

Probability and Statistics: Lecture-28

Monsoon-2020

by Dr. Pawan Kumar (IIIT, Hyderabad)

on October 16, 2020

» Joint Cumulative Distribution Function...

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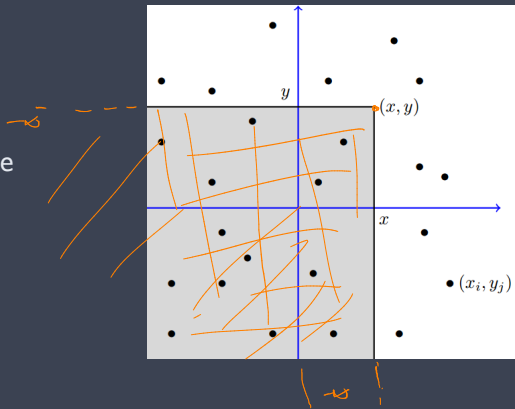
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Sample
Space

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Moreover, we have the following **properties**

$$F_{XY}(\infty, \infty) = 1, \quad F_{XY}(-\infty, y) = 0,$$

$P(\underbrace{\{X \leq -\infty\}}_{\emptyset} \cap \{Y \in \mathcal{Y}\})$
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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 $X \sim \text{Bernoulli}(p)$ and $Y \sim \text{Bernoulli}(q)$ be independent, where $0 < p, q < 1$.

» Example of Joint PMF and Joint CDF...

$$P_{X,Y}(x=y, y=x) = -$$
$$X = \{0, 1\}, Y = \{0, 1\}$$

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 .

» Answer to previous problem...

The joint range of R.V.s X and Y is given by

$$R_{XY} = \{(0,0), (0,1), (1,0), (1,1)\}$$

[$\underline{0} \equiv \text{Head}$, $\underline{0} \equiv \text{Tail}$]

$$P_{XY}(i,j) = \underbrace{P_X(i) P_Y(j)}_{\text{Since } X, Y \text{ are ind.}} \quad \text{for } i, j = 0, 1$$

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

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

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

$$P_{XY}(1,1) = pq$$

Joint CDF

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

$$1. F_{XY}(x,y) = 0 \quad \text{if } x < 0$$

$$2. F_{XY}(x,y) = 0 \quad \text{if } y < 0$$

$$3. F_{XY}(x,y) = 1 \quad \text{if } \underline{x \geq 1} \text{ \& } \underline{y \geq 1}$$

$$\underline{0 \leq x < 1 \text{ and } y \geq 1}$$

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

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

$$= P(X=0) \quad \begin{matrix} \uparrow \\ \text{added} \\ \text{all } y \end{matrix}$$

$$= 1-p \leftarrow \text{marginal CDF w.r.t } X$$

» Answer to previous problem...

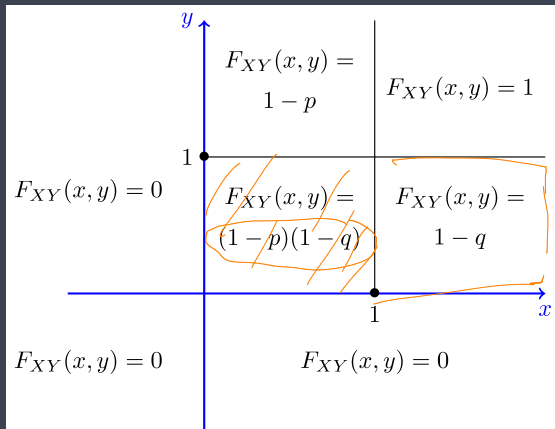
For $0 \leq y < 1$ and $x \geq 1$

$$F_{X,Y}(x,y) = 1 - q \quad (\text{check as for } x)$$

Finally for $0 \leq x < 1$ and $0 \leq y < 1$

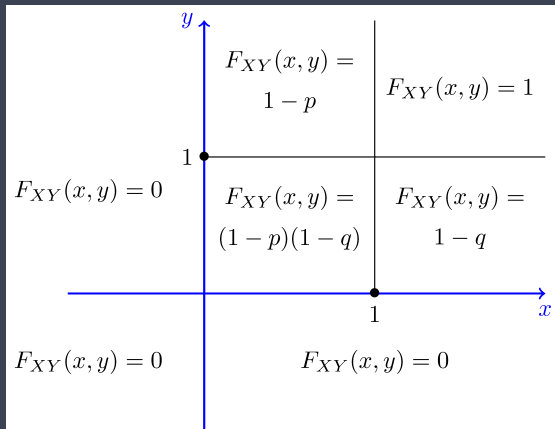
$$\begin{aligned} F_{X,Y}(x,y) &= P(X \leq x, Y \leq y) \\ &= P(X=0, Y=0) \\ &= \underline{\underline{(1-p)(1-q)}} \end{aligned}$$

