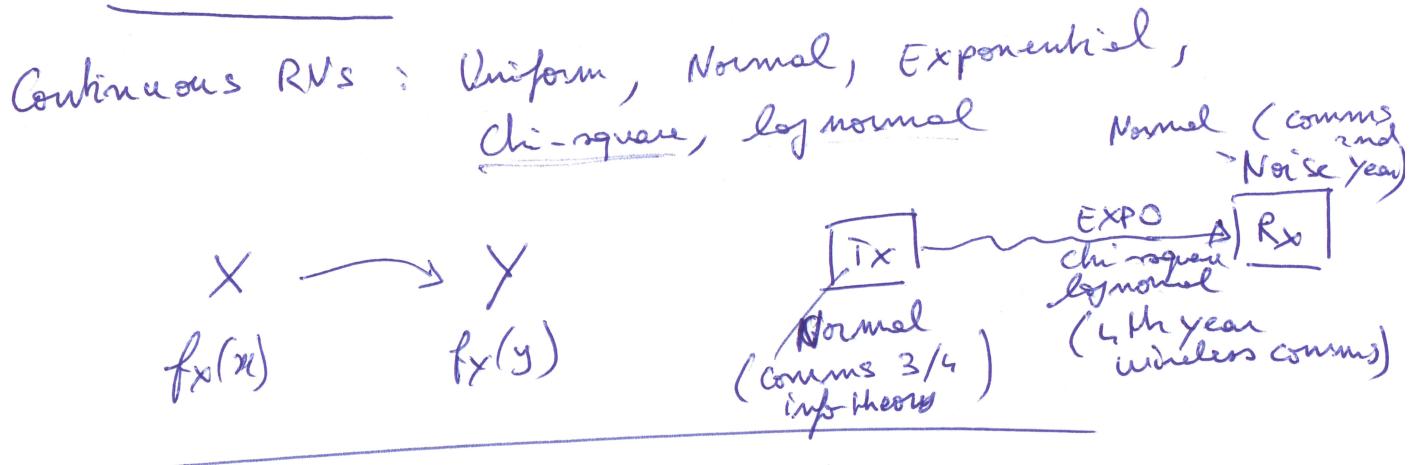


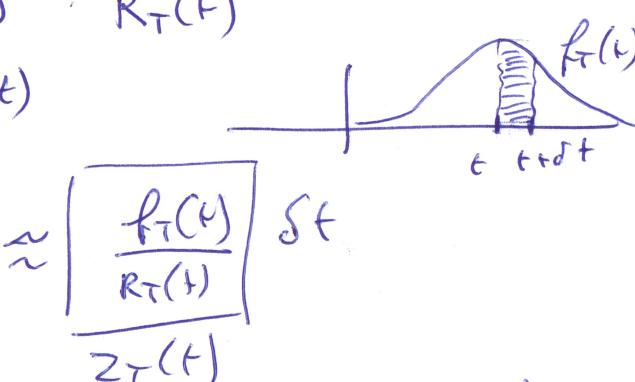
Lecture 9/10



$$\begin{aligned}
 & \text{Let } \\
 & \int_{-\infty}^{2.35} \sqrt{\frac{2}{\pi}} e^{-2(u-2)^2} du \\
 & = \int_{-\infty}^{2.35} \sqrt{\frac{1}{2\pi \cdot \frac{1}{4}}} e^{-\frac{(u-2)^2}{2 \cdot \frac{1}{4}}} du \quad f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \\
 & \qquad \qquad \qquad \text{PDF of } UN N(\mu=2, \sigma^2=\frac{1}{4}) \quad N(0,1) \\
 & = F_U(2.35) = P(U \leq 2.35) \quad U = \sigma Z + \mu \\
 & = P\left(\frac{1}{2}Z + 2 \leq 2.35\right) \quad = \frac{1}{2}Z + 2 \\
 & = P(Z \leq 0.7) = 0.758
 \end{aligned}$$

