

Joint PMF: $f(x,y) = P(X=x, Y=y)$

Marginal PMFs: $f_X(x) = \sum_y f(x,y)$

$$f_Y(y) = \sum_x f(x,y)$$

(*)

Ex. Flip 3 coins,

$$X = \begin{cases} 0 & \text{last flip is T} \\ 1 & \text{last flip is H} \end{cases}$$

$Y = \# \text{ heads among 3 flips}$

$f(x,y)$	0	1	2	3	
0	$f(0,0) = \frac{1}{8}$	$f(0,1) = \frac{3}{8}$	$\frac{1}{8}$	0	$f_X(0) = \frac{1}{2}$
1	0	$\frac{1}{8}$	$\frac{2}{8}$	$\frac{1}{8}$	$f_X(1) = \frac{1}{2}$
	$f_Y(0) = \frac{1}{8}$	$f_Y(1) = \frac{3}{8}$	$f_Y(2) = \frac{3}{8}$	$f_Y(3) = \frac{1}{8}$	

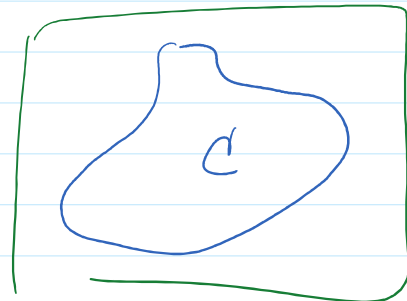
Defn: Joint PDF

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If X and Y are continuous we call

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}$$

the joint PDF if $\forall C \subset \mathbb{R}^2$



$$P((X, Y) \in C) = \int_C f(x, y) dx dy$$

Univariate case:

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

Facts: (1) $F(x, y) = \int_{-\infty}^x \int_{-\infty}^y f(u, v) du dv$

(uni: $F(x) = \int_{-\infty}^x f(t) dt$)

(2) $f(x, y) = \frac{\partial^2 F}{\partial x \partial y}$

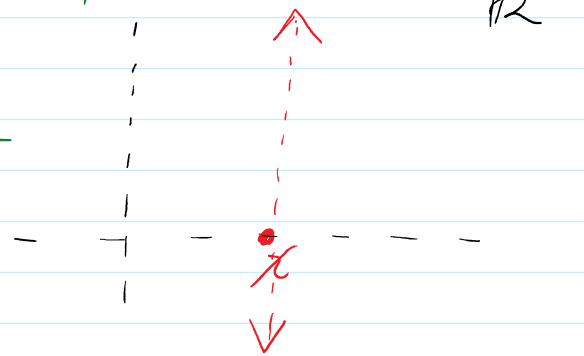
(uni: $f(x) = \frac{dF}{dx}$)

③ $f(x,y) \geq 0$ and $\iint_{\mathbb{R}^2} f(x,y) dx dy = 1$

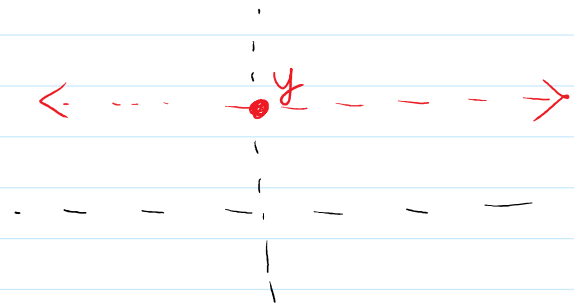
(Univ: $f(x) \geq 0$ and $\int_{\mathbb{R}} f(x) dx = 1$)

Theorem: Rel. btwn Joint/Marginals

① $f_x(x) = \int_{\mathbb{R}} f(x,y) dy$

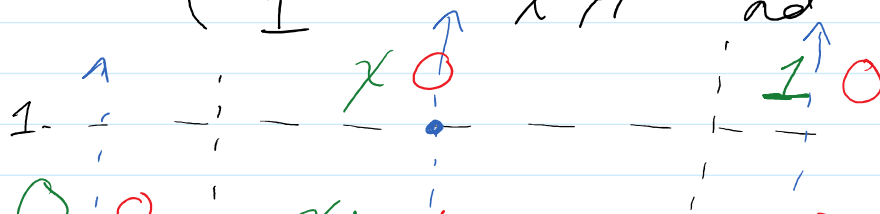


② $f_y(y) = \int_{\mathbb{R}} f(x,y) dx$



Ex.

$$F(x,y) = \begin{cases} 0 & x \leq 0 \text{ or } y \leq 0 \\ xy & (x,y) \in [0,1]^2 \\ x & y > 1 \text{ and } 0 < x < 1 \\ y & x > 1 \text{ and } 0 < y < 1 \\ 1 & x > 1 \text{ and } y > 1 \end{cases}$$



for $0 < y < 1$
 $y \sim U(0,1)$.

