

Notes 4. Math 481/581 Mathematical Statistics I

(p.1)

Uniform Distribution on a region R of the plane.

If R is a region of the plane with area $|R| < \infty$, then a joint density for a uniformly distributed random pair (X, Y) is defined by:

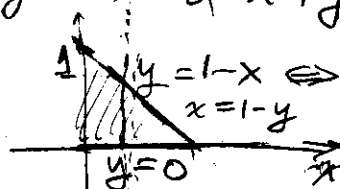
$$f(x, y) = \begin{cases} \frac{1}{|R|} & , (x, y) \in R \\ 0 & , \text{else.} \end{cases}$$

(f is constant for $(x, y) \in R$ and zero otherwise)

Example. Let $R = \{(x, y) : x \geq 0, y \geq 0, \text{ and } x + y \leq 1\}$.

Then $|R| = \frac{1}{2} \cdot 1 \cdot 1 = 1/2$.

$$f(x, y) = \begin{cases} 2 & , (x, y) \in R \\ 0 & \text{else.} \end{cases}$$



The marginal density of X is obtained as:

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_0^{1-x} 2 dy = \boxed{2(1-x), 0 \leq x \leq 1}$$

$$\text{Similarly } f_Y(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_0^{1-y} 2 dx = \boxed{2(1-y), 0 \leq y \leq 1}$$

Obviously X and Y are not independent r.v.'s since $f(x, y) \neq f_X(x) \cdot f_Y(y)$.

Example. Let X and Y be independent and each uniformly distributed on $[0, T]$.

$$\text{Thus } f_X(x) = \begin{cases} 1/T & , 0 \leq x \leq T \\ 0 & , \text{else} \end{cases} \text{ and } f_Y(y) = \begin{cases} 1/T & , 0 \leq y \leq T \\ 0 & , \text{else} \end{cases}$$

$$\text{So by independence } f(x, y) = f_X(x) f_Y(y) = \begin{cases} 1/T^2 & , 0 \leq x \leq T, 0 \leq y \leq T \\ 0 & \text{else} \end{cases}$$

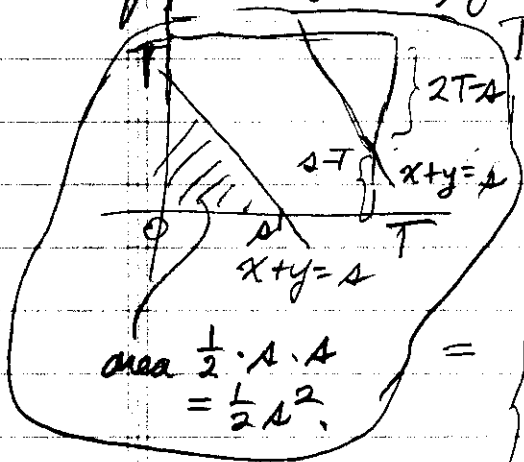
Thus the random pair (X, Y) is uniformly distributed on the square $\begin{matrix} 0 \leq x \leq T, \\ 0 \leq y \leq T, \end{matrix}$

Notes 4

Ex. 2 Example. Let X and Y be independent with each uniformly distributed on $[0, T]$. Find the cumulative distribution function of $S := X + Y$.

We want to calculate $P(S \leq a)$ for each $a \in [0, 2T]$. Since (X, Y) has density $f(x, y) = 1/T^2$, $0 \leq x \leq T$, $0 \leq y \leq T$ and $f(x, y) = 0$, else, and since $P(X + Y \leq a) = \iint_{x+y \leq a} f(x, y) dx dy$,

we integrate the constant $1/T^2$ over the part of the square $0 \leq x, y \leq T$ such that $x + y \leq a$.



Thus $P(S \leq a) = P(X + Y \leq a)$

$$= \begin{cases} \frac{\frac{1}{2} a^2}{T^2}, & 0 \leq a \leq T \\ 1 - \frac{\frac{1}{2} (2T-a)^2}{T^2} & T \leq a \leq 2T \end{cases}$$

$$= \begin{cases} \frac{1}{2} \left(\frac{a}{T}\right)^2 & 0 \leq a \leq T \\ 1 - \frac{1}{2} \left(2 - \frac{a}{T}\right)^2 & T \leq a \leq 2T. \end{cases}$$

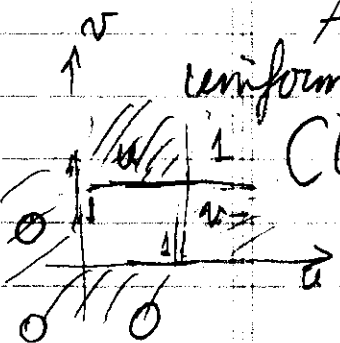
The density of S is therefore $\frac{d}{da} P(S \leq a)$ or

$$f_S(a) = \begin{cases} a/T^2 & 0 \leq a \leq T \\ \left(2 - \frac{a}{T}\right) \left(\frac{1}{T}\right) & T \leq a \leq 2T. \end{cases}$$

