .

Define the following mapping:

$$C' : RanF \times RanG \rightarrow [0, 1]$$

where  $C'(F(x), G(y)) := H(x, y)$ .

Hence, it is unique on  $DomC' = RanF \times RanG$ .

## Lemma 2

Let  $\mathbf{C}'$  be a subcopula. Then there exists a copula  $\mathbf{C}$  such that it coincides with  $\mathbf{C}'$  over the  $\text{Dom}\mathbf{C}'$  and this extension is non-unique.

**Proof:** Because of the continuity of the subcopula  $\mathbf{C}'$ , it will be a subcopula  $\mathbf{C}''$  over  $\overline{\mathbf{S}_1} \times \overline{\mathbf{S}_2}$ .

For any point  $(\mathbf{a}, \mathbf{b}) \in \mathbb{I}^2$ , denote by

- $\mathbf{a}_1, \mathbf{b}_1$  be the greatest element in  $\mathbf{S}_1, \mathbf{S}_2$  respectively such that

$$\mathbf{a}_1 \leq \mathbf{a}, \mathbf{b}_1 \leq \mathbf{b}$$

- $\mathbf{a}_2, \mathbf{b}_2$  be the least element in  $\mathbf{S}_1, \mathbf{S}_2$  respectively such that

$$\mathbf{a}_1 \geq \mathbf{a}, \mathbf{b}_1 \geq \mathbf{b}$$

# Proof.

$$\lambda_1 = \begin{cases} (a - a_1)/(a_2 - a_1), & \text{if } a_1 < a_2 \\ 1, & \text{if } a_1 = a_2 \end{cases}$$

$$\mu_1 = \begin{cases} (b - b_1)/(b_2 - b_1), & \text{if } b_1 < b_2 \\ 1, & \text{if } b_1 = b_2 \end{cases}$$

$$C(a, b) = (1 - \lambda_1)(1 - \mu_1)C''(a_1, b_1) + (1 - \lambda_1)\mu_1C''(a_1, b_2) \\ + \lambda_1(1 - \mu_1)C''(a_2, b_1) + \lambda_1\mu_1C''(a_2, b_2).$$

One can check  $C(a, b)$  is **grounded** and **2 - increasing**.

Creating a rectangle  $B$  in this unit square requires one more point  $(c, d) \in \mathbb{I}^2$  with  $c \geq a, d \geq b$ .

Let  $c_1, d_1, c_2, d_2, \lambda_2, \mu_2$  in the same way with  $a$  and  $b$ .

# Proof.

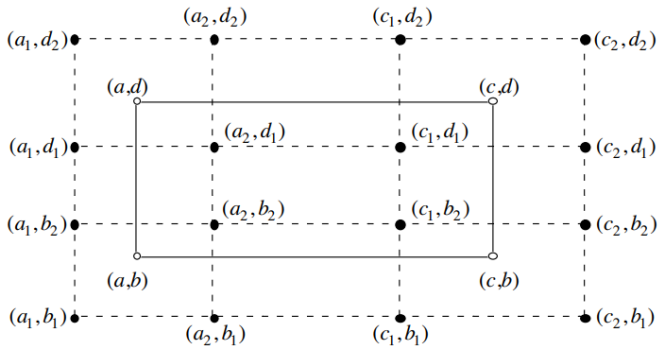


Figure: Scheme of points.

The  $\mathbf{C}$  - volume of  $\mathbf{B}$  can be decomposed into the summation of nine  $\mathbf{C}$  - volumes of nine sub-rectangles, which are all non-negative.

## PART III. Copulas and Random variables.

# Some results.

**Lemma:** Let  $\mathbf{X}$  be continuous random variable with distribution function  $\mathbf{F}$ , then  $\mathbf{F}(\mathbf{X})$  is a standard uniform random variable, i.e.,  $\mathbf{F}(\mathbf{X}) \sim U(0, 1)$ .

