

STAT 1151

Introduction to Probability

Lecture 4 Joint Probability Distributions

Xiaomei Tan

Announcement

1. Midterm:

- Time: 18:30-20:30 November 3rd (Friday Week 9)
- Location: 综C 204

2. Office hour:

- No office hour this week
- Time: 13:30-16:30 October 7th (Saturday)
9:00-12:00 October 8th (Sunday)
- Location: 3-317B

Last Lecture

Chapter 2. Probability

Conditional Probability of An Event

Conditional Probability

Independence

Product Rule

Total Probability

Bayes' Rule

Chapter 3. Random Variables & Probability Distributions

Random Variables

Discrete Random Variables

Continuous Random Variables

Discrete Probability Distributions

Probability Function

Cumulative Distribution Function

Continuous Probability Distributions

Probability Density Function

Cumulative Distribution Function

Outline

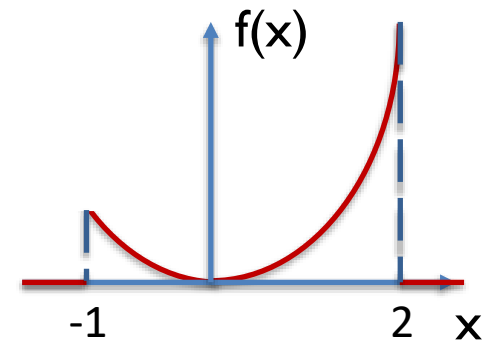
- Chapter 3 Random Variables and Probability Distributions
 - Continuous Probability Distributions
 - Joint Probability Distributions
 - Two discrete random variables
 - Two continuous random variables
 - More than two random variables

Probability Density Function

Example:

Suppose that the error in the reaction temperature ($^{\circ}\text{C}$) for a controlled laboratory experiment is a continuous random variable X having the probability density function

$$f(x) = \begin{cases} \frac{x^2}{3}, & -1 < x < 2 \\ 0, & \text{elsewhere} \end{cases}$$



(a) Verify that $f(x)$ is a density function.

1. $f(x) \geq 0$, for all $x \in R$ ✓

2. $\int_{-\infty}^{\infty} f(x) dx = 1$

➡ $\int_{-\infty}^{\infty} \frac{x^2}{3} dx = \frac{x^3}{9} \Big|_{-1}^2 = \frac{8}{9} + \frac{1}{9} = 1$ ✓

3. $P(a < X < b) = \int_a^b f(x) dx$

(b) Find $P(0 < X \leq 1)$.

➡ $P(0 < X \leq 1) = \int_0^1 \frac{x^2}{3} dx = \frac{x^3}{9} \Big|_0^1 = \frac{1}{9}$

(c) Find $P(X \leq 1)$.

❓ Cumulative distribution function

Cumulative Distribution Function

Definition:

The **cumulative distribution function** $F(x)$ of a continuous random variable X with density function $f(x)$ is

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t)dt, \quad \text{for } -\infty < x < \infty$$

where $f(x) = \frac{dF(x)}{dx}$, if the derivative exists.

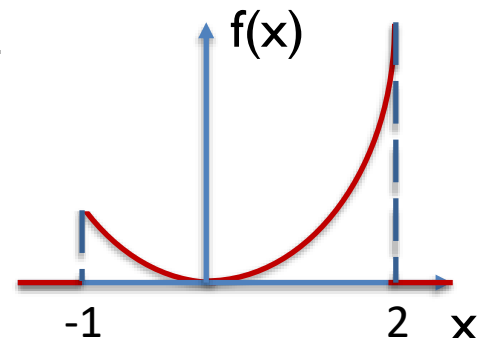
Example in last page:

- For $-1 \leq x < 2$, $F(x) = \int_{-\infty}^x f(t)dt = \int_{-1}^x \frac{t^2}{3} dt = \frac{x^3 + 1}{9}$

- For $x < -1$, $F(x) = 0$

- For $x \geq 2$, $F(x) = 1$

So $F(x) = \left\{ \begin{array}{ll} 0 & x < -1 \\ \frac{x^3 + 1}{9} & -1 \leq x < 2 \\ 1 & x \geq 2 \end{array} \right.$



Cumulative Distribution Function

Example:

The Department of Energy (DOE) puts projects out on bid and generally estimates what a reasonable bid should be. Call the estimate b . The DOE has determined that the density function of the winning (low) bid is

$$f(x) = \begin{cases} \frac{5}{8b}, & \frac{2b}{5} \leq x \leq 2b \\ 0, & \text{elsewhere} \end{cases}$$

Find $F(x)$.