Reliability

- RV T random time to failure (non-negative)
- CDF of T $F_T(t) = P(T \leq t) = \int_0^t f_T(u) du$
failure time distribution
- PDF of T $f_T(t)$ failure time density

$$f_T(t) = \frac{d}{dt} R_T(t) = \frac{d}{dt} (1 - F_T(t)) = -\frac{d}{dt} R_T(t)$$
- $R_T(t) = 1 - F_T(t) = 1 - P(T \leq t) = P(T > t)$
reliability function
- $A = \{t < T < t + \delta t\}$ 
- $B = \{t < T\}$
- $$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(A)}{P(B)} \approx \frac{\frac{f_T(t) \delta t}{R_T(t)}}{P(T > t)}$$

- $Z_T(t)$ hazard rate
(measure of propensity to failure at time t)
- cumulative hazard function

$$H_T(t) = \int_0^t Z_T(u) du$$

$$= \int_0^t \frac{f_T(u)}{R_T(u)} du = - \left[\ln R_T \right]_0^t$$

$$= - \ln R_T(t) + \underbrace{\ln R_T(0)}_{P(T > 0) = 1}$$

$$H_T(t) = - \ln R_T(t)$$

$$R_T(t) = e^{-H_T(t)}$$

(2)

$$\underline{\text{ex}} \quad \text{hazard rate} \quad \underline{\underline{z_T(t) = \lambda}}$$

$$H_T(t) = \int_0^t z_T(u) du = \int_0^t \lambda du = \lambda t$$

$$R_T(t) = e^{-H_T(t)} = e^{-\lambda t}$$

$$f_T(t) = -\frac{d}{dt} R_T(t) = \underline{\underline{\lambda e^{-\lambda t}}}$$

TN EXP(λ)

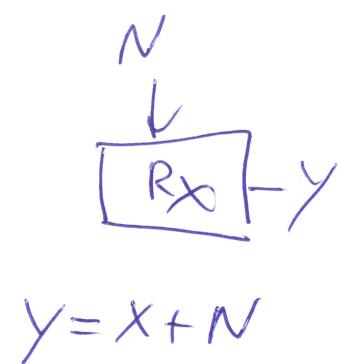
→ reminiscent of memoryless property

$$P(T > a+b | T > a) = P(T > b)$$

- $E[T]$ mean time to failure (MTTF)

$$E(T) = \int_0^\infty t f_T(t) dt = \frac{\int_0^\infty R_T(t) dt}{(\text{MTTF} < \infty)}$$

Jointly Distributed RVs



RV X	CDF	$F_X(x) = P(X \leq x)$
	PDF PMF	$f_X(x)$

RVs X, Y joint CDF

$$F_{X,Y}(a, b) = P(X \leq a \cap Y \leq b)$$

AND

Discrete RVs

X, Y

- ex X throw a die $\{1, 2, 3, 4, 5, 6\}$
 Y toss a coin $\{H, T\}$

$$S = \{(1, H), (1, T), (2, H), (2, T), \dots, (6, H), (6, T)\}$$

Joint PMF

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

Valid PMF : $f_{X,Y}(x,y) \geq 0 \quad \forall x, y$

$$\sum_x \sum_y f_{X,Y}(x,y) = 1$$

$$P((X, Y) \in A) = \sum_x \sum_y f_{X,Y}(x,y) \quad \text{for } (x, y) \in A$$

ex

		1	2	3	$f_{Y X}(y)$	
		5	0.2	0.1	0	0.3
		6	0.2	0.1	0.1	0.4
		7	0.1	0.1	0.1	0.3
$f_X(x)$		0.5	0.3	0.2		

