

# **Probability and Statistics**

UG2, Core course, IIIT,H

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November 12, 2021

**1** Mixed Random Variable  
Solved Problems

- 2** Joint Distributions: Two Random Variables  
Computing probabilities with joint distributions  
Joint Cumulative Distribution function  
Marginal CDF  
Example of Joint PMF and Joint CDF  
Computing Probability of a Rectangular Patch  
Conditional PMF and Conditional CDF  
Independent Random Variables

Conditional Expectation  
Functions of Two Random Variables

- 3** Joint Continuous Random Variables  
Solved Problems

- 4** Multiple Random Variables  
Joint PDF and Joint CDF of Multiple Random Variables  
Sums of Random Variables  
Random Vectors  
Functions of Random Vectors and Method of Transformations

## Outline

- ① Mixed Random Variable  
Solved Problems
- ② Joint Distributions: Two Random Variables
- ③ Joint Continuous Random Variables
- ④ Multiple Random Variables

## CDF of a mixed RV as a sum of Continuous and Discrete CDF...

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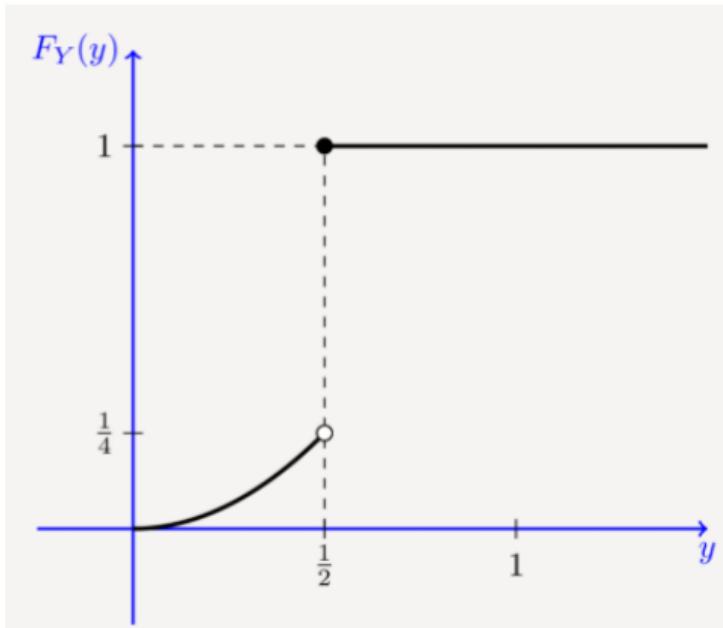
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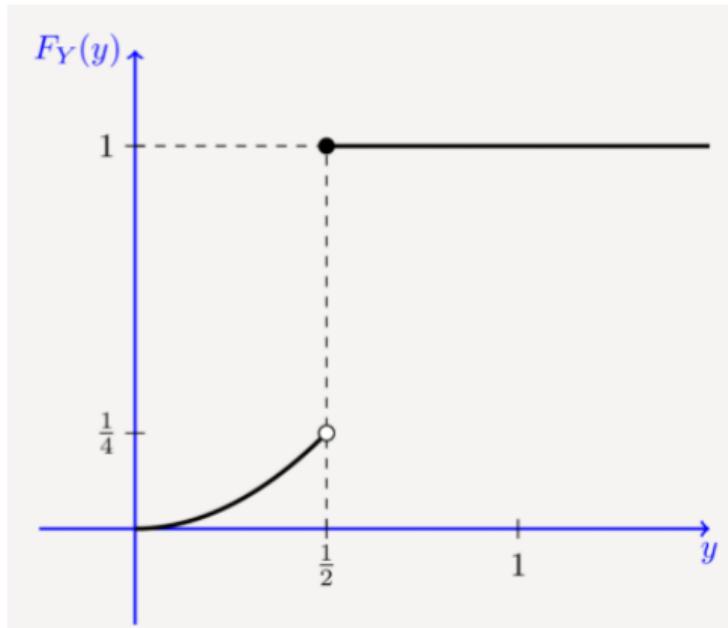
$$E[Y] = \int_{-\infty}^{\infty} y c(y) dy + \sum_{y_k} y_k P(Y = y_k)$$

