EE901 PROBABILITY AND RANDOM PROCESSES

Module 6
Multiple Random
Variables

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Joint Probability Density Function

• If there exists a function $f_{X,Y}(x,y)$ such that

$$F_{X,Y}(x,y) = \int_{-\infty}^{x} \int_{-\infty}^{y} f_{X,Y}(x,y) dxdy$$

it is known as the joint probability density function of X, Y.

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

Example: Dart Throw

- $\Omega = B$. Each outcome ω is a 2D coordinate (x, y).
- Assume a uniform probability measure $\mathbb{P}(A) = |A|/a^2$ for any set.
- Let $X(\omega)$ and $Y(\omega)$ denote the x and y coordinate of the outcome.



• Joint CDF $F_{X,Y}(x,y) = \mathbb{P}(X \le x, Y \le y)$

$$F_{X,Y}(x,y) = \begin{cases} 0 & \text{if } x < 0 \text{ OR } y < 0 \\ 1 & \text{if } x > a \text{ AND } y > a \\ \frac{1}{a^2} \min(x,a) \min(y,a) & \text{otherwise} \end{cases}$$

• Joint PDF

$$f_{X,Y}(x,y) = \begin{cases} \frac{1}{a^2} & \text{if } a \ge x, y \ge 0\\ 0 & \text{otherwise} \end{cases}$$

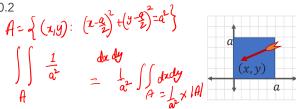
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Probability Law in terms of PDF

• For any 2D Borel set A

$$\mathbb{P}[(X,Y) \in A] = \int_A f_{X,Y}(x,y) dx dy$$

- What is probability that the dart hits within 0.2 distance of the center?
- A=Circle of radius 0.2



Marginal Distribution

- The individual distribution of X and Y are called marginal distribution.
- It is possible to compute the marginal distribution from the joint distribution.
- Let us first compute marginal CDFs of X and Y

$$\begin{split} F_{X,Y}(x,y) &= \mathbb{P}(X \leq x, Y \leq y) \\ F_{X,Y}(x,\infty) &= \mathbb{P}(X \leq x, Y \leq \infty) \\ &= \mathbb{P}(X \leq x) = F_X(x) \\ F_{X,Y}(x,y) &= \begin{cases} 0 & \text{if } x < 0 \text{ OR } y < 0 \\ 1 & \text{if } x > a \text{ AND } y > a \\ \frac{1}{a^2} \min(x,a) \min(y,a) & \text{otherwise} \end{cases} \\ F_{X}(x) &= F_{X,Y}(x,\infty) = \begin{cases} 0 & \text{if } x < 0 \\ 1 & \text{if } x > a \\ x/a & \text{otherwise} \end{cases} \end{split}$$

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Marginal PMFs

• Let us first compute marginal PMFs of X and Y

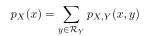
$$p_{X,Y}(x,y) = \mathbb{P}(X = x, Y = y)$$

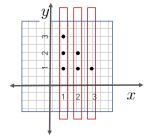
$$\sum_{y \in \mathcal{R}_Y} p_{X,Y}(x,y) = \sum_{y \in \mathcal{R}_Y} \mathbb{P}(X = x, Y = y) = \mathbb{P}(\bigcup_{y \in \mathcal{R}_Y} \{X = x, Y = y\})$$

$$= \mathbb{P}(\{X = x, Y \in \mathcal{R}_Y\})$$

$$= \mathbb{P}(\{X = x\}) = p_X(x)$$

Marginal PMFs





Consider the PMF. Each point is equi-probable.

$$p_X(x) = p_{X,Y}(x,1) + p_{X,Y}(x,2) + p_{X,Y}(x,3)$$

$$p_X(1) = p_{X,Y}(1,1) + p_{X,Y}(1,2) + p_{X,Y}(1,3) = \frac{3}{6} = 0.5$$
$$p_X(2) = p_{X,Y}(2,1) + p_{X,Y}(2,2) + p_{X,Y}(2,3) = \frac{2}{6} = 0.33$$

$$p_X(2) = p_{X,Y}(2,1) + p_{X,Y}(2,2) + p_{X,Y}(2,3) = \frac{2}{6} = 0.33$$

$$p_X(3) = p_{X,Y}(3,1) + p_{X,Y}(3,2) + p_{X,Y}(3,3) = \frac{1}{6} = 0.17$$

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Marginal PDFs

• Let us first compute marginal PDFs of X and Y

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

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

$$F_X(x) = \int_{-\infty}^x f_X(x) dx$$

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

$$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy$$
 $f_X(x) = \frac{d}{dx} F_{X,Y}(x,\infty)$

