

Probability and Statistics

4 – Continuous Random Variables

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Joint Random Variables

Definition (4.20)

Let X, Y be random variables with respect to the same probability measure \Pr , i.e. with respect to the same triple $(\Omega, \mathcal{A}, \Pr)$. The *joint cumulative distribution function* of X and Y , $F_{XY} : \mathbb{R}^2 \rightarrow [0, 1]$, is defined by:

$$\begin{aligned} F_{XY}(x, y) &:= \Pr(X \leq x, Y \leq y) \\ &= \Pr(X^{-1}((-\infty, x]) \cap Y^{-1}((-\infty, y])) \quad \text{for all } x, y \in \mathbb{R} \end{aligned}$$

Joint Cumulative Distribution Functions

Lemma (4.21)

Let X, Y be random variables with respect to the same probability measure \Pr and $(x_1, x_2] \times (y_1, y_2]$ be a rectangle. Then:

$$\begin{aligned}\Pr(x_1 < X \leq x_2, y_1 < Y \leq y_2) \\ = F_{XY}(x_2, y_2) - F_{XY}(x_1, y_2) - F_{XY}(x_2, y_1) + F_{XY}(x_1, y_1)\end{aligned}$$

Marginal Cumulative Distribution Functions

Lemma (4.22)

Let F_{XY} be the joint cumulative distribution function of two random variables X and Y . The (marginal) cumulative distribution functions of X and Y are determined by F_{XY} as follows:

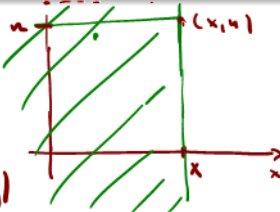
$$F_X(x) = \lim_{y \rightarrow \infty} F_{XY}(x, y)$$

$$F_Y(y) = \lim_{x \rightarrow \infty} F_{XY}(x, y)$$

$$F_X(x) = P_r(X \leq x) = P_r\left(\bigcup_{n \in \mathbb{N}} \underbrace{\{\omega | X(\omega) \leq x, Y(\omega) \leq n\}}_{A_n}\right)$$

$$\stackrel{4.3}{=} \lim_{n \rightarrow \infty} P_r(X \leq x, Y \leq n)$$

$$= \lim_{n \rightarrow \infty} F_{XY}(x, n) = \lim_{y \rightarrow \infty} F_{XY}(x, y)$$



Independent Random Variables

Definition (4.23)

Two random variables X and Y , defined with respect to the same probability measure \Pr , are called *independent*, if for all intervals $S, T \subseteq \mathbb{R}$:

$$\Pr(X \in S, Y \in T) = \Pr(X \in S) \cdot \Pr(Y \in T)$$

$$A = X^{-1}(S), B = Y^{-1}(T)$$

$$\Pr(A \cap B) = \Pr(A) \cdot \Pr(B)$$

Independent Random Variables

Lemma (4.24)

Let X, Y be random variables with respect to the same probability measure \Pr . Then X and Y are independent, if and only if one of the following holds:

(i) For all $x_1, x_2, y_1, y_2 \in \mathbb{R}$ with $x_1 < x_2$ and $y_1 < y_2$:

$$S = (x_1, x_2], T = (y_1, y_2]$$

$$\Pr(x_1 < X \leq x_2, y_1 < Y \leq y_2) = (F_X(x_2) - F_X(x_1)) \cdot (F_Y(y_2) - F_Y(y_1))$$

(ii) For all $x, y \in \mathbb{R}$:

$$F_{XY}(x, y) = F_X(x) \cdot F_Y(y)$$

$$S = (-\infty, x], T = (-\infty, y]$$

Proof of Lemma (4.24)

Show: $\Pr((X, Y) \in S \times T) = \Pr(X \in S) \cdot \Pr(Y \in T)$ for all intervals $S, T \subseteq \mathbb{R}$

(i) $S = \mathbb{R} = (-\infty, \infty)$

(ii) $S = (-\infty, x)$

see (4.2), (4.3)

(iii) $S = [x, \infty)$

(iv) $S = (x, \infty)$

(v) $S = (x_1, x_2)$

(vi) $S = (x_1, x_2]$

(vii) $S = [x_1, x_2)$

(viii) $S = [x_1, x_2]$

Proof of Lemma (4.24)

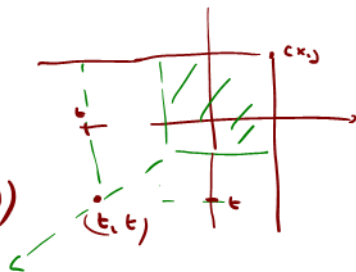
E.g. (i) \Rightarrow (ii)

$$\bar{F}_{X,Y}(x,y) \stackrel{(4.3)}{=} \lim_{t \rightarrow -\infty} P(t < X \leq x, t < Y \leq y)$$

$$= \lim_{t \rightarrow -\infty} (F_X(x) - \bar{F}_X(t)) \cdot (F_Y(y) - \bar{F}_Y(t))$$

$$= \lim_{t \rightarrow -\infty} (F_X(x) - \bar{F}_X(t)) \cdot \lim_{t \rightarrow -\infty} (F_Y(y) - \bar{F}_Y(t))$$

$$= (F_X(x) - 0) \cdot (F_Y(y) - 0) = F_X(x) \cdot F_Y(y)$$



Proof of Lemma (4.24)