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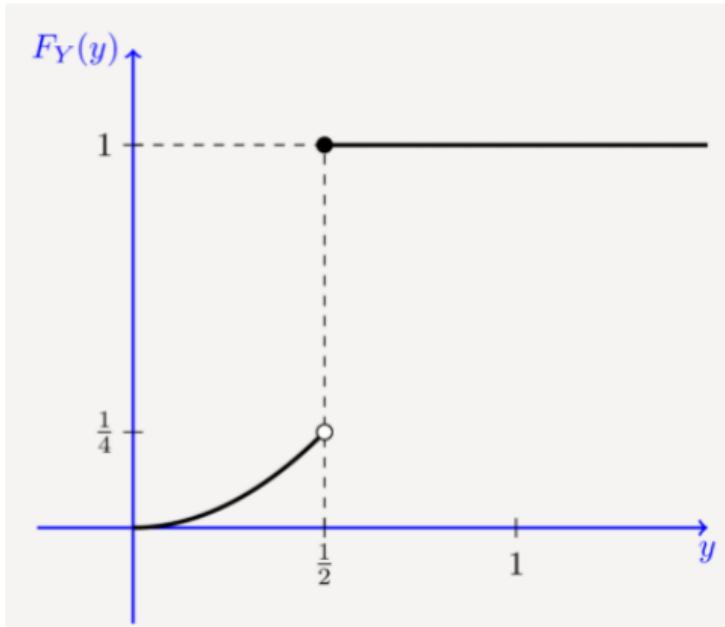


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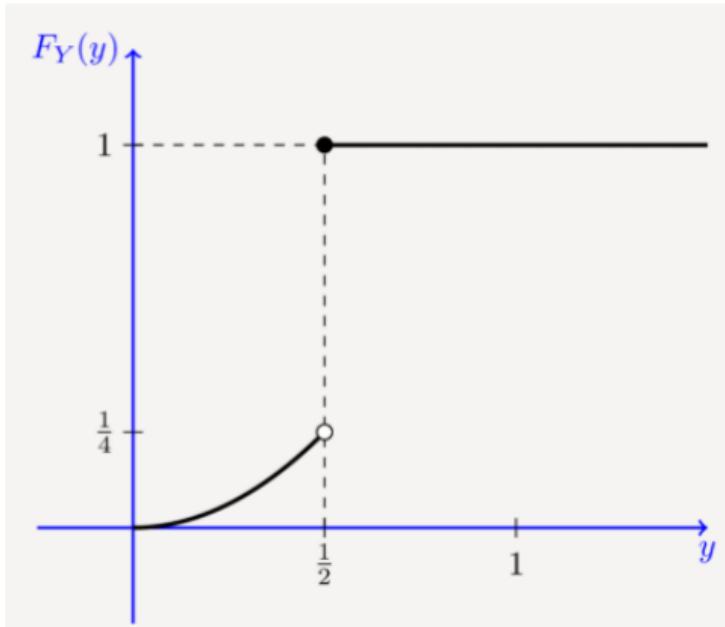
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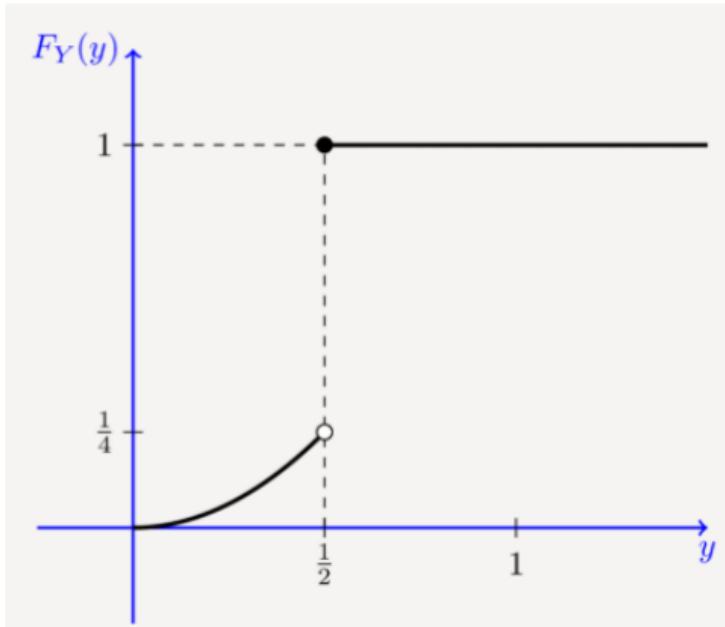