► For any random variable  $\mathbf{X}$  with distribution function  $\mathbf{F}$ , the *quantile function*  $\mathbf{F}^{\leftarrow}$  is defined by

$$\mathbf{F}^{\leftarrow}(y) = \inf\{x \in \mathbb{R} : \mathbf{F}(x) \geq y\}, \quad y \in [0, 1],$$

with the convention that  $\inf \emptyset = \infty$ .

**Lemma:** Let  $\mathbf{X} \sim U(0, 1)$  and  $\mathbf{F}$  be any distribution function. Then  $\mathbf{F}^{\leftarrow}(\mathbf{X})$  is a random variable with distribution function  $\mathbf{F}$ .



# Sklar's Theorem.

## Sklar's Theorem

Let  $\mathbf{X}$  and  $\mathbf{Y}$  be random variables with distribution functions  $F$  and  $G$ , respectively, and joint distribution function  $H$ . Then there exists a copula  $C$  satisfying. Then there exists a **copula**  $C$  such that for all  $\mathbf{x}, \mathbf{y}$  in  $\overline{\mathbb{R}}$ ,

$$H(\mathbf{x}, \mathbf{y}) = C(F(\mathbf{x}), G(\mathbf{y})). \quad (2)$$

If  $F$  and  $G$  are continuous, then  $C$  is **unique**; otherwise,  $C$  is **uniquely determined** on  $\text{Ran}F \times \text{Ran}G$ .

- ▶ The copula in this theorem will be called the **copula of  $\mathbf{X}$  and  $\mathbf{Y}$** , denoted  $C_{\mathbf{X}\mathbf{Y}}$

## Remark.

- ▶ Copula is a multivariate distribution function with standard uniform univariate margins, that is,  $\mathbf{U}(\mathbf{0}, \mathbf{1})$  margins.
- ▶ For any  $(\mathbf{u}, \mathbf{v}) \in \text{RanF} \times \text{RanG}$ , (2) can be rewritten as follows

$$H(F^{\leftarrow}(\mathbf{u}), G^{\leftarrow}(\mathbf{v})) = C(\mathbf{u}, \mathbf{v}).$$

- ▶ A random vector has a continuous joint distribution function if and only if it has continuous margins.
- ▶ If one can estimate the copula and the margin from a given data, then the multivariate is provided.

## Example.

Let  $\mathbf{X}, \mathbf{Y}$  be random variables with joint distribution  $\mathbf{H}$

$$H(x, y) = \begin{cases} \frac{(x+1)(e^y - 1)}{x + 2e^y - 1}, & (x, y) \in [-1, 1] \times [0, \infty] \\ 1 - e^{-y}, & (x, y) \in (1, \infty] \times [0, \infty] \\ 0, & \text{else where.} \end{cases}$$

with margins  $\mathbf{F}, \mathbf{G}$  given by

$$F(x) = \begin{cases} 0, & x < -1 \\ (x+1)/2, & x \in [-1, 1] \\ 1, & x > 1 \end{cases}, \quad G(y) = \begin{cases} 0, & y < 0 \\ 1 - e^{-y}, & y \geq 0 \end{cases}$$

## Example.

Obviously  $F, G$  are continuous functions,  $\text{Ran}F = \text{Ran}G = \mathbb{I}$  and the quantile functions of  $F, G$  are

$$F^{\leftarrow}(x) = 2x - 1, \quad G^{\leftarrow}(y) = -\ln(1 - y).$$

Because  $\text{Ran}F = \text{Ran}G = \mathbb{I}$ , one can find the copula

$$C(u, v) = \frac{uv}{u + v - uv}.$$

Hence,

$$H(x, y) = \frac{F(x)G(y)}{F(x) + G(y) - F(x)G(y)}$$

# Independent Copula

## Theorem

Let  $\mathbf{X}$  and  $\mathbf{Y}$  be continuous random variable. Then  $\mathbf{X}$  and  $\mathbf{Y}$  are independent if and only if  $\mathbf{C}_{\mathbf{XY}} = \mathbf{\Pi}$