» Plot of Joint CDF



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

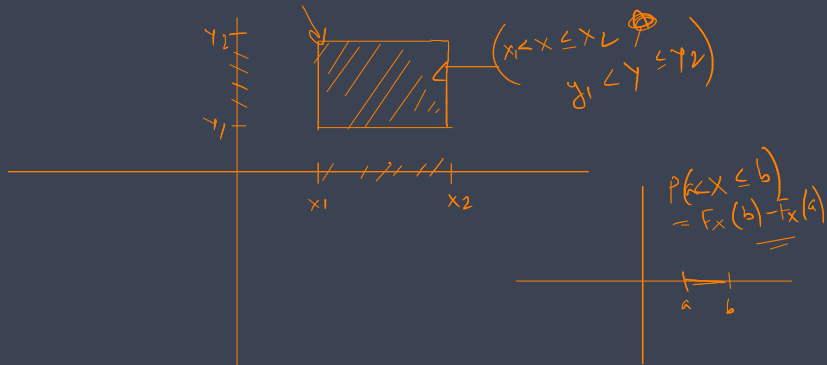
» Plot of Joint CDF



- * 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

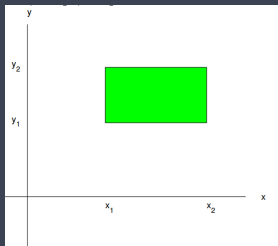
A result

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



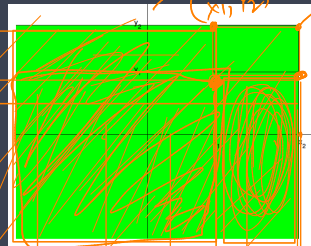
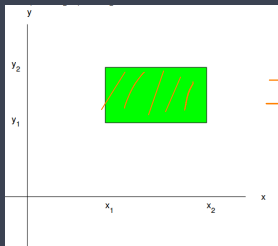
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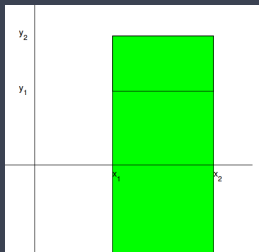
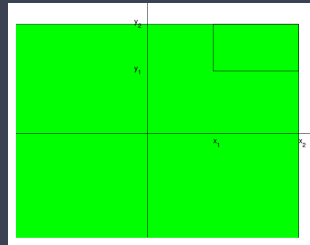
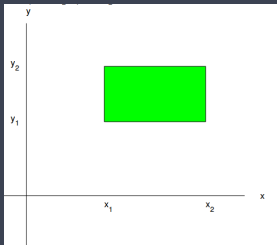


$F_{XY}(x_2, y_2)$
 (x_2, y_1)

$P(X \leq x_2, Y \leq y_2)$

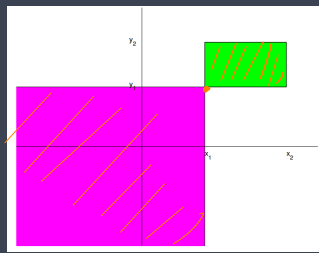
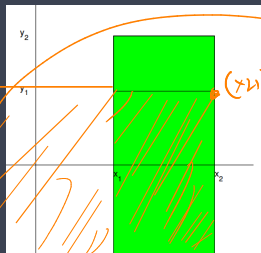
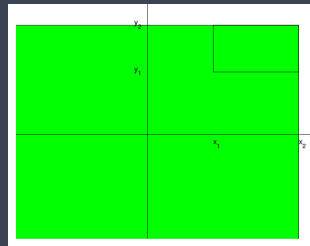
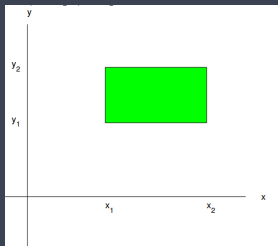
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Example Motivation for Conditional PMF and CDF

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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:

$$R_X = \{1, 2, 3, 4, 5, 6\} \quad A = \{x < 5\}$$

$$P_{X|A}(1) = P(X=1 \mid \underbrace{X < 5}_A) = \frac{P(X=1 \text{ and } \underbrace{X < 5})}{P(X < 5)} = \frac{P(X=1)}{P(X < 5)} = \frac{1}{4}.$$

Similarly

$$P_{X|A}(2) = P_{X|A}(3) = P_{X|A}(4) = \frac{1}{4}$$

$$P_{X|A}(5) = P_{X|A}(6) = \underline{\underline{0}}$$

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$$P_{X|Y}(x_i, y_j) = \frac{P_{XY}(x_i, y_j)}{P_Y(y_j)}$$

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

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$$\underline{P_{X|Y}(x_i|y_j)} = \underline{P(X = x_i | Y = y_j)} = \frac{P_{XY}(x_i, y_j)}{P_Y(y_j)} = \frac{P_X(x_i)P_Y(y_j)}{P_Y(y_j)} = P_X(x_i)$$

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Consider the set of points in set G defined as follows

$$G = \{(x, y) \mid x, y \in \mathbb{Z}, \quad |x| + |y| \leq 2\}.$$

Handwritten notes:

$$x = \{0, 1, -1, 2, -2\}$$
$$y = \{0, 1, -1, 2, -2\}$$
$$G = \{(0,0), (0,1), (0,-1), (1,0), (-1,0), (1,1), (1,-1), (-1,1), (-1,-1), (2,0), (-2,0), (0,2), (0,-2), (1,2), (-1,2), (2,1), (-2,1), (1,-2), (-1,-2), (2,-1), (-2,-1)\}$$

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1. Find the **joint and marginal PMFs** of X and Y .
2. Find the **conditional PMF** of X given $Y = 1$.