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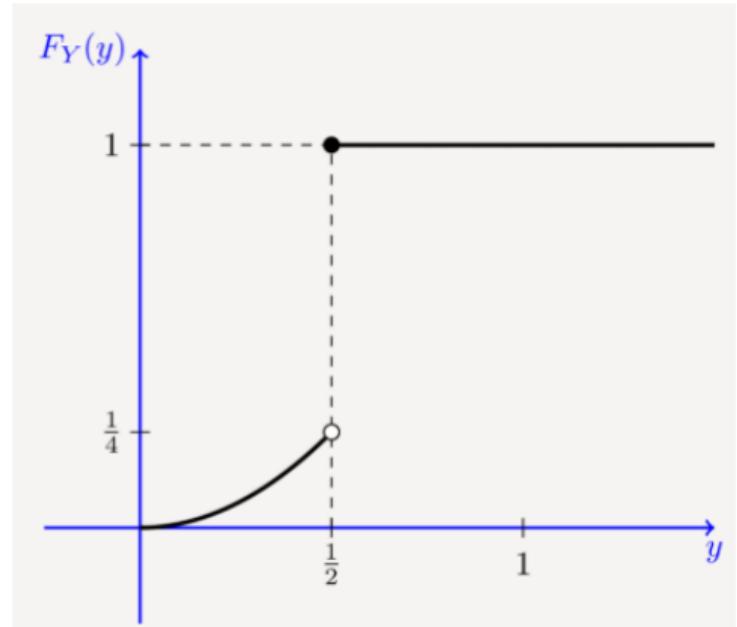
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$$F_Y(y) = \begin{cases} 0, & y < y_1 \\ y^2, & y_1 \leq y < y_2 \\ 1, & y \geq y_2 \end{cases}$$



Check that

$$\left\{ \int_{-\infty}^{\infty} c(y) dy + \sum_{y_k} P(Y = y_k) = 1 \right.$$

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and the discrete part is

$$\left\{ \begin{array}{ll} D(y) = & \text{RMF} \\ \begin{cases} 3/4 & y \geq 1/2 \\ 0 & y < 1/2 \end{cases} & \end{array} \right.$$

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$$P(a \leq X \leq b) = F_X(b) - F_X(a)$$

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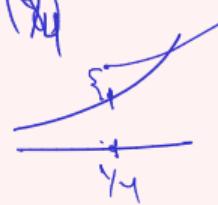
- Find  $P(1/4 \leq Y \leq 3/8)$
- Find  $P(Y \geq 1/4)$
- Find  $E[Y]$

$$\begin{aligned} 1 - P(Y < 1/4) &= 1 - F_Y(1/4) + P_Y(1/4) \\ &= 1 - \left(\frac{1}{4}\right)^2 + 0 \\ c(y) &= \frac{d}{dy} C(y) = \frac{d}{dy} y^2 = 2y \quad \begin{cases} 2y, & 0 \leq y \leq 1/2 \\ 0, & \text{otherwise.} \end{cases} \end{aligned}$$

If  $Y$  was cont.  
 $P(Y \leq y) = P(Y < y)$

$$F_Y(y) = P(Y \leq y)$$

$$F_Y(1/4) = P(Y \leq 1/4)$$



Answer to previous problem...

$$\textcircled{1} \quad P\left(\frac{1}{4} \leq Y \leq \frac{3}{8}\right) = F_Y\left(\frac{3}{8}\right) - F_Y\left(\frac{1}{4}\right)$$