- For $2b/5 \leq x \leq 2b$,
$$F(x) = \int_{-\infty}^x \frac{5}{8b} dt = \int_{2b/5}^x \frac{5}{8b} dt = \left. \frac{5t}{8b} \right|_{2b/5}^x = \frac{5x}{8b} - \frac{1}{4}$$

- For $x < 2b/5$, $F(x) = 0$
- For $x > 2b$, $F(x) = 1$

$$F(x) = \begin{cases} 0, & x < \frac{2b}{5} \\ \frac{5x}{8b} - \frac{1}{4}, & \frac{2b}{5} \leq x \leq 2b \\ 1, & x > 2b \end{cases}$$

Using $F(x)$ to Compute Probabilities

Proposition:

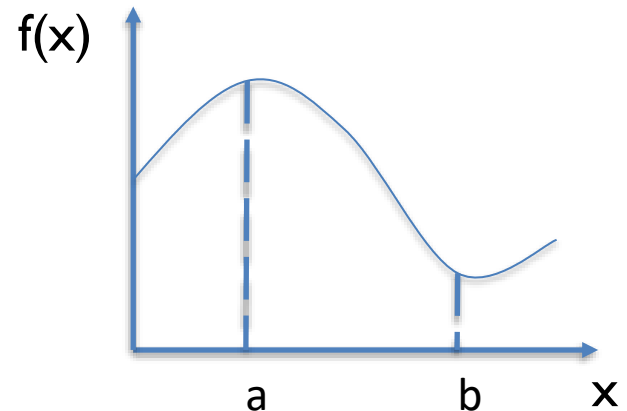
Let X be a continuous random variable with probability density function $f(x)$ and cumulative distribution function $F(x)$.

- For any number a ,

$$P(X > a) = 1 - F(a)$$

- For any two numbers a and b with $a < b$,

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



Using $F(x)$ to Compute Probabilities

Example:

Suppose the probability density function of the magnitude X of a dynamic load on a bridge (in newtons) is given by

$$f(x) = \begin{cases} \frac{1}{8} + \frac{3}{8}x, & 0 \leq x \leq 2 \\ 0, & \text{elsewhere} \end{cases}$$

Find $P(X > 1)$ and $P(1 \leq X \leq 1.5)$.

- For $0 \leq x \leq 2$, $F(x) = \int_{-\infty}^x (\frac{1}{8} + \frac{3}{8}t) dt = \frac{t}{8} + \frac{3}{16}t^2 \Big|_0^x = \frac{x}{8} + \frac{3}{16}x^2$
- For $x < 0$, $F(x) = 0$
- For $x > 2$, $F(x) = 1$

$$F(x) = \begin{cases} 0, & x < 0 \\ \frac{x}{8} + \frac{3}{16}x^2, & 0 \leq x \leq 2 \\ 1, & x > 2 \end{cases}$$

Joint Probability Distributions

Many problems in probability involve several random variables simultaneously.

Ex. - #successful free throws (F) and #successful three-point shots (T) of a basket player

- weight (W) and height (H)
- To determine the likelihood of success in college based on a student's high school data: college entrance exam score (E), high school class rank (R), and grade-point average at the end of freshman year (F)

Consider joint probability distributions for

- Two discrete random variables
- Two continuous random variables
- More than two random variables

Two Discrete Random Variables

Definition:

Consider two discrete random variables, X and Y .

The function $f(x, y)$ is a **joint probability distribution** if the following conditions hold:

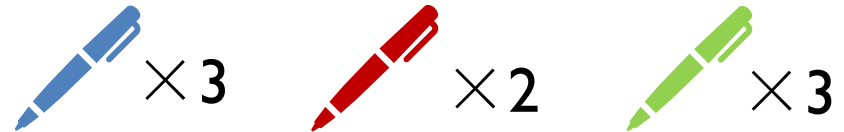
1. $f(x, y) \geq 0$ for all (x, y)
2. $\sum_x \sum_y f(x, y) = 1$
3. $P(X = x, Y = y) = f(x, y)$

For any region A in the xy plane, $P[(X, Y) \in A] = \sum \sum_A f(x, y)$

Two Discrete Random Variables

Example:

Two ballpoint pens are selected at random from a box that contains 3 blue pens, 2 red pens, and 3 green pens. If X is the number of **blue** pens selected and Y is the number of **red** pens selected.

