

Conditionnement - Indépendance

X, Y couple VA idem (x_1, \dots, x_n)

VAD

loi jointe de (X, Y)

$$P(X=x, Y=y)$$

loi conditionnelle de X sachant Y

$$P(X=x | Y=y) = \frac{P(X=x, Y=y)}{P(Y=y)}$$

loi marginale de X

$$P(X=x) = \sum_y P(X=x, Y=y)$$

Esperance

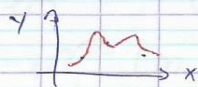
$$E[g(X, Y)] = \sum_x \sum_y g(x, y) P(X=x, Y=y)$$

Esperance conditionnelle de X sachant Y

$$E[g(X, Y) | Y=y] = \sum_x g(x, y) P(X=x | Y=y)$$

$$E[E(X|Y)] = E[X]$$

$$= \sum_y E(X|Y=y) P(Y=y)$$



VAC

densité jointe $f_{(X,Y)}(x,y)$

$$P((X, Y) \in \mathcal{D}) = \iint_{\mathcal{D}} f_{(X,Y)}(x,y) dx dy$$

densité conditionnelle

$$f_{X|Y=y}(x) = \frac{f_{(X,Y)}(x,y)}{f_Y(y)}$$

$$P(X \in A | Y=y) = \int_A f_{X|Y=y}(x) dx$$

densité marginale

$$f_X(x) = \int_{\mathcal{R}} f_{(X,Y)}(x,y) dy$$

$$E[g(X, Y)] = \iint_{\mathcal{R}^2} g(x, y) f_{(X,Y)}(x,y) dx dy$$

$$\rightarrow \int_{\mathcal{R}} g(x, y) f_{X|Y=y}(x) dx$$

$$\int E(X|Y=y) f_Y(y) dy$$

la "meilleure" fct de X qui approche Y est

$$Y(X) = E(Y|X)$$

"meilleure" au sens de p.d.f. scl $\langle T, u \rangle = E(Tu)$

$$E(Y|X) = \operatorname{argmin}_{\varphi \text{ mesurable}} E(Y - \varphi(X))^2$$

Indép

$$P(X \in A, Y \in B) \\ = P(X \in A) P(Y \in B) \quad \forall A, B \in \mathcal{B}(\mathbb{R})$$

idem

$$f_{(X,Y)}(x,y) = f_x(x) f_y(y)$$

$$E[g(X)h(Y)] = E[g(X)]E[h(Y)] \quad \text{indép} \Rightarrow \text{Cov}(X,Y) = 0$$

fonct génératrice

$$\forall z \in [-1, 1] \\ G_x(z) = E[z^X]$$

fonct caractéristique $\forall z \in \mathbb{R}$

$$\chi_x(z) = E[e^{izX}] \\ = \int e^{izx} f_x(x) dx$$

$$= \mathcal{F}(f)(z)$$

$$f = \frac{1}{2\pi} \int e^{-izx} \hat{f}$$

$$X, Y \text{ indep} \quad \chi_{X+Y}(z) = \chi_X(z) \chi_Y(z)$$

$$\text{et } f_{X+Y}(t) = f_X(t) * f_Y(t) \\ = \int_{\mathbb{R}} f_X(t-y) f_Y(y) dy$$

$$F_X(x) = P(X \leq x) \quad f_X(x) = F_X'(x) \\ F_{X+Y}(t) = P(X+Y \leq t) = \int_{\mathbb{R}} P((X,Y) \in \mathcal{D}) \quad \text{avec } \mathcal{D} = \{(x,y) \in \mathbb{R}^2 \mid x+y \leq t\}$$

$$= \iint_{\mathcal{D}} \underbrace{f_{(X,Y)}(x,y)}_{f_X(x) f_Y(y) \text{ car indep}} dx dy$$

$$= \int_{\mathbb{R}} \left(\int_{-\infty}^{t-y} f_X(x) dx \right) f_Y(y) dy$$

$$= \int_{\mathbb{R}} F_X(t-y) f_Y(y) dy$$

$$f_{X+Y}(t) = \frac{\partial}{\partial t} F_{X+Y}(t) = \frac{\partial}{\partial t} \int_{\mathbb{R}} \int_{-\infty}^{t-y} f_X(x) dx f_Y(y) dy \\ = \frac{\partial}{\partial t} \int_{\mathbb{R}} F_X(t-y) f_Y(y) dy \\ = \int_{\mathbb{R}} \frac{\partial}{\partial t} F_X(t-y) f_Y(y) dy$$

$$= f_x \cdot f_y (t)$$

Rappel $(u, v) = \varphi(x, y)$ $\varphi: \mathcal{E}^2$ diffeo

$$P((x, y) \in \mathcal{D}) = \iint_{\mathcal{D}} f_{xy}(x, y) dx dy$$

$$= \iint_{\varphi(\mathcal{D})} f_{(u, v)}(u, v) du dv$$

$$f_{(u, v)}(u, v) = f_{(x, y)}(\varphi^{-1}(u, v)) \times \left| \det \left(\nabla \varphi^{-1}(u, v) \right) \right|$$

$$\Psi = (\Psi_1, \dots, \Psi_d) \quad \nabla \Psi = \frac{\partial \Psi_i}{\partial x_j} \quad 1 \leq i, j \leq d$$

$$\Psi_1(x_1, \dots, x_d)$$

$$\vdots$$

$$\Psi_d(x_1, \dots, x_d)$$