$0 < \frac{3}{8} < \frac{1}{2}$

$$= \left(\frac{3}{8}\right)^2 - \left(\frac{1}{4}\right)^2 + 0$$

= ~~—~~ ~~A&B~~

\textcircled{2} done

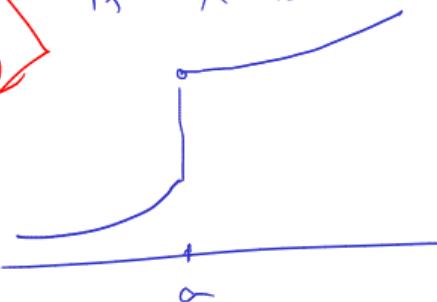
$$\textcircled{3} \quad E[Y] = \int_0^{Y_2} y \cdot 2y \, dy + \frac{1}{2} \cdot P(Y=Y_2)$$

*jump at  $y=Y_2$*

$$= 2 \left[ \frac{y^3}{3} \right]_0^{Y_2} + \frac{1}{2} \cdot \frac{3}{4} = \frac{2}{3} \cdot \left(\frac{1}{2}\right)^3 + \frac{1}{2} \cdot \frac{3}{4}$$

$$\begin{aligned} P(a \leq X \leq b) &= P(a < X < b) \\ &= P(a < X < b) \\ &= P(a \leq X < b) \\ &= P(a < X \leq b) \end{aligned}$$

if  $X$  is continuous.



$$\left(\frac{1}{2}\right)^3 + \frac{1}{2} \cdot \frac{3}{4}$$

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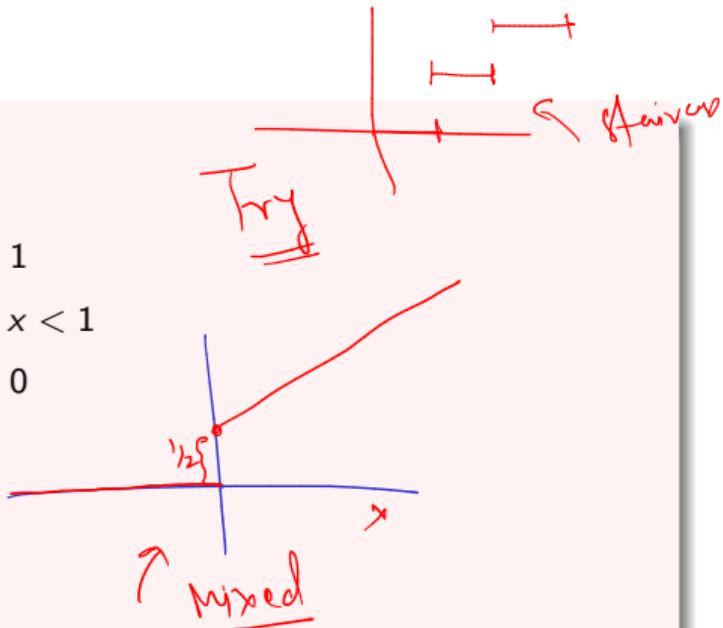
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- 4 Find  $P(X = 0 | X \leq 0.5)$



**Answer to previous problem...**

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## Solved Problem 2

### Problem 2

Let  $X \sim \text{Uniform}(-2, 2)$  be a continuous random variable. Let  $Y = g(X)$  where

PDF =

$$g(x) = \begin{cases} 1 & x > 1 \\ x & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

Find the CDF of  $Y$ .