Marginal PDFs

• Marginal PDFs of X and Y $f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) \mathrm{d}y$ $f_X(x) = \frac{\mathrm{d}}{\mathrm{d}x} F_{X,Y}(x,\infty)$

Example
$$f_{X,Y}(x,y) = \begin{cases} \frac{1}{a^2} & \text{if } a \ge x, y \ge 0 \\ 0 & \text{otherwise} \end{cases}$$

otherwise
$$F_X(x) = F_{X,Y}(x,\infty) = \begin{cases} 0 & \text{if } x < 0 \\ 1 & \text{if } x > a \\ x/a & \text{otherwise} \end{cases}$$

$$f_X(x) = \int_0^a \frac{1}{a^2} dy = \frac{1}{a}$$

otherwise $f_X(x) = \int_0^a 0 dy = 0$

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Independence of RVs

• Random variables X and Y are mutually independent if

$$\mathbb{P}\left(X\in B_1,Y\in B_2\right)=\mathbb{P}\left(X\in B_1\right)\mathbb{P}\left(Y\in B_2\right)$$
 for any sets B_1 and B_2

• If $B_1 = (-\infty, x]$ and $B_2 = (-\infty, y]$,

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

$$P(X \le x, Y \le y) = P(X \le x) P(Y \le y).$$

Independence of RVs

• Random variables X and Y are mutually independent if

for any sets B_1 and B_2

• For DRVs, let us take $B_1=\{x\}, B_2=\{y\}$

$$\mathbb{P}(X = x, Y = y) = \mathbb{P}(X = x)\mathbb{P}(Y = y)$$
$$p_{X,Y}(x, y) = p_X(x) p_Y(y)$$

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Independence of RVs

• Random variables X and Y are mutually independent if

$$\mathbb{P}\left(X\in B_1,Y\in B_2\right)=\mathbb{P}\left(X\in B_1\right)\mathbb{P}\left(Y\in B_2\right)$$
 for any sets B_1 and B_2

• For CRVs, let us take $B_1 = (x, x + \epsilon)$ and $B_2 = (y, y + \epsilon)$

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

Functions of Multiple Random Variables

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Function of Two RVs

- Let X_1 and X_2 have joint PDF $f_{X_1,X_2}(x_1,x_2)$.
- Let Z be a function of X_1 and X_2

• Z is also a random variable.

Expectation of A Function of Two RVs

- Let X and Y be two random variables.
- Consider a function $g: \mathcal{R}(X) \times \mathcal{R}(Y) \to \mathbb{R}$. $\mathrm{E}_{\underline{X},\underline{Y}}[g(X,Y)] = \int \int \underline{g(x,y)} f_{X,Y}(x,y) \mathrm{d}x \mathrm{d}y \qquad \sum_{\substack{X,Y \\ Y,Y}} g(X,Y) \underbrace{f_{X,Y}(x,y)}_{X,Y} = \underbrace{f_{X}(x)} \underbrace{f_{Y}(y)}_{Y,Y} \underbrace{f_{Y}(y)}_{Y,Y} = \underbrace{f_{X}(x)} \underbrace{f_{X}(y)}_{Y,Y} = \underbrace{f_{X}(x)} \underbrace{f_{X}(y)}_{Y,Y} = \underbrace{f_{X}(x)} \underbrace{f_{X}(x)}_{Y,Y} = \underbrace{f_{X}($
- Let g(x,y) = u(x)v(y)

$$\begin{aligned} \mathbf{E}_{X,Y}[g(X,Y)] &= \int \int u(x)v(y)f_X(x)f_Y(y)\mathrm{d}x\mathrm{d}y \\ &= \int u(x)f_X(x)\mathrm{d}x \int v(y)f_Y(y)\mathrm{d}y \\ &= \mathbf{E}_X[u(X)] \ \mathbf{E}_Y[v(Y)] \end{aligned}$$