» Example of Joint and Marginal PDF and Conditional PMF...

$$\checkmark \underline{P_{XY}(x=x_i, y=y_i)} = \underline{P_X(x_i)} = \sum_{y_i} P_{XY}(x_i, y_i)$$

Example

Consider the set of points in set G defined as follows

$$G = \{(x, y) \mid x, y \in \mathbb{Z}, \quad |x| + |y| \leq 2\}.$$

If we pick a point (X, Y) from this grid at random, then the probability of choosing a point is $1/13$.

1. Find the joint and marginal PMFs of X and Y .
2. Find the conditional PMF of X given $Y = 1$.
3. Are X and Y independent?

$$P_{X|Y}(x=x_i | y=1) = \frac{P_{XY}(x_i, 1)}{P_Y(1)} = \underline{P_X(x_i)}$$

» Answer to previous problem...

① Joint PMF

$$P_{XY}(x,y) = \begin{cases} \frac{1}{13} & (x,y) \in G \\ 0 & \text{otherwise} \end{cases}$$

$$P_Y(y) = \sum_{x_i \in R_X} P_{XY}(x_i, y)$$

$$P_X(x) = \sum_{y_i \in R_Y} P_{XY}(x, y_i)$$

$$P_X(\underline{-2}) = P_{XY}(\underline{-2}, 0) = \frac{1}{13}$$

$$P_X(-1) = P_{XY}(-1, -1) + P_{XY}(-1, 1) + P_{XY}(-1, 1) \\ = \frac{3}{13}$$

$$P_X(0) = P_{XY}(0, -2) + P_{XY}(0, -1) + P_{XY}(0, 0) + P_{XY}(0, 1) + P_{XY}(0, 2)$$

$$P_X(i) = \begin{cases} \frac{5}{13} & \text{for } i = 2, -2 \\ \frac{1}{13} & \text{for } i = -1, 1 \\ \frac{3}{13} & \text{for } i = 0 \\ \frac{5}{13} & \text{for } i = 0 \\ 0 & \text{otherwise} \end{cases}$$

» Answer to previous problem...

② $y=1$ possible $x: -1, 0, 1$

$$P_{X|Y}(i, 1) = \frac{P_{X,Y}(i, 1)}{P_Y(1)} = \frac{Y_{13}}{3Y_{13}} = \frac{1}{3}$$

for $-1, 0, 1$

$$P_{X|Y}(i|1) = \begin{cases} \frac{1}{3} & \text{for } i = -1, 0, 1 \\ 0 & \text{otherwise} \end{cases}$$

$\Rightarrow \underbrace{X|Y=1}_{\text{over the set } \{-1, 0, 1\}}$ is uniformly disto.

③ Is X, Y ind.

$$\text{Is } P_{X|Y}(i, 1) \neq P_X(i)$$

$\forall i$

$\Rightarrow X$ & Y are not ind.

» Definition of Conditional Expectation...

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Definition of Conditional Expectation

Let A be any event.

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Let A be any event. Let X and Y be two random variables with ranges R_X and R_Y respectively.

» Definition of Conditional Expectation...

$$E(X|Y) = \sum_i x_i \underbrace{P(X=x_i|Y)}$$

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Let A be any event. Let X and Y be two random variables with ranges R_X and R_Y respectively. Then the **conditional expectations** are defined as follows

$$E[X | A] = \sum_{x_i \in R_X} x_i P_{X|A}(x_i)$$

$$E[X | Y = j_j] = \sum_{x_i \in R_X} x_i P_{X|Y}(x_i | y_j)$$

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» Example of Conditional Expectation

$$P_{X|Y} = \begin{cases} \frac{1}{3} & x = -1, 0, 1 \\ 0 & \text{otherwise} \end{cases}$$
$$E[X|Y=1] = (-1) \frac{1}{3} + 0 \cdot \frac{1}{3} + 1 \cdot \frac{1}{3}$$

Example

Consider the set of points in set G defined as follows

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1. Find $E[X \mid Y = 1]$

» Example of Conditional Expectation

$$P_{X|Y}(x_i | -1 < Y < 2)$$

$$= \frac{\sum x_i P_{X,Y}}{\sum P_{X,Y}}$$

Example

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1. Find $E[X \mid Y = 1]$
2. Find $E[X \mid -1 < Y < 2]$

» Example of Conditional Expectation

$$\sum_i p_{x|y}$$

Example

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If we pick a point (X, Y) from this grid at random, then the probability of choosing a point is $1/13$.

1. Find $E[X \mid Y = 1]$
2. Find $E[X \mid -1 < Y < 2]$
3. Find $E[\underbrace{|X|} \mid -1 < Y < 2]$