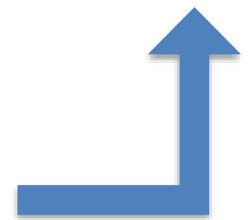


Possible (x, y) :

$(0, 0)$, $(0, 1)$, $(0, 2)$, $(1, 0)$, $(1, 1)$, and $(2, 0)$

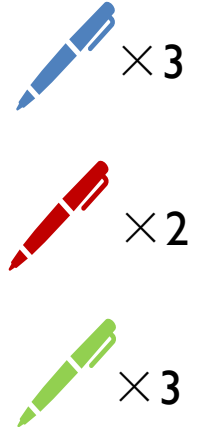
		y		
		0	1	2
x	$f(x, y)$	$\binom{3}{2} / \binom{8}{2} = \frac{3}{28}$	$\binom{2}{1} \binom{3}{1} / \binom{8}{2}$	$\binom{2}{2} / \binom{8}{2}$
	0	$\binom{3}{1} \binom{3}{1} / \binom{8}{2}$	$\binom{3}{1} \binom{2}{1} / \binom{8}{2}$	0
	1	$\binom{3}{2} / \binom{8}{2}$	0	0
	2			

Joint probability table



Two Discrete Random Variables




		y		
		0	1	2
x	$f(x, y)$			
	0	$3/28$	$3/14$	$1/28$
	1	$9/28$	$3/14$	0
	2	$3/28$	0	0



(a) Find the joint probability function $f(x, y)$.

(b) Find $P[(X, Y) \in A]$, where A is the region $\{(x, y) | x + y \leq 1\}$.

Two Discrete Random Variables

		y			Row Totals	
f(x, y)		0	1	2		
x	0	3/28	3/14	1/28	5/14	 × 3
	1	9/28	3/14	0	15/28	 × 2
	2	3/28	0	0	3/28	 × 3
					1	

(c) Find the probability that only one blue pen was selected.

The probability distribution **$g(x)$** of X alone results from holding x fixed and summing the joint probability distributions $f(x, y)$ over the values of Y .

Marginal Distributions of X

Two Discrete Random Variables

Definition:

For discrete random variables X and Y ,

- the **Marginal Distributions** of X alone is $g(x) = \sum_y f(x, y)$

- the **Marginal Distributions** of Y alone is $h(y) = \sum_x f(x, y)$

$$g(0) = f(0,0) + f(0,1) + f(0,2) = 5/14$$

$$h(0) = f(0,0) + f(1,0) + f(2,0) = 15/28$$

		y				
		f(x, y)	0	1	2	Row Totals
x	0	3/28	3/14	1/28	5/14	$g(0)$
	1	9/28	3/14	0	15/28	$g(1)$
	2	3/28	0	0	3/28	$g(2)$
Column Totals		15/28	3/7	1/28	1	
		$h(0)$	$h(1)$	$h(2)$		

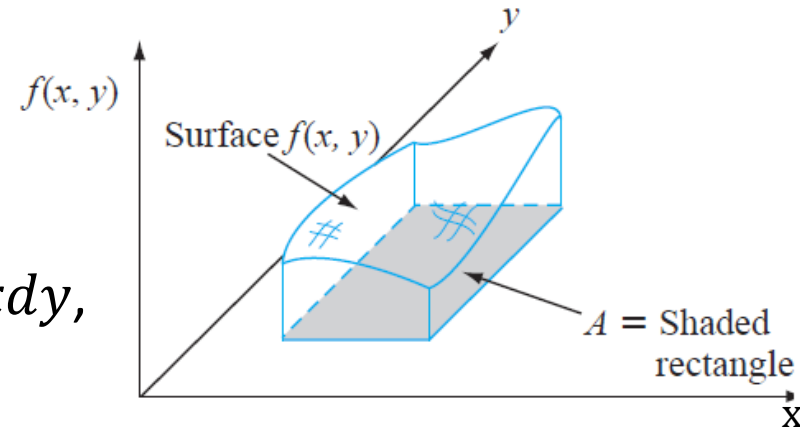
Two Continuous Random Variables

Definition:

Consider two continuous random variables, X and Y .

The function $f(x, y)$ is a **joint (probability) density function** if the following conditions hold:

1. $f(x, y) \geq 0$ for all (x, y) ,
2. $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$,
3. $P[(X, Y) \in A] = \int \int_A f(x, y) dx dy$,
for any region A in the xy plane.