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Expectation of A Function of Two RVs

• X and Y are mutually independent. For general function g(x,y)

$$\mathbf{E}_{X,Y}[g(X,Y)] = \int \int g(x,y) f_{X,Y}(x,y) \mathrm{d}x \mathrm{d}y$$

$$\mathbb{E}\left[g(X,Y)\right] = \int \int g(x,y) \underline{f}_{X,Y}(x,y) dx dy$$

$$= \int \int g(x,y) f_X(x) f_Y(y) dx dy$$

$$= \int \int g(x,y) f_X(x) dx f_Y(y) dy$$

$$Given y, think $g(x,y)$ as function of $h(x) = g(x,y)$. Then
$$\int g(x,y) f_X(x) dx = \int h(x) f_X(x) dx$$

$$= \mathbb{E}_{\underline{X}}\left[\underline{h(X)}\right] = \mathbb{E}_{X}\left[\underline{g(X,y)}\right]$$$$

Given
$$y$$
, think $g(x,y)$ as function of x :
$$\underbrace{\frac{h(x) = g(x,y)}{\int g(x,y)f_X(x)\mathrm{d}x} = \int h(x)f_X(x)\mathrm{d}x}_{==\underline{\mathbb{E}_X}\left[\underline{h(X)}\right] = \underline{\mathbb{E}_X}\left[\underline{g(X,y)}\right]}_{==\underline{\mathbb{E}_X}\left[\underline{h(X)}\right] = \underline{\mathbb{E}_X}\left[\underline{g(X,y)}\right]$$

Expectation of A Function of Two RVs

• X and Y are mutually independent. For general function g(x,y)

$$E_{X,Y}[g(X,Y)] = \int \int g(x,y)f_{X,Y}(x,y)dxdy$$

$$\mathbb{E}\left[g(X,Y)\right] = \int \int g(x,y) f_{X,Y}(x,y) \mathrm{d}x \, \mathrm{d}y$$

$$= \int \int g(x,y) f_{X}(x) f_{Y}(y) \mathrm{d}x \, \mathrm{d}y$$

$$= \int \int g(x,y) f_{X}(x) \mathrm{d}x f_{Y}(y) \mathrm{d}y$$

$$= \int \mathbb{E}_{X}\left[g(X,y)\right] f_{Y}(y) \mathrm{d}y$$

$$= \mathbb{E}_{Y}\left[\mathbb{E}_{X}\left[g(X,Y)\right]\right]$$
Take the expectation with respect to each random variable one by one.

Only for independent RVs

$$\mathbb{E}\left(g(x,y)\right) = \int g(x,y) \, dx = \int g(x,y) \, dx$$

Only for independent RVs

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Distribution of A Function of Two RVs

- Let $Z(\omega) = g(X_1(\omega), X_2(\omega)) \quad \forall \omega \in \Omega$
- For any set B

$$\mathbb{P}[Z \in B] = \mathbb{P}[g(X_1, X_2) \in B]$$

$$= \iint_{g(x_1, x_2) \in B} f_{X_1, X_2}(x_1, x_2) dx_1 dx_2$$

• Example: Let X and Y are two random variables with $f_{X,Y}(x,y) = \begin{cases} \frac{1}{a^2} & \text{if } a \geq x, y \geq 0 \\ 0 & \text{otherwise} \end{cases}$ Compute the distribution of Z = X + Y.

Example: Sum of Two RVs

Example: Let X and Y are two random variables with $f_{X,Y}(x,y) = \begin{cases} \frac{1}{a^2} & \text{if } a \geq x,y \geq 0 \\ 0 & \text{otherwise} \end{cases}$ Compute the distribution of Z = X + Y. $F_Z(z) = \mathbb{P}(Z \leq z) = \mathbb{P}(X + Y \leq z) = \int_{x+y \leq z} f_{X,Y}(x,y) \mathrm{d}x \mathrm{d}y$

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Example: Sum of Two RVs

Example: Let X and Y are two random variables with $f_{X,Y}(x,y) = \begin{cases} \frac{1}{a^2} & \text{if } a \geq x, y \geq 0 \\ 0 & \text{otherwise} \end{cases}$ Compute the distribution of Z = X + Y.