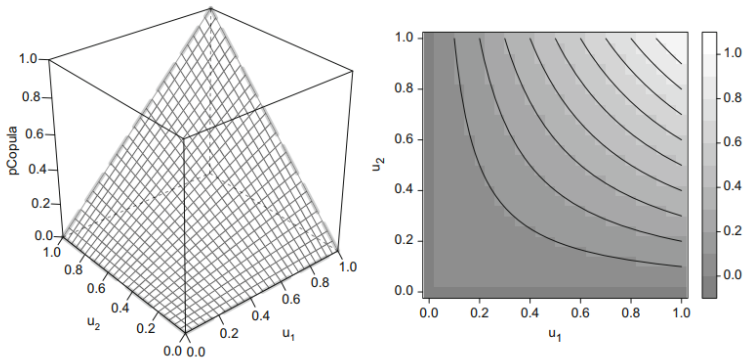


Figure: Graph of product copula and its contour graph.

# Invariance principle.

## Theorem.

Let  $\mathbf{X}$  and  $\mathbf{Y}$  are continuous random variables with distribution functions  $\mathbf{F}$  and  $\mathbf{G}$  respectively and a copula  $\mathbf{C}_{\mathbf{X}\mathbf{Y}}$ . If  $\alpha, \beta$  are strictly increasing on  $\mathbf{Ran}\mathbf{X} \times \mathbf{Ran}\mathbf{Y}$ , respectively, then

$$\mathbf{C}_{\mathbf{X}\mathbf{Y}} = \mathbf{C}_{\alpha(\mathbf{X})\beta(\mathbf{Y})}.$$

- ▶ If  $\alpha$  is strictly increasing and  $\beta$  is strictly decreasing, then

$$\mathbf{C}_{\alpha(\mathbf{X}),\beta(\mathbf{Y})}(\mathbf{u}, \mathbf{v}) = \mathbf{u} - \mathbf{C}_{\mathbf{X}\mathbf{Y}}(\mathbf{u}, 1 - \mathbf{v})$$

- ▶ If  $\alpha$  is strictly decreasing and  $\beta$  is strictly increasing, then

$$\mathbf{C}_{\alpha(\mathbf{X}),\beta(\mathbf{Y})}(\mathbf{u}, \mathbf{v}) = \mathbf{v} - \mathbf{C}_{\mathbf{X}\mathbf{Y}}(1 - \mathbf{u}, \mathbf{v})$$

- ▶ If  $\alpha$  and  $\beta$  are both strictly decreasing, then

$$\mathbf{C}_{\alpha(\mathbf{X}),\beta(\mathbf{Y})}(\mathbf{u}, \mathbf{v}) = \mathbf{u} + \mathbf{v} - 1 + \mathbf{C}_{\mathbf{X}\mathbf{Y}}(1 - \mathbf{u}, 1 - \mathbf{v})$$

# The Fréchet-Hoeffding Bounds

## Theorem.

If  $X$  and  $Y$  are two random variables with  $F, G$  are their distribution function respectively. Then for any  $x, y$  in  $\overline{\mathbb{R}}$ , and any  $u, v$  in  $[0, 1]$ , one has

$$\max(u + v - 1, 0) \leq C(u, v) \leq \min(u, v)$$

$$\max(F(x) + G(y) - 1, 0) \leq H(x, y) \leq \min(F(x), G(y))$$

- Call  $W(u, v) = \max(u + v - 1, 0)$ ,  $M(u, v) = \min(u, v)$ .  
Then  $W(u, v)$  and  $M(u, v)$  are copulas

# The Fréchet-Hoeffding Bounds

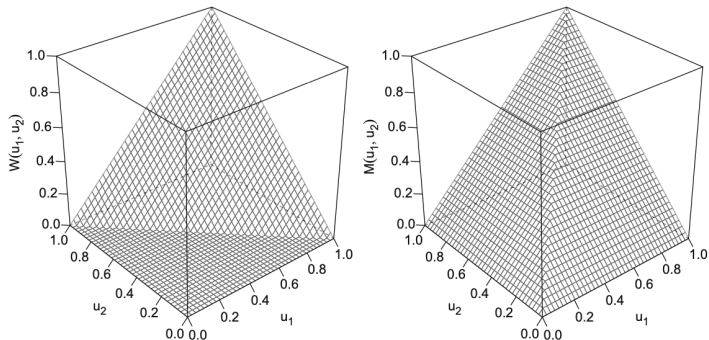


Figure: The Fréchet-Hoeffding Bounds Lower bound and Upper bound.



