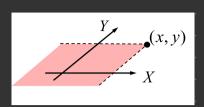


# B52 Nov 3 Lec 1 Notes

Joint CDF

For arbitrary RVs X, Y, joint CDF provides probabilities of type



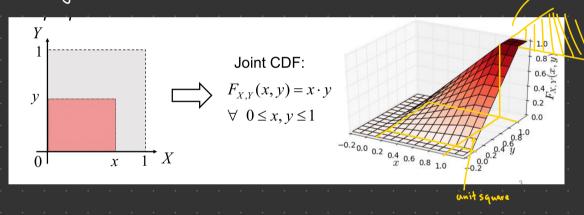
Joint CDF describes joint distributions of a RVs. We can use this to find marginal CDFs.

Ex 1: (2D Uniform)

Let X,Y be 20, uniform RVs over unit square

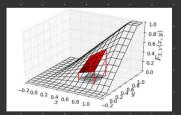
If x, y >1, Fx, x (x, y)=1

Probability of any subset  $A \subseteq [0,1]^2$  is proportional to area of A.



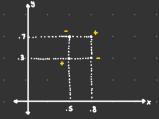
Ex 2: (Continued from Ex 1)

Find 
$$P(0.5 \le X \le 0.8, 0.3 \le Y \le 0.7)$$
  
=  $(0.3) \times (0.4)$   
=  $0.12$ 



GR.

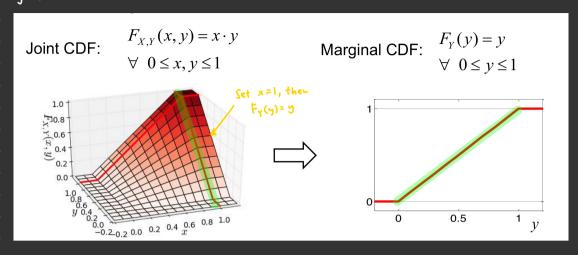
= 
$$F_{xy}$$
 (0.8,0.7) -  $F_{xy}$  (0.5,0.7)  
-  $F_{xy}$  (0.8,0.3) +  $F_{xy}$  (0.5,0.3)



= 0.12

## Ex. 3 ( (ontinued from Ex 1)

Find marginal CDF of Y.



### Joint CDF:

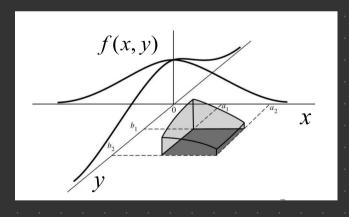
Joint CDF is useful for finding probabilities of rectangular areas of

Need different tool for calculating probabilities

joint PDF is a function fx, y (x,y)

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

We can think of probabilities as volume contained under function fx, and over



Properties of joint PDFs:

- (i) f<sub>x,x</sub> (x,y)≥0 . (ii) SS fx,x (x,y) dxdy = 1

- Relationship between joint PDF and joint CDF

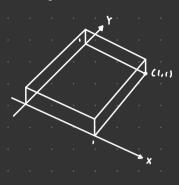
  (i)  $F_{x,Y}(x,y) = \int_{-\infty}^{x} \int_{-\infty}^{y} f_{x,Y}(s,t) dsdt$ ,  $\forall x,y \in \mathbb{R}$ (ii)  $f_{x,Y}(x,y) = \frac{\partial^{2} F_{x,Y}(x,y)}{\partial x \partial y}$ ,  $\forall x,y$  where derivatives expressions.

#### Ex 4 (2D Uniform)

Consider RVs X,Y with joint CDF Fx,x (x,y) = x,y , 0 ≤ x,y ≤ 1.

(i) Find their joint PDF:

$$f_{xy}(x,y) = \frac{\partial^2 F_{xy}(x,y)}{\partial x \partial y} = \frac{\partial}{\partial x} \left( \frac{\partial}{\partial y} x \cdot y \right)$$
$$= \frac{\partial}{\partial x} (x)$$
$$= 1 , \forall x,y \in (0,1)$$



(ii) Calculate P(X<Y)

$$P(X < Y) = \iint_{R} f_{xy}(x,y) dxdy$$

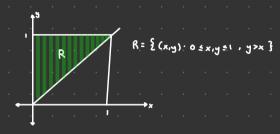
$$= \int_{0}^{1} \left[ \int_{x}^{1} f_{xy}(x,y) dy \right] dx$$

$$= \int_{0}^{1} \int_{x}^{1} 1 dy dx$$

$$= \int_{0}^{1} \left[ y \right]_{x}^{1} dx$$

 $= \int_0^1 (1-x) dx$ 

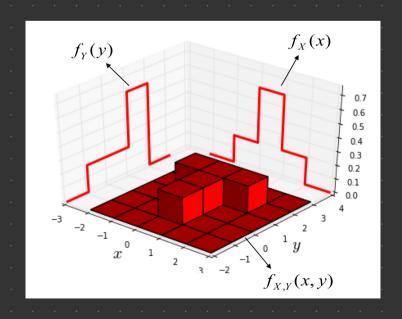
 $=1-\left[\frac{1}{2}x^{2}\right]_{0}^{1}=\frac{1}{2}$ 



# Marginal PDF

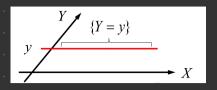
For RVs X,Y, with joint PDF fx,x, the marginal PDF of X and Y are given by

$$f_X(x) = \int_{-\infty}^{\infty} f_{x,Y}(x,y) dy$$
  
 $f_Y(y) = \int_{-\infty}^{\infty} f_{x,Y}(x,y) dx$ 



#### Conditional PDF

With continuous RVs X,Y, can condition on lower dimensional space with 0 probability e.g. condition on  $EY = y^3$ , which is a line.



Conditioning on EY= y3 restricts effective "sample space" from real plane to real line EY= y3.

 $\begin{cases} (x,Y) \text{ values live in real plane } \mathbb{R}^2 \\ (x,Y=y) \text{ values live in real plane } \mathbb{R} \end{cases}$ 

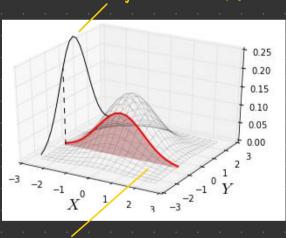
Let continuous RVs X,Y have joint PDF fx, (x,y).

Conditional PDF of X given Y=y defined as

$$f_{xir}(x|y) = \frac{f_{x,r}(x,y)}{f_{r}(y)}$$
, for  $f_{r}(y) > 0$ 

 $f_{x|Y}(x|y)$  cuts slice of  $f_{x|Y}(x,y)$  at Y=y and scales it by  $Y_{fY}(y)$  so that it integrates to 1.

marginal PDF of T, fy (y) = slice area



slice of joint PDF fx,y(x,y), along Y=<u>y</u>