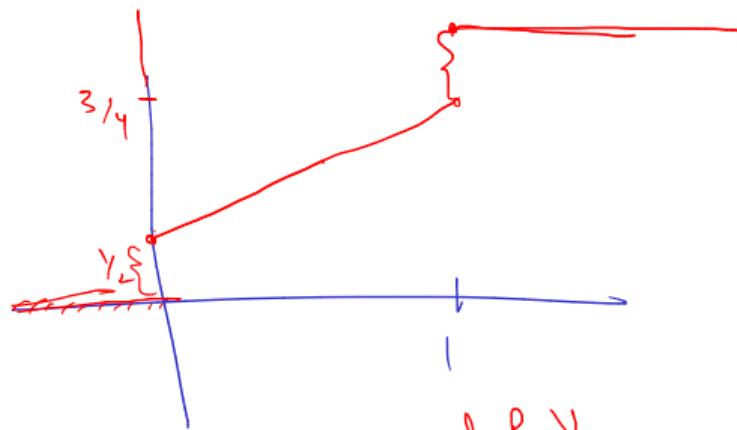
$$\boxed{R_Y = [0, 1]} \Rightarrow F_Y(0) = 0, \quad y < 0 \quad | \quad \text{if } 0 < y \leq 1 \quad \text{corresp.}$$
$$\Rightarrow F_Y(1) = 1, \quad y > 1$$

$$\text{For } 0 < y < 1, \quad F_Y(y) = P(Y \leq y) = P(X \leq y) = F_X(y)$$
$$= \int_{-2}^y \frac{1}{4} dx = \frac{1}{4} [x]_{-2}^y = \frac{y+2}{4}$$

Answer to previous problem...

Summary

$$F_y(y) = \begin{cases} \frac{1}{4}(y+2), & 0 \leq y \leq 1 \\ 1, & y > 1 \\ 0, & \text{otherwise} \end{cases}$$



Example of mixed R.V

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$$R_{XY} = \{(x, y) \mid P_{XY} > 0\}$$

$$(x_5, y_{10}) = 0$$

- In particular, if  $R_X = \{x_1, x_2, \dots\}$ ,  $R_Y = \{y_1, y_2, \dots\}$ , then

$$R_{XY} \subset R_X \times R_Y = \{(x_i, y_j) \mid x_i \in R_X, y_j \in R_Y\}$$

- Sum of joint probabilities must sum to 1:  $\sum_{(x_i, y_j) \in R_{XY}} P_{XY}(x_i, y_j) = 1$

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$P_{XY}$

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## Solved Example

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Let  $X$  and  $Y$  be two random variables with joint PMF as follows:

	$Y=0$	$Y = 1$	$Y = 2$
$X = 0$	$\frac{1}{6}$	$\frac{1}{4}$	$\frac{1}{8}$
$X = 1$	$\frac{1}{8}$	$\frac{1}{6}$	$\frac{1}{6}$

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- 1 Find  $P(X = 0, Y \leq 1)$
- 2 Find the marginal PMFs of  $X$  and  $Y$

## Solved Example

$$P_Y(1) = \sum_{i=1}^2 P_{XY}(x_i, 1)$$

$$= P(0, 1) + P(1, 1)$$

$$= \cancel{Y_6} + \cancel{Y_6}$$

$$\begin{aligned} &= \cancel{\cancel{\cancel{P(Y=1, X=0)}}} \\ &\quad \rightarrow P(X=0) \\ &= Y_4 \end{aligned}$$

Let  $X$  and  $Y$  be two random variables with joint PMF as follows:

	$Y=0$	$Y=1$	$Y=2$
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1 Find  $P(X = 0, Y \leq 1)$

2 Find the marginal PMFs of  $X$  and  $Y$

3 Find  $P(Y = 1 | X = 0)$

$$\begin{aligned} P_X(1) &= P_{XY}(1, 0) + P_{XY}(1, 1) \\ &\quad + P_{XY}(1, 2) \\ &= \text{read from table} \end{aligned}$$

$$\begin{aligned} &= P_{XY}(0, 0) + P_{XY}(0, 1) + P_{XY}(0, 2) \\ &= \boxed{Y_6 + Y_4 + Y_8} \\ &= \frac{Y_6 + Y_8 + Y_8}{Y_6 + Y_8 + Y_8} \end{aligned}$$

## Solved Example

$$P(A \cap B) = P(A) P(B)$$
$$P(x=\alpha, y=\gamma) = P(x=\alpha) P(y=\gamma)$$

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Let  $X$  and  $Y$  be two random variables with joint PMF as follows:

	$Y=0$	$Y=1$	$Y=2$
$X=0$	$\frac{1}{6}$	$\frac{1}{4}$	$\frac{1}{8}$
$X=1$	$\frac{1}{8}$	$\frac{1}{6}$	$\frac{1}{6}$

- 1 Find  $P(X = 0, Y \leq 1)$
- 2 Find the marginal PMFs of  $X$  and  $Y$
- 3 Find  $P(Y = 1 | X = 0)$
- 4 Are  $X$  and  $Y$  independent?