In particular, if A is the two-dimensional rectangle $\{(x, y): a \leq x \leq b, c \leq y \leq d\}$, then

$$P[(X, Y) \in A] = P(a \leq X \leq b, c \leq Y \leq d) = \int_c^d \int_a^b f(x, y) dx dy$$

Two Continuous Random Variables

Example:

A privately owned business operates both a drive-in facility and a walk-in facility. On a randomly selected day, let X and Y , respectively, be the proportions of the time that the drive-in and the walk-in facilities are in use, and suppose that the joint density function of these random variables is

$$f(x, y) = \begin{cases} \frac{2}{5}(2x + 3y), & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

(a) Verify $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \, dx \, dy = 1$

Two Continuous Random Variables

Example:

A privately owned business operates both a drive-in facility and a walk-in facility. On a randomly selected day, let X and Y , respectively, be the proportions of the time that the drive-in and the walk-in facilities are in use, and suppose that the joint density function of these random variables is

$$f(x, y) = \begin{cases} \frac{2}{5}(2x + 3y), & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

(b) Find $P[(X, Y) \in A]$, where $A = \{(x, y) | 0 < x < 0.5, 0.25 < y < 0.5\}$

Two Continuous Random Variables

Definition:

For continuous random variables X and Y ,

- the **Marginal Distributions** of X alone is $g(x) = \int_{-\infty}^{\infty} f(x, y) dy$
- the **Marginal Distributions** of Y alone is $h(y) = \int_{-\infty}^{\infty} f(x, y) dx$

The marginal probability density function $g(x)$ of X alone results from holding x fixed and integrating the joint density function $f(x, y)$ over y .

Two Continuous Random Variables

Revisit example:

A privately owned business operates both a drive-in facility and a walk-in facility. On a randomly selected day, let X and Y , respectively, be the proportions of the time that the drive-in and the walk-in facilities are in use, and suppose that the joint density function of these random variables is

$$f(x, y) = \begin{cases} \frac{2}{5}(2x + 3y), & 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

(c) Find $g(x)$ and $h(y)$.

$$g(x) = \int_{-\infty}^{\infty} f(x, y) dy = \left\{ \right.$$

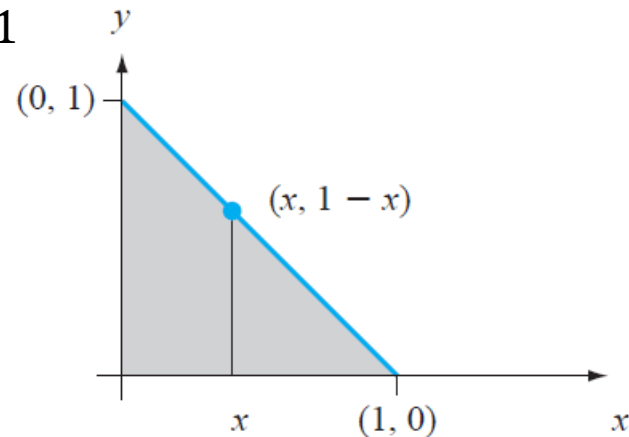
$$h(y) = \int_{-\infty}^{\infty} f(x, y) dx = \left\{ \right.$$

Two Continuous Random Variables

Consider a more complicated case that the region of positive density is not a rectangle:

A company sells fruit salad consisting of three types of fruits: apples, oranges, and grapes. Suppose the net weight of each salad is exactly 1 lb, but the weight contribution of each type of fruit is random. Because the three weights sum to 1, a joint probability model for any two gives all necessary information about the weight of the third type. Let X = the weight of apples in a selected salad and Y = the weight of oranges. Then the region of positive density is $D = \{(x, y): 0 \leq x \leq 1, 0 \leq y \leq 1, x + y \leq 1\}$, the shaded region pictured in Figure. The joint density function is

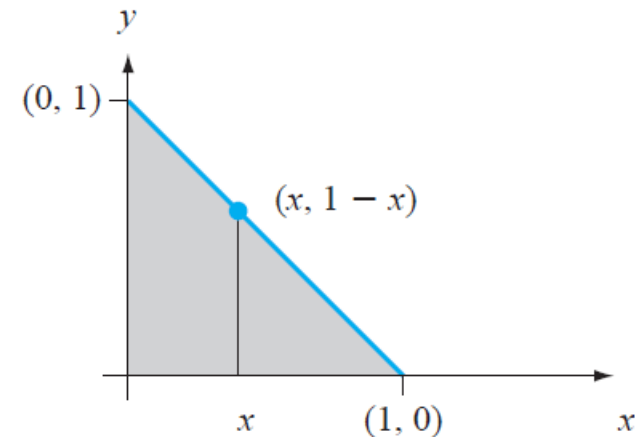
$$f(x, y) = \begin{cases} 24xy, & 0 \leq x \leq 1, 0 \leq y \leq 1, x + y \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$