Ex, let

$$f(x,y) = 6xy^2 \quad \text{for } 0 < x < 1 \\ 0 < y < 1$$

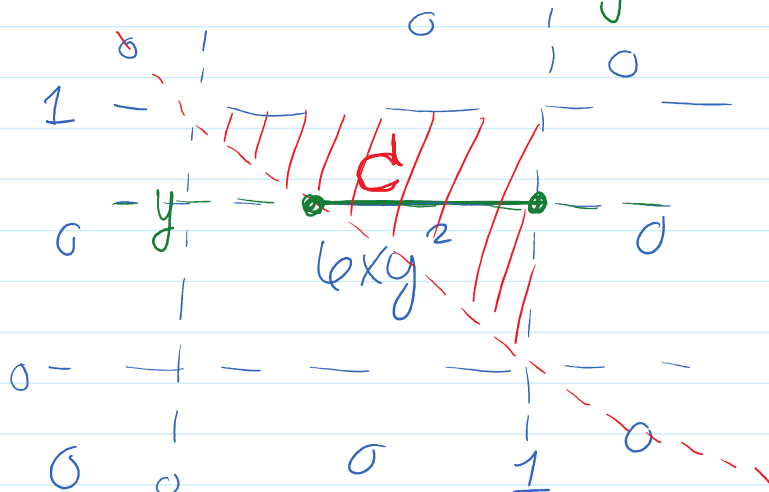
$$P(X+Y > 1)$$

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

$$= \int_C f(x,y) dx dy$$

$$= \int_0^1 \int_{1-y}^1 6xy^2 dx dy$$

$$= \dots = 9/10$$



$$x+y=1$$

$$x=1-y \quad \text{or} \quad y=1-x$$

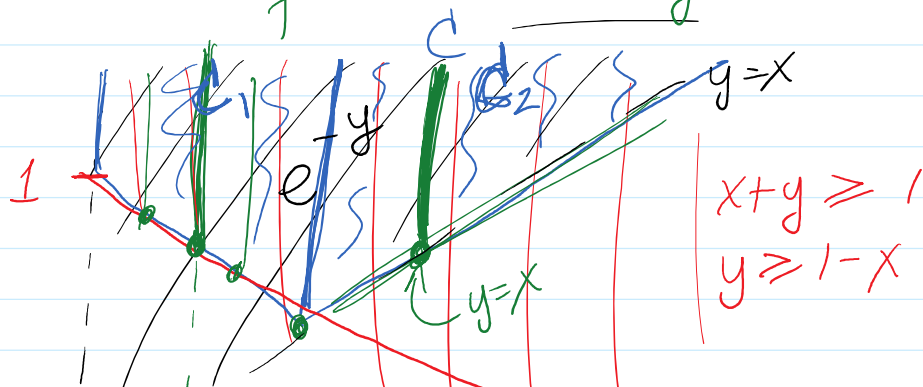
$$x+y > 1 \quad \text{then} \quad y > 1-x$$

Ex,

let $f(x,y) = e^{-y}$ for $0 < x < y$

$$P(X+Y \geq 1)$$

$$= \int f(x,y) dx dy$$



$$x+y \geq 1 \\ y \geq 1-x$$

$$\begin{aligned}
 &= \int_C f(x,y) dx dy \\
 &= \int_{C_1} f dx dy + \int_{C_2} f dx dy \\
 &= \int_0^{1/2} \int_{1-x}^{\infty} e^{-y} dy dx + \int_{1/2}^1 \int_x^{\infty} e^{-y} dy dy \\
 &= \dots = \boxed{\frac{2}{\sqrt{e}} - \frac{1}{e}}
 \end{aligned}$$

Defn: Bivariate Expectation

If (X,Y) is Biv RV and $g: \mathbb{R}^2 \rightarrow \mathbb{R}$
then

$$\mathbb{E}[g(X,Y)] = \begin{cases} \sum_x \sum_y g(x,y) f(x,y) & \text{(discrete)} \\ \iint_{\mathbb{R}^2} g(x,y) f(x,y) dx dy & \text{(continuous)} \end{cases}$$

PMF

(Univar: $\mathbb{E}[g(X)] = \int g(x) f(x) dx$)

$E_{X,Y}$
 $g(x,y) = xy$

let

$$f(x,y) = 1$$

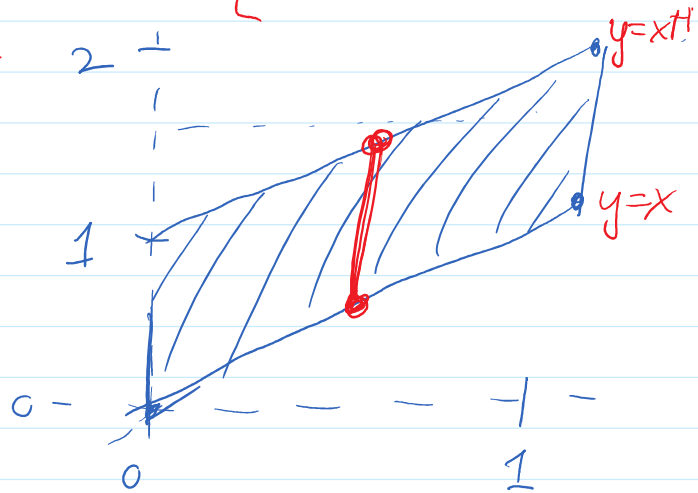
for

$$0 < x < 1$$

$$x < y < x+1$$

$$E[XY] = \int_0^1 \int_x^{x+1} xy(1) dy dx$$

$$= 7/12$$



Theorem: Biv Expectation is Linear

If $g_1: \mathbb{R}^2 \rightarrow \mathbb{R}$, $g_2: \mathbb{R}^2 \rightarrow \mathbb{R}$,
 and $a, b \in \mathbb{R}$, then

$$E[a g_1(X,Y) + b g_2(X,Y)]$$

$$= a E[g_1(X,Y)] + b E[g_2(X,Y)].$$