Answer to previous problem...



Answer to previous problem...



## Joint Cumulative Distribution Function...

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*empty set*

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## Example of Joint PMF and Joint CDF...

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Solved Example on Joint PDF and Joint CDF

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Solved Example on Joint PDF and Joint CDF

Let  $X \sim \text{Bernoulli}(\underline{p})$  and  $Y \sim \text{Bernoulli}(\underline{q})$  be independent, where  $0 < \underline{p}, \underline{q} < 1$ .

## Example of Joint PMF and Joint CDF...

$p = \text{prob. of Head}$

### Solved Example on Joint PDF and Joint CDF

Let  $X \sim \text{Bernoulli}(p)$  and  $Y \sim \text{Bernoulli}(q)$  be independent, where  $0 < p, q < 1$ . Find the joint PMF and joint CDF for  $X$  and  $Y$ .

$$X, Y = \{(0,0), (0,1), (1,0), (1,1)\}, \quad 0 = \text{Head}, 1 = \text{Tail}.$$

$$P_{XY}(i,j) = P_X(i) P_Y(j) \quad [X, Y \text{ ind.}], \quad i, j = 0, 1$$

$$P_{XY}(0,0) = P_X(0) P_Y(0) = p \cdot q$$

$$P_{XY}(0,1) = P_X(0) P_Y(1) = p \cdot (1-q)$$

$$P_{XY}(1,0) = P_X(1) P_Y(0) = (1-p)q$$

$$P_{XY}(1,1) = P_X(1) P_Y(1) = (1-p)(1-q)$$

Joint CDF

$$F_{XY}(x,y) = P_{XY}(X \leq x, Y \leq y)$$

$$\begin{cases} ① F_{XY}(x,y) = 0, & \text{if } x < 0 \\ ② F_{XY}(x,y) = 0, & \text{if } y < 0 \\ ③ F_{XY}(x,y) = 1, & \text{if } x \geq 1, y \geq 1 \end{cases}$$

Answer to previous problem...

$$\text{For } \begin{cases} 0 \leq x < 1, y > 1 \end{cases}$$

$$F_{XY}(x, y) = P_{XY}(X \leq x, Y \leq y)$$

$$= P_{XY}(X = 0, Y \leq 1)$$

$$= P_X(X = 0) = p$$

$$\text{For } \begin{cases} 0 \leq y < 1, X > 1 \end{cases}$$

$$F_{XY}(x, y) = P_{XY}(X \leq 1, Y = 0)$$

$$= P_Y(Y = 0) = q$$

~~Finally for~~

$$\begin{cases} 0 \leq x < 1 \text{ and } 0 \leq y < 1 \end{cases}$$

$$F_{XY}(x, y) = P(0, 0)$$

$$= p \{ \cdot \}$$

Answer to previous problem...