Two Continuous Random Variables

$$f(x, y) = \begin{cases} 24xy, & 0 \leq x \leq 1, 0 \leq y \leq 1, x + y \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

(a) Verify $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$



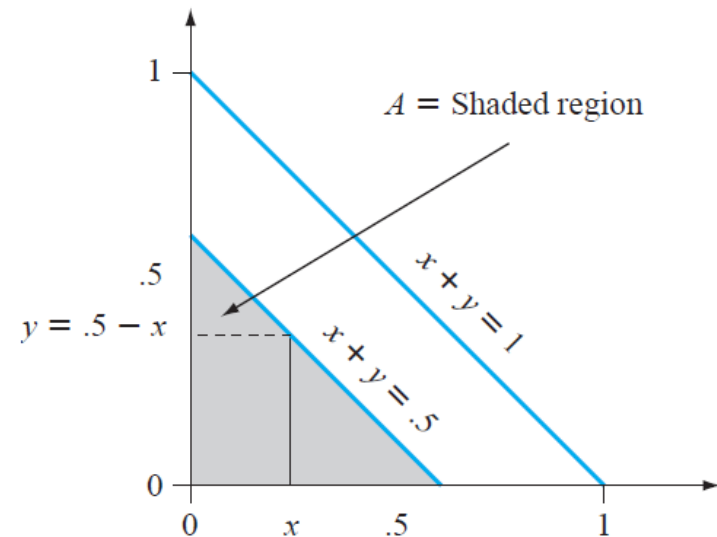
Two Continuous Random Variables

$$f(x, y) = \begin{cases} 24xy, & 0 \leq x \leq 1, 0 \leq y \leq 1, x + y \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

(b) Find the probability that apples and oranges make up at most 50% of the salad.

Let $A = \{(x, y): 0 \leq x \leq 1, 0 \leq y \leq 1, \text{ and } x + y \leq 0.5\}$

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy =$$



Conditional Probability Distribution

Conditional probability of event B, given event A:

$$P(B|A) = \frac{P(B \cap A)}{P(A)}, \text{ provided } P(A) > 0$$

If A and B are defined by **discrete/continuous random variables** **X=x** and **Y=y**, respectively:

$$P(Y = y|X = x) = \frac{P(X = x, Y = y)}{P(X = x)} = \frac{f(x, y)}{g(x)}, \text{ provided } g(x) > 0$$



A function of y with x fixed (discrete rvs)
or joint density divided by marginal distribution (continuous rvs)

Conditional Probability Distribution

Definition:

Let X and Y be two random variables, discrete or continuous. The **conditional distribution of Y** given that $X=x$ is

$$f(y|x) = \frac{f(x, y)}{g(x)}, \text{ provided } g(x) > 0$$

Similarly, the **conditional distribution of X** given that $Y=y$ is

$$f(x|y) = \frac{f(x, y)}{h(y)}, \text{ provided } h(y) > 0$$

The probability that X falls between a and b when it is known that the $Y = y$ is

$$P(a < X < b | Y = y) = \sum_{a < x < b} f(x|y), \quad \text{for discrete } X \text{ and } Y$$

$$P(a < X < b | Y = y) = \int_a^b f(x|y) dx, \quad \text{for continuous } X \text{ and } Y$$

Conditional Probability Distribution

Revisit example:

Two pens are randomly selected. X is the number of **blue** pens selected and Y is the number of **red** pens selected.



(a) Find the conditional distribution of X , given that $Y = 1$.

(b) Find $P(X=0|Y=1)$.

		y				
		$f(x, y)$	0	1	2	Row Totals
x	0	3/28	3/14	1/28	5/14	$g(0)$
	1	9/28	3/14	0	15/28	$g(1)$
	2	3/28	0	0	3/28	$g(2)$
Column Totals		15/28	3/7	1/28	1	
		$h(0)$	$h(1)$	$h(2)$		

Conditional Probability Distribution

Example:

Given the joint density function

$$f(x, y) = \begin{cases} \frac{x(1 + 3y^2)}{4}, & 0 < x < 2, 0 < y < 1 \\ 0, & \text{elsewhere} \end{cases}$$

(a) Find $g(x)$ and $h(y)$.