# Example

- Considering the **Frank Copula**

$$C^F(u, v) = \frac{1}{9} \log \left( 1 + \frac{(e^{9u} - 1)(e^{9v} - 1)}{e^9 - 1} \right), (u, v) \in [0, 1]^2$$

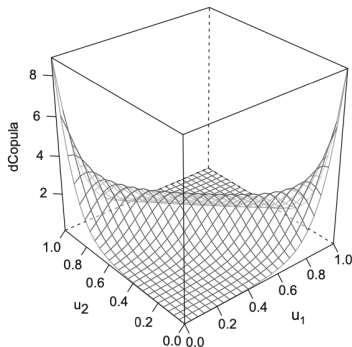
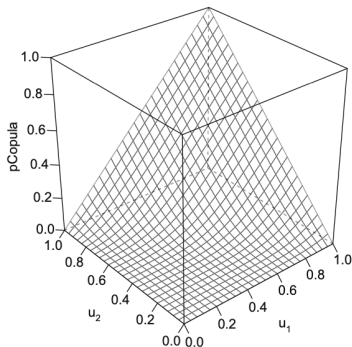


Figure: Frank copula and its density.

# Example

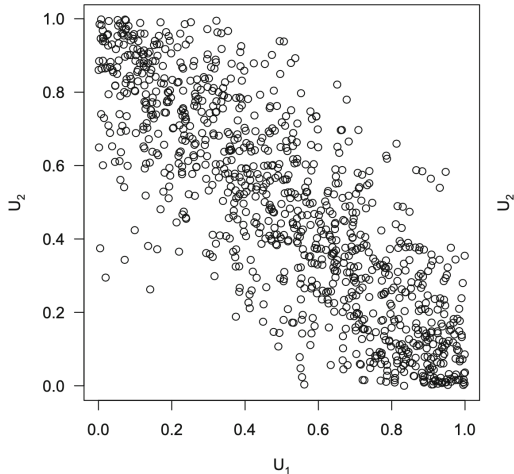


Figure:  $n = 1000$  independent observations  $(U_1, U_2) \sim C^F$ .

**Natural Question:** What happens when the equality occurs?

- ▶ A subset  $S$  of  $\overline{\mathbb{R}^2}$  is nondecreasing if for any  $(x, y), (u, v)$  in  $S$ ,  $x < u$  implies  $y \leq v$ .

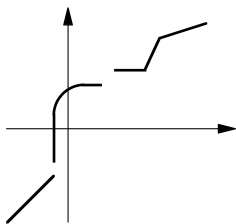


Figure: An example of nondecreasing set.

- ▶ A subset  $S$  of  $\overline{\mathbb{R}^2}$  is nonincreasing if for any  $(x, y), (u, v)$  in  $S$ ,  $x < u$  implies  $y \geq v$ .

# The equalities.

- ▶ The **support** of a copula  $\mathbf{C}$  is the complement of the union of all open subset of  $\mathbb{I}^2$  with  $\mathbf{C}$  - measure zero.

## Theorem.

Let  $\mathbf{X}$  and  $\mathbf{Y}$  be random variables with joint distribution function  $\mathbf{H}$ . Then  $\mathbf{H}$  is equal to its **Fréchet-Hoeffding upper bound** if and only if the support of  $\mathbf{H}$  is a **nondecreasing subset** of  $\overline{\mathbb{R}^2}$ .

Analogously,  $\mathbf{H}$  is equal to its **Fréchet-Hoeffding lower bound** if and only if the support of  $\mathbf{H}$  is a **nonincreasing subset** of  $\overline{\mathbb{R}^2}$ .

**Example:** The Fréchet-Hoeffding upper bound is a copula and its support set is the main diagonal.