## Plot of Joint CDF

- Figure shows the values of  $F_{XY}(x, y)$  in different regions

## Plot of Joint CDF

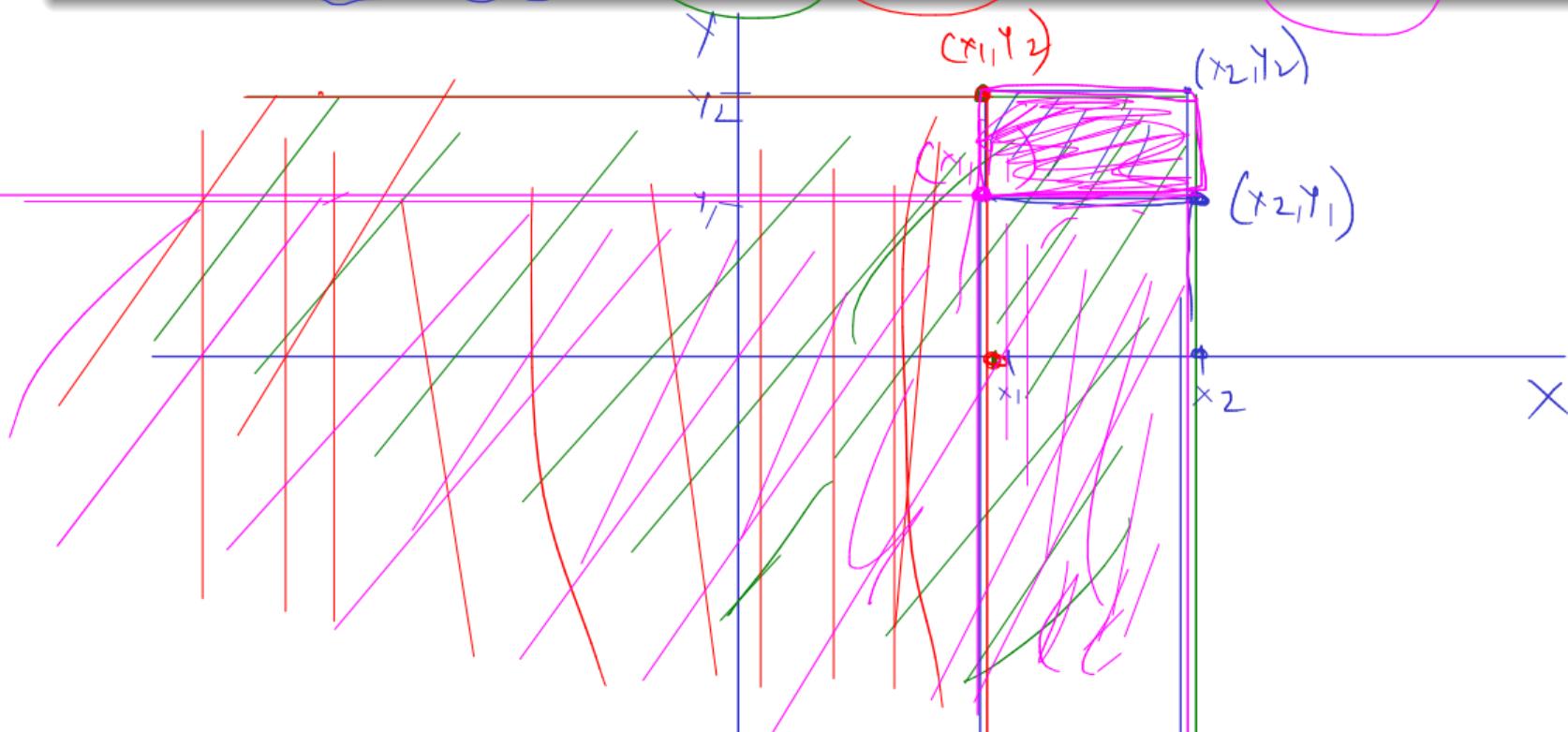


- Figure shows the values of  $F_{XY}(x, y)$  in different regions
- Note that in general we need three dimensional graph to show a joint CDF of two random variables

A result

$$P(a < X \leq b) = F_X(b) - F_X(a)$$

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



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Example Motivation for Conditional PMF and CDF

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I roll a fair die.

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I roll a fair die. Let  $X$  be the observed number.

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### Example Motivation for Conditional PMF and CDF

I roll a fair die. Let  $X$  be the observed number. Find the conditional PMF of  $X$  given that we know the observed number was less than 5.

Solution:

## Conditional PMF and Conditional CDF...

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Definition of Conditional PMF and Conditional CDF

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$$\begin{aligned} P_{X|A}(x_i) &= \underbrace{P(X = x_i | A)}_{\text{conditional PMF}} \\ &= \frac{P(X = x_i \text{ and } A)}{P(A)}, \quad \text{for any } x_i \in R_X \end{aligned}$$

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Conditional PMF of  $X$  given  $Y \dots$

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for **any**  $x_i \in R_X$  and  $y_j \in R_Y$ .

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