(b) Find $f(x|y)$.

(c) Find $P(0.25 < X < 0.5 | Y = 1/3)$.

Independent Random Variables

If $f(x|y)$ does not depend on y , the outcome of Y has no impact on the outcome of X . We say X and Y are independent random variables.

Definition:

Let X and Y be two random variables, discrete or continuous, with joint probability distribution $f(x, y)$ and marginal distributions $g(x)$ and $h(y)$, respectively. The random variables X and Y are said to be **statistically independent** if and only if

$$f(x, y) = g(x)h(y) \quad \Leftrightarrow \quad f(x|y) = g(x)$$

for all (x, y) within their range.

$$\Leftrightarrow f(y|x) = h(y)$$

Independent Random Variables

Checking for statistical independence of discrete rvs requires more attention. $f(x, y) = g(x)h(y)$ may apply to some but not all (x, y) !!

Example:

$$f(1,1) = g(1)h(1)$$

For other (x,y) ,

e.g., $f(0,0) \neq g(0)h(0)$

	f(x, y)	y			Row Totals	
		0	1	2		
x	0	1/4	0	0	1/4	$g(0)$
	1	0	1/4	1/4	1/2	$g(1)$
	2	0	1/4	0	1/4	$g(2)$
Column Totals		1/4	1/2	1/4	1	
		$h(0)$	$h(1)$	$h(2)$		

More Than Two Random Variables

Let X_1, X_2, \dots, X_n be random variables.

- If X_1, X_2, \dots, X_n are all discrete random variables, the **joint probability function** of the rvs is

$$f(x_1, x_2, \dots, x_n) = P(X_1 = x_1, X_2 = x_2, \dots, X_n = x_n)$$

- If X_1, X_2, \dots, X_n are continuous random variables, the **joint density function** is

$$f(x_1, x_2, \dots, x_n) = \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f(x_1, x_2, \dots, x_n) dx_1 \dots dx_n$$

More Than Two Random Variables

- The **marginal distribution** of rv X_1 , for example, is

$$g(x_1) = \begin{cases} \sum_{x_2} \dots \sum_{x_n} f(x_1, x_2, \dots, x_n), & \text{for discrete case} \\ \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f(x_1, x_2, \dots, x_n) dx_2 \dots dx_n, & \text{for continuous case} \end{cases}$$

- The **joint marginal distribution** of rvs X_1 and X_2 , for example, is

$$g(x_1, x_2) = \begin{cases} \sum_{x_3} \dots \sum_{x_n} f(x_1, x_2, \dots, x_n), & \text{for discrete case} \\ \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f(x_1, x_2, \dots, x_n) dx_3 \dots dx_n, & \text{for continuous case} \end{cases}$$

More Than Two Random Variables

- The **joint conditional distribution** of rvs X_1, X_2 , and X_3 , given that $X_4 = x_4, \dots, X_n = x_n$ for example, is

$$f(x_1, x_2, x_3 | x_4, \dots, x_n) = \frac{f(x_1, x_2, \dots, x_n)}{g(x_4, \dots, x_n)}$$

- **Mutual statistical independence** of rvs X_1, X_2, \dots, X_n , with marginal distributions $f_1(x_1), f_2(x_2), \dots, f_n(x_n)$, respectively

$$f(x_1, x_2, \dots, x_n) = f_1(x_1) f_2(x_2) \dots f_n(x_n)$$

for all (x_1, x_2, \dots, x_n) within their range.

More Than Two Random Variables

Example:

Suppose that the shelf life (years) of a certain perishable food product packaged in cardboard containers is a random variable whose probability density function is given by

$$f(x) = \begin{cases} e^{-x}, & x > 0 \\ 0, & \text{elsewhere} \end{cases}$$

Let X_1, X_2 , and X_3 represent the shelf lives for three of these containers selected **independently** and find $P(X_1 < 2, 1 < X_2 < 3, X_3 > 2)$.

$$f(x_1, x_2, x_3) = \begin{cases} e^{-x_1} e^{-x_2} e^{-x_3} = e^{-x_1-x_2-x_3}, & x_1, x_2, x_3 > 0 \\ 0, & \text{elsewhere} \end{cases}$$

$$P(X_1 < 2, 1 < X_2 < 3, X_3 > 2) = \int_2^\infty \int_1^3 \int_0^2 e^{-x_1-x_2-x_3} dx_1 dx_2 dx_3$$

$$* \int e^{-x} dx = -e^{-x}$$