$$F_Z(z) = \mathbb{P}(Z \le z) = \mathbb{P}(X + Y \le z) = \int_{x+y \le z} f_{X,Y}(x,y) \mathrm{d}x \mathrm{d}y$$

$$\mathbb{P}[Z \leq z] = \begin{cases} 0 & z < 0 \\ \frac{1}{2}z^2 & 1 > z > 0 \\ 1 - \frac{1}{2}(2 - z)^2 & 2 > z > 1 \\ 1 & z \geq 2 \end{cases}$$

Joint MGF

· Joint MGF of RVs are defined as

$$M_{X,Y}(t,s) = \mathbb{E}[e^{tX+sY}] = \int \int e^{tx+sy} f_{X,Y}(x,y) dxdy$$

• If random variables X and Y are mutually independent, then

$$\mathbf{E}_{X,Y}[u(X)v(Y)] = \mathbf{E}_X[u(X)] \ \mathbf{E}_Y[v(Y)]$$

$$M_{X,Y}(t,s) = \mathbb{E}[\underline{e^{tX+sY}}] = \mathbb{E}[\underline{e^{tX}}] \mathbb{E}[\underline{e^{sY}}] = \underbrace{M_X(t)M_Y(s)}$$

• Can be helpful in computation of functions of RVs.

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Example: Sum of Two RVs

Example: Let X and Y are two independent exponential random variables with

$$f_X(x) = \lambda e^{-\lambda x} 1(x \ge 0) \checkmark$$

$$f_{X,Y}(x,y) = \lambda e^{-\lambda x} \lambda e^{-\lambda y} 1(x \ge 0) 1(y \ge 0)$$

$$f_{Y,Y}(y) = \lambda e^{-\lambda x} 1(y \ge 0) \checkmark$$

Compute the distribution of Z = X + Y.

$$F_Z(z) = \mathbb{P}(Z \le z) = \mathbb{P}(X + Y \le z) = \int_{x+y \le z} \underbrace{f_{X,Y}(x,y) dxdy}$$

Example: Sum of Two RVs

Example: Let X and Y are two independent exponential random variables with

$$f_X(x) = \lambda e^{-\lambda x} 1(x \ge 0)$$

$$f_Y(y) = \lambda e^{-\lambda y} 1(y \ge 0)$$

$$f_{X,Y}(x,y) = \lambda e^{-\lambda x} \lambda e^{-\lambda y} 1(x \ge 0) 1(y \ge 0)$$

$$M_X(t) = \frac{1}{1 - t/\lambda} \qquad M_Y(t) = \frac{1}{1 - t/\lambda}$$

Compute the distribution of Z = X + Y.

$$M_{Z}(t) = \mathbb{E}[e^{tZ}] = \mathbb{E}[e^{tX+tY}] = \underline{M_{X}(t)M_{Y}(t)} = \underbrace{\begin{pmatrix} 1 \\ 1-t/\lambda \end{pmatrix}}_{=} \underbrace{\begin{pmatrix} 1 \\ 1-t/\lambda \end{pmatrix}^{2}}_{=} \sim \underbrace{\begin{pmatrix} 1 \\ 1-t/\lambda$$

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Example: Sum of n RVs

Example: Let $\{X_i\}$ are n independent exponential random variables with

$$f_{X_i}(x) = \lambda e^{-\lambda x} \mathbf{1}(x \ge 0)$$

$$f_{X_i}(x) = \lambda e^{-\lambda x} \mathbf{1}(x \ge 0)$$
 $M_{X_i}(t) = \frac{1}{1 - t/\lambda}$

Compute the distribution of $Z = \sum_i X_i$.

The distribution of
$$Z = \sum_i X_i$$
.
$$M_Z(t) = \mathbb{E}[e^{tZ}] = \mathbb{E}[e^{tX_1 + tX_2 + \dots + tX_n}] = M_{X_1}(t) M_{X_2(t)} \cdots M_{X_n}(t)$$

$$= \prod_{i > 1} M_{X_i}(t) \prod_{i > 1} M_{X_i}(t)$$

$$= \prod_{i > 1} M_{X_i}(t) \prod_{i > 1} M_{X_i}(t)$$

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