Copula

A copula is a joint distribution with uniform marginal distributions. The following is a copula for $|k| \leq 1$:

$$C(u, v) = \begin{cases} uv (1 + k(1-u)(1-v)) & , 0 \leq u \leq 1, 0 \leq v \leq 1, \\ 1 & \text{where } u \geq 1 \text{ and } v \geq 1, \\ 0 & \text{where } u \text{ or } v \leq 0, \\ u & , \text{ where } 0 \leq u \leq 1 \text{ and } v \geq 1, \\ v & , \text{ where } 0 \leq v \leq 1 \text{ and } u \geq 1. \end{cases} \text{ (Goulet)}$$



Notes

(p.3)

We have that

$C(u, v)$ is a joint cumulative distribution function of a pair U, V because

$C(u, v)$ is non-decreasing and continuous. Indeed we have a joint density $c(u, v) := \frac{\partial^2 C(u, v)}{\partial u \partial v} = \alpha(1-2u)(1-2v) + 1$, $0 \leq u, v \leq 1$

that is non-negative since $|1-2u| \leq 1$ and $|1-2v| \leq 1$

for $0 \leq u, v \leq 1$ and since $|\alpha| \leq 1$ by assumption.

Furthermore $C(u, \infty) := P(U \leq u, V < \infty)$

$$= P(U \leq u) = \begin{cases} 0 & u \leq 0 \\ u & 0 \leq u \leq 1 \\ 1 & u \geq 1 \end{cases}$$

$$\text{Similarly } C(\infty, v) = P(V \leq v) = \begin{cases} 0 & v \leq 0 \\ v & 0 \leq v \leq 1 \\ 1 & v \geq 1 \end{cases}$$

while $C(\infty, \infty) := \lim_{u, v \rightarrow \infty} C(u, v) = 1$.

Thus C is a joint cdf. with uniform marginals.

In general, if $F(x)$ and $G(y)$ are distribution functions

Then $H(x, y) = F(x)G(y) \{1 + \alpha(1-F(x))(1-G(y))\}$.

is a joint cdf. with marginals F and G

whenever $|\alpha| \leq 1$. The density of $H(x, y)$ is by the chain rule:
 $h(x, y) = c(F(x), G(y)) \cdot f(x)g(y)$ when $F(x)$ and $G(y)$ have densities $f(x)$ and $g(y)$ respectively.

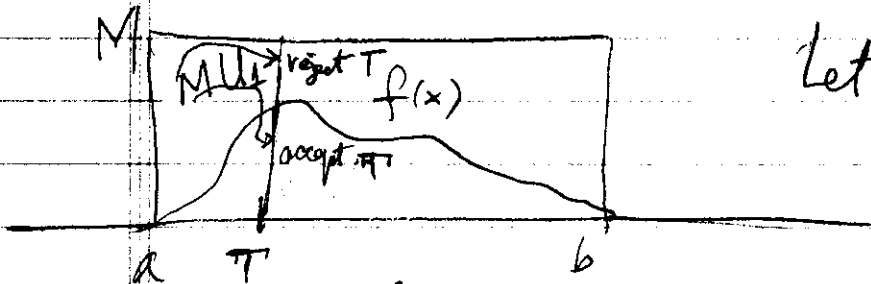
Note that $c(u, v)$ is not constant unless $\alpha = 0$

so only in the case $\alpha = 0$ does the copula correspond to the case of a joint distribution of independent random variables

Notes 4

(p.4) Rejection Method for Simulating a random Variable.

Let f be a density that is non-zero only on an interval $[a, b]$, and let M be a constant $\geq f(x)$, $x \in [a, b]$.



$$\text{Let } m(x) = \begin{cases} \frac{1}{b-a} & a \leq x \leq b \\ 0 & \text{else} \end{cases}$$

be the density of a Uniform r.v. on $[a, b]$.

We first generate T with density $m(x)$.

$$(T = a + (b-a)U_0 \text{ for } U_0 \sim \text{Uniform}[0, 1])$$

Second generate $U_1 \sim \text{Uniform}[0, 1]$ independent of T

if $M U_1 \leq f(T)$ then let $X_i = T$ (accept T), otherwise reject T and go through these two steps again.

Claim X has density f .

$$\begin{aligned} P(X \in [x, x+dx]) &= P(T \in [x, x+dx] \mid \text{accept } T) \\ &= \frac{P(T \in [x, x+dx] \text{ and accept } T)}{P(\text{accept } T)} \\ &= \frac{P(M U_1 \leq f(T) \text{ and } T \in [x, x+dx])}{\int_a^b \frac{f(x)}{M} \frac{dx}{b-a}} \\ &= \frac{(f(x)/M) \cdot (dx)}{\int_a^b \frac{f(x)}{M} \frac{dx}{b-a}} = \frac{f(x) dx}{\int_a^b f(x) dx} = \frac{f(x) dx}{1} \\ &= f(x) dx \quad \text{Q.E.D.} \end{aligned}$$