$$P(X=1, Y=5) = 0.2$$

$$A = \{(1, 5), (2, 6), (3, 7)\}$$

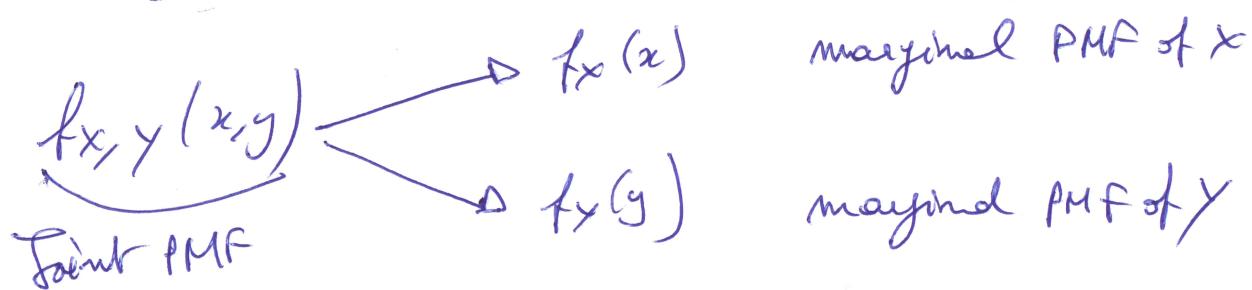
$$P((X, Y) \in A) =$$

$$P(X=1, Y=5) +$$

$$P(X=2, Y=6) + P(X=3, Y=7)$$

$$= f_{X,Y}(1, 5) + f_{X,Y}(2, 6) \\ + f_{X,Y}(3, 7)$$

$$= 0.2 + 0.1 + 0.1 = 0.4$$



$$f_X(x) = \sum_y f_{X,Y}(x,y)$$

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

ex

		1	2	3
$f_X(x)$	0.5	0.3	0.2	

		5	6	7
$f_Y(y)$	0.3	0.4	0.3	

$$1) f_X(x) \geq 0$$

$$f_Y(y) \geq 0$$

$$2) \sum_x f_X(x) = 1$$

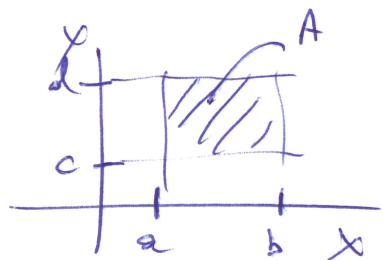
$$\sum_y f_Y(y) = 1$$

Continuous RVs

Joint PDF

$$P(X \in A) = \int_A f_X(x) dx$$

$$P((X,Y) \in A) = \iint_A \underbrace{f_{X,Y}(x,y)}_{\text{Joint PDF}} dx dy$$



$$P(X, Y \in A) = \int_a^b \int_c^d f_{X,Y}(x,y) dy dx$$

Valid PDF

- $f_{X,Y}(x,y) \geq 0 \quad \forall x,y$
- $\iint_{-\infty}^{+\infty} f_{X,Y}(x,y) dx dy = 1$

Joint CDF $F_{X,Y}(a,b) = \int_{-\infty}^a \int_{-\infty}^b f_{X,Y}(x,y) dx dy$

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

$$f_X(x) = \frac{d}{dx} F_X(x)$$

ex $f_{X,Y}(x,y) = \begin{cases} \frac{6}{5}(x+y^2) & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$

Valid PDF?

- $f_{X,Y}(x,y) \geq 0 \quad \forall x,y$

$$\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dx dy = \int_0^1 \int_0^1 \frac{6}{5}(x+y^2) dx dy = 1$$

$$f_{x,y}(x,y) \quad \begin{matrix} \rightarrow \\ \leftarrow \end{matrix} \quad \begin{matrix} f_x(x) & \text{marginal PDF of } x \\ f_y(y) & \text{marginal PDF of } y \end{matrix}$$

$$f_x(x) = \int_{-\infty}^{+\infty} f_{x,y}(x,y) dy$$

$$f_y(y) = \int_{-\infty}^{+\infty} f_{x,y}(x,y) dx$$

$$f_{x,y}(x,y) \quad \begin{matrix} \rightarrow \\ \leftarrow \end{matrix} \quad \begin{matrix} f_x(x) \\ f_y(y) \end{matrix}$$

ex rectangular domain

$$f_{x,y}(x,y) = \begin{cases} \frac{6}{5}(x+y^2) & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