(ii) \Rightarrow (i)

$$P_r(x_1 < X \leq x_2, y_1 < Y \leq y_2)$$

$$\stackrel{(4.24)}{=} F_{X,Y}(x_2, y_2) - F_{X,Y}(x_1, y_2) - \bar{F}_{X,Y}(x_2, y_1) + \bar{F}_{X,Y}(x_1, y_1)$$

$$\stackrel{(\dots)}{=} F_X(x_2) F_Y(y_2) - F_X(x_1) F_Y(y_2) - \bar{F}_X(x_2) \bar{F}_Y(y_1) + \bar{F}_X(x_1) \bar{F}_Y(y_1)$$

$$= (F_X(x_2) - F_X(x_1)) (F_Y(y_2) - F_Y(y_1))$$

□

Independent Random Variables

Lemma (4.25)

Let X, Y be joint independent random variables and $\underline{h, k : \mathbb{R} \rightarrow \mathbb{R}}$ be piecewise continuous functions, then $h(X)$ and $k(Y)$ are independent.

Proof of Lemma (4.25)

$$S, T \subseteq \mathbb{R} \text{ intervals} \quad h^{-1}(S) = \bigcup_{i=1}^m S_i, \quad h^{-1}(T) = \bigcup_{j=1}^n T_j$$

$$\Pr(h(X) \in S, h(Y) \in T) = \Pr(X \in h^{-1}(S), Y \in h^{-1}(T))$$

$$= \Pr(X \in \bigcup_{i=1}^m S_i, Y \in \bigcup_{j=1}^n T_j)$$

$$= \Pr((X, Y) \in \bigcup_{i,j} S_i \times T_j) = \sum_{i,j} \Pr((X, Y) \in S_i \times T_j)$$

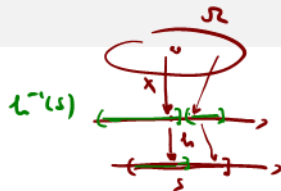
assumption: X, Y are independent! \downarrow

$$= \sum_{i,j} \Pr(X \in S_i) \cdot \Pr(Y \in T_j) = \sum_{i=1}^m \Pr(X \in S_i) \cdot \sum_{j=1}^n \Pr(Y \in T_j)$$

$$= \Pr(X \in \bigcup_{i=1}^m S_i) \cdot \Pr(Y \in \bigcup_{j=1}^n T_j)$$

$$= \Pr(X \in h^{-1}(S)) \cdot \Pr(Y \in h^{-1}(T))$$

$$= \Pr(h(X) \in S) \cdot \Pr(h(Y) \in T)$$



Jointly Continuous Random Variables

Definition (4.26)

Two random variables X and Y are jointly continuous random variables, if there exists a function $f_{XY} : \mathbb{R}^2 \rightarrow \mathbb{R}_0^+$, such that

$$F_{XY}(x, y) = \int_{-\infty}^y \int_{-\infty}^x f_{XY}(s, t) ds dt$$

for all $x, y \in \mathbb{R}$. The function f_{XY} is called a joint probability density function of X and Y .

Jointly Continuous Random Variables

Lemma (4.27)

Let f_{XY} be a joint probability density function of two random variables X and Y . The (marginal) probability density functions of X and Y are determined by f_{XY} as follows:

$$f_X(x) = \int_{-\infty}^{\infty} f_{XY}(x, t) dt \quad \checkmark$$

$$f_Y(y) = \int_{-\infty}^{\infty} f_{XY}(s, y) ds$$

$$F_X(x) = \lim_{y \rightarrow \infty} F_{XY}(x, y) = \lim_{y \rightarrow \infty} \int_{-\infty}^x \left(\int_{-\infty}^y f_{XY}(s, t) dt \right) ds = \int_{-\infty}^x \underbrace{\left(\int_{-\infty}^{\infty} f_{XY}(s, t) dt \right)}_{f(s)} ds$$

is a pdf for X

Jointly Continuous Random Variables

Lemma (4.28)

Let f_{XY} be a joint probability density function of two random variables X and Y . If f_{XY} is continuous, then for all $x, y \in \mathbb{R}$:

$$\frac{\partial^2}{\partial y \partial x} F_{XY}(x, y) = \frac{\partial^2}{\partial x \partial y} F_{XY}(x, y) = f_{XY}(x, y)$$

$$\begin{aligned} \frac{\partial^2}{\partial y \partial x} F_{XY}(x, y) &= \frac{\partial}{\partial x} \left(\frac{\partial}{\partial y} \left(\int_{-\infty}^y \left(\int_{-\infty}^x f_{XY}(s, t) \, ds \right) dt \right) \right) \\ &= \frac{\partial}{\partial x} \left(\int_{-\infty}^y f_{XY}(x, t) \, dt \right) = f_{XY}(x, y) \end{aligned}$$