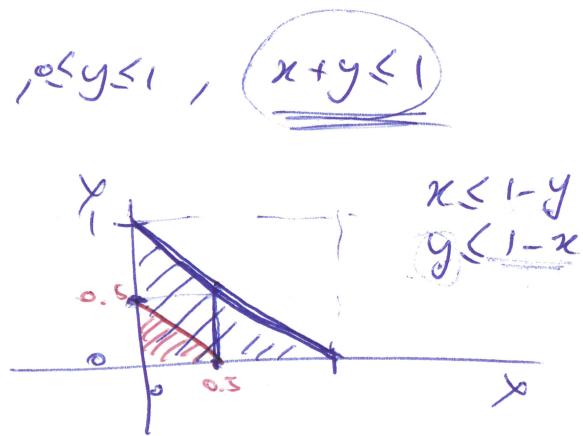
$$\begin{aligned} f_x(x) &= \int_{-\infty}^{+\infty} f_{x,y} dy = \int_0^1 \frac{6}{5}(x+y^2) dy \\ &= \frac{6}{5} \left[xy + \frac{y^3}{3} \right]_0^1 = \left\{ \begin{array}{l} \frac{6}{5} \left[x + \frac{1}{3} \right] \\ 0 \end{array} \right. \quad \begin{array}{l} 0 \leq x \leq 1 \\ \text{otherwise} \end{array} \end{aligned}$$

$$\begin{aligned} f_{x,y}(x,y) &= f_x(x) f_y(y) \\ x, y &\text{ dependent} \\ f_y(y) &= \int_{-\infty}^{+\infty} f_{x,y} dx = \int_0^1 \frac{6}{5}(x+y^2) dx \\ &= \frac{6}{5} \left[\frac{x^2}{2} + y^2 x \right]_0^1 = \left\{ \begin{array}{l} \frac{6}{5} \left(\frac{1}{2} + y^2 \right) \quad 0 \leq y \leq 1 \\ 0 \quad \text{otherwise} \end{array} \right. \end{aligned}$$

$$P(X \leq 0.5) = \int_0^{0.5} f_x(x) dx$$

ex non rectangular domain

$$f_{x,y}(x,y) = \begin{cases} 2xy & 0 \leq x \leq 1, 0 \leq y \leq 1, \\ & x+y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$



Valid joint PDF?

- $f_{x,y}(x,y) \geq 0 \quad \forall x, y$

- $\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f_{x,y} dx dy = \int_0^1 \int_0^{1-x} 2xy \, dy \, dx$
 $= \int_0^1 \int_0^{1-x} 2xy \, dx \, dy$
 $= 1$

- $A = \{(x,y) : 0 \leq x \leq 1, 0 \leq y \leq 1, x+y \leq 0.5\}$

$$P((X,Y) \in A) = \int_0^{0.5} \int_0^{0.5-x} 2xy \, dy \, dx$$
 $= \int_0^{0.5} \int_0^{0.5-y} 2xy \, dx \, dy$

- marginal of x

$$f_x(x) = \int_{-\infty}^{+\infty} f_{x,y} \, dy$$

$$\begin{aligned} &x+y \leq 1 \\ &y \leq 1-x \end{aligned}$$

$$= \int_0^{1-x} f_{x,y}(x,y) \, dy$$

Independent RVs

A, B events

$$P(A|B) = P(A)$$

A, B independent

$$P(A \cap B) = P(A)P(B)$$

X, Y RVs independent

$$\bullet P(X \leq x, Y \leq y) = P(X \leq x) P(Y \leq y) \quad \stackrel{f_{x,y}}{=}$$

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

$$\bullet f_{x,y}(x,y) = f_X(x) f_Y(y) \quad \text{continuous} \quad \stackrel{f_{x,y}}{=}$$

$$P(X=x, Y=y) = P(X=x) P(Y=y) \quad \text{discrete}$$

More than 2 RVs

$$X_1, \dots, X_m$$

Joint CDF $F_{X_1, X_2, \dots, X_m}(x_1, \dots, x_n) = P(X_1 \leq x_1, X_2 \leq x_2, \dots, X_m \leq x_n)$

Joint PDF $f_{X_1, X_2, \dots, X_m}(x_1, \dots, x_n) = \frac{\partial^n F_{X_1, X_2, \dots, X_m}}{\partial x_1 \dots \partial x_m}$

Independent RVs

$$F_{X_1, X_2, \dots, X_m}(x_1, \dots, x_n) = F_{X_1}(x_1) \dots F_{X_m}(x_m)$$

$$f_{X_1, X_2, \dots, X_m}(x_1, \dots, x_n)$$

$$f_{X_1, X_2, \dots, X_m}(x_1, \dots, x_n) = f_{X_1}(x_1) \dots f_{X_m}(x_m)$$

$$f_{X_1, \dots, X_m}(x_1, \dots, x_m)$$



Conditional probability

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

Conditional PMF / PDF

$$f_{Y|X}(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)} \geq 0$$

Valid PDF/PMF

- $f_{Y|X}(y|x) \geq 0$
- $\int_{-\infty}^{+\infty} f_{Y|X}(y|x) dy = 1$

$$\int_{-\infty}^{+\infty} \frac{f_{X,Y}(x,y)}{f_X(x)} dy = \frac{1}{f_X(x)} \underbrace{\int_{-\infty}^{+\infty} f_{X,Y}(x,y) dy}_{f_Y(y)} = 1$$

$$\cdot P(Y \leq y | X=x) = \int_{-\infty}^y f_{Y|X}(y|x) dy$$

• X, Y independent

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

$$P(A|B) = P(A)$$

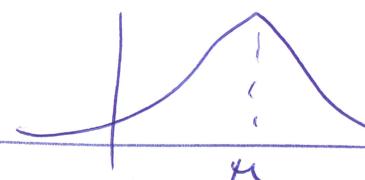
Any

$$f_{x,y}(x,y) = \begin{cases} \frac{6}{5}(x+y^2) & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0 & \text{else} \end{cases}$$

$$f_y(y) = \begin{cases} \frac{6}{5}y^2 + \frac{3}{5} & 0 \leq y \leq 1 \\ 0 & \text{else} \end{cases}$$

$$f_{x|y}(x|0.3) = \frac{f_{x,y}(x, 0.3)}{f_y(0.3)} = \begin{cases} \frac{\frac{6}{5}(x+0.3^2)}{\frac{6}{5} \cdot 0.3^2 + \frac{3}{5}} & 0 \leq x \leq 1 \\ 0 & \text{else} \end{cases}$$

Expectation

$$X \quad \mu = E(X) = \int_{-\infty}^{+\infty} x f_x(x) dx$$


$$Y = g(X) \quad E(Y) = \begin{aligned} & \text{1) first calculate } f_Y(y), \\ & \text{then calculate } E(Y) \\ & = \int_{-\infty}^{+\infty} y f_Y(y) dy \\ & \text{2) } E(g(X)) = \int_{-\infty}^{+\infty} g(x) f_X(x) dx \end{aligned}$$

$$\mathbb{E}(g(X, Y)) = \begin{cases} \sum_x \sum_y g(x, y) f_{x,y}(x, y) & \text{discrete} \\ \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} g(x, y) f_{x,y}(x, y) dx dy & \text{continuous} \end{cases}$$

- $f(x,y) = X$

$$\begin{aligned} E(f(x,y)) &= E(X) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} x f_{x,y}(x,y) dy dx \\ &= \int_{-\infty}^{+\infty} x \underbrace{\int_{-\infty}^{+\infty} f_{x,y}(x,y) dy}_{f_x(x)} dx \\ &= \int_{-\infty}^{+\infty} x f_x(x) dx = \underline{\underline{E(X)}} \end{aligned}$$

- $f(x,y) = (x - \mu_x)^2 \quad \mu_x = E(X)$

$$\begin{aligned} E(f(x,y)) &= E((x - \mu_x)^2) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} (x - \mu_x)^2 f_{x,y}(x,y) dy dx \\ &= \int_{-\infty}^{+\infty} (x - \mu_x)^2 \underbrace{\int_{-\infty}^{+\infty} f_{x,y}(x,y) dy}_{f_x(x)} dx \\ &= \underline{\underline{\text{Var}(X)}} \end{aligned}$$

- $f(x,y) = x + y$

$$\begin{aligned} E(f(x,y)) &= E(x+y) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} (x+y) f_{x,y}(x,y) dy dx \\ &= \left(\int_{-\infty}^{+\infty} x f_{x,y}(x,y) dy dx \right) \overset{E(x)}{+} \left(\int_{-\infty}^{+\infty} y f_{x,y}(x,y) dy dx \right) \overset{E(y)}{=} \end{aligned}$$

$$E(x+y) = E(X) + E(Y)$$

- $f(x,y) = XY$

$$\begin{aligned} E(XY) &= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} xy f_{x,y}(x,y) dx dy \\ &\neq E(X) E(Y) \end{aligned}$$

ex

		X			$f_{x,y}(y)$
		1	2	3	
Y	5	0.2	0.1	0	0.3
	6	0.2	0.1	0.1	0.4
	7	0.1	0.1	0.1	0.3
		$f_x(x)$	0.5	0.3	0.2

$$\rho(x, y) = x + y$$

$$E(x+y) = \underbrace{E(x)}_{\sum x f_x(x)} + \underbrace{E(y)}_{\sum y f_y(y)}$$

$$E(x+y) = \underbrace{\sum_x \sum_y (x+y) \underbrace{f_{x,y}(x,y)}_{}}$$

Conditional expectation

$E[X|Y]$ } is a random variable
} is a function of Y $\rho(Y)$

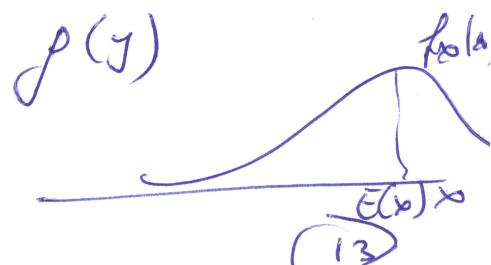
$$Y=y$$

$$\rho(y) = E[X|Y=y] = \begin{cases} \sum_i x_i P(X=x_i | Y=y) & \text{discrete} \\ \int_{-\infty}^{+\infty} x f_{x|y}(x|y) dx & \text{continuous} \end{cases}$$

$E[X|Y=y]$ is a function of y

$$\begin{array}{cccc} y & 0 & 1 & 2 \\ \hline E[X|Y=y] & E[X|Y=0] & E[X|Y=1] & E[X|Y=2] \end{array}$$

$E(X)$ is a number



• X, Y independent

$$\forall y \quad E(X|Y=y) = \int_{-\infty}^{+\infty} x \ f_{X|Y}(x|y) dx$$

$$\stackrel{\text{ind}}{=} \int_{-\infty}^{+\infty} x \ f_X(x) dx = \underline{\underline{E(X)}}$$

$$f_{X|Y} = f_X$$

• $E(X|Y=y) = g(y)$

$$E(g(Y)) = \int_{-\infty}^{+\infty} g(y) f_Y(y) dy$$

$$E_y(E(X|Y)) \stackrel{(a)}{=} \int_{-\infty}^{+\infty} g(y) f_Y(y) dy$$

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

$$g(y) = E(X|Y=y) = \boxed{\int_{-\infty}^{+\infty} x f_{X|Y}(x|y) dx}$$

$$(a) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} x \underbrace{f_{X|Y}(x|y) f_Y(y)}_{f_{X,Y}(x,y)} dx dy$$

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

$$= \int_{-\infty}^{+\infty} x \underbrace{\int_{-\infty}^{+\infty} f_{X|Y}(x|y) dy}_{f_X(x)} dx$$

$$= \int_{-\infty}^{+\infty} x f_X(x) dx = \underline{\underline{E(X)}}$$

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