

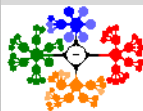
Lecture 5

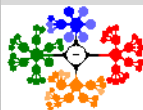
Random Variables and Probability Distributions

Lecture Information

Ceng272 *Statistical Computations* at March 15, 2010

Dr. Cem Özdoğan
Computer Engineering Department
Çankaya University





1 Random Variables and Probability Distributions

Concept of a Random Variable

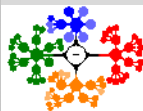
Discrete Probability Distributions

Continuous Probability Distributions

Joint Probability Distribution

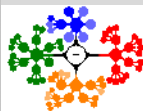
Concept of a Random Variable I

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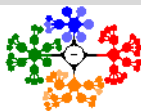
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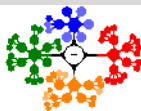
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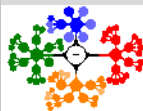
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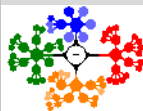
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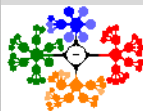
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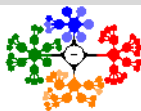
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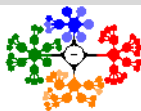
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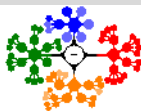
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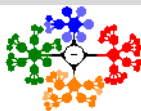
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Sample Space	y
RR	2
RB	1
BR	1
BB	0



Concept of a Random Variable II

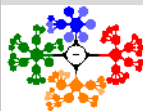


Example: Number of defective (D) products when 3 products are tested.

Outcomes in Sample Space	x: value of X
DDD	3
DDN	2
DND	2
DNN	1
NDD	2
NDN	1
NND	1
NNN	0

Concept of a Random Variable III

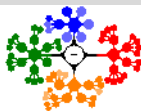
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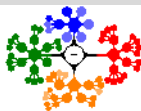


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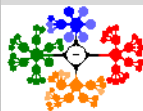


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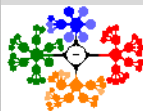


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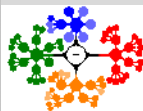


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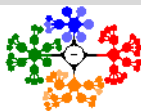
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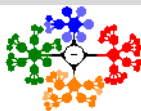


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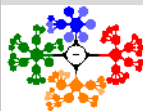
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NND	3
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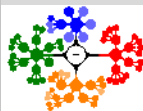
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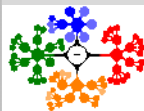
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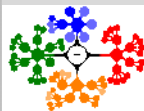
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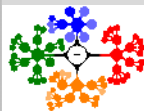
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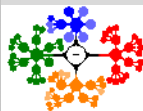
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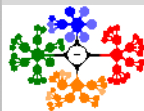


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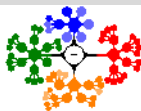
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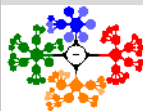
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Continuous sample space: If a sample space contains an infinite number of possibilities equal to the number of points on a line segment.



Discrete Probability Distributions I

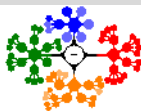
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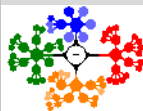
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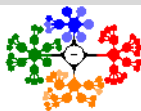
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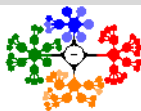
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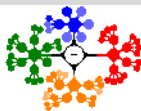
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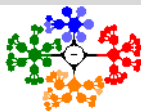
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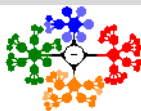
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- The probability distribution of a discrete random variable can be presented in the form of a mathematical formula, a table, or a graph-probability histogram or barchart.



Discrete Probability Distributions II



Example: Let X be the random variable: number of heads in 3 tosses of a fair coin.

Sample Space	x
TTT	0
TTH	1
THT	1
THH	2
HTT	1
HTH	2
HHT	2
HHH	3

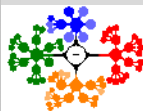
$P(X = x)$: Probability that outcome is a specific x value.

x	0	1	2	3
$P(X=x)$	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$

Discrete Probability Distributions III

Random Variables and
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Concept of a Random
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Discrete Probability
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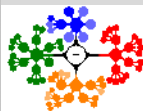
Continuous Probability
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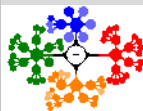
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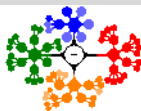
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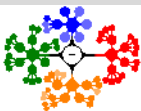
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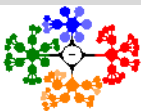
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f(x)	$\frac{10}{28}$	$\frac{15}{28}$	$\frac{3}{28}$



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f(x)	$\frac{10}{28}$	$\frac{15}{28}$	$\frac{3}{28}$

$$f(0) = P(X = 0) = \frac{\binom{3}{0} \binom{5}{2}}{\binom{8}{2}} = \frac{10}{28}$$

$$f(1) = P(X = 1) = \frac{\binom{3}{1} \binom{5}{1}}{\binom{8}{2}} = \frac{15}{28}$$

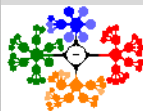
$$f(2) = P(X = 2) = \frac{\binom{3}{2} \binom{5}{0}}{\binom{8}{2}} = \frac{3}{28}$$

Discrete Probability Distributions IV

- **Definition 3.5:**

The **Cumulative distribution function** $F(x)$ of a discrete random variable X with probability distribution $f(x)$ is

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



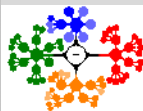
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- **Example 3.10:** Find the cumulative distribution of the random variable X in Example 3.9.



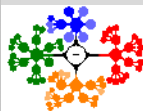
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$$f(0) = \frac{1}{16}, f(1) = \frac{4}{16}, f(2) = \frac{6}{16}, f(3) = \frac{4}{16}, f(4) = \frac{1}{16},$$

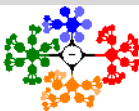
$$F(0) = f(0) = \frac{1}{16}$$

$$F(1) = f(0) + f(1) = \frac{5}{16}$$

$$F(2) = f(0) + f(1) + f(2) = \frac{11}{16}$$

$$F(3) = f(0) + f(1) + f(2) + f(3) = \frac{15}{16}$$

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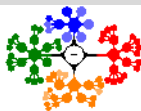
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$$F(x) = \left\{ \begin{array}{l} 0 \text{ for } x < 0 \\ \frac{1}{16} \text{ for } 0 \leq x < 1 \\ \frac{5}{16} \text{ for } 1 \leq x < 2 \\ \frac{11}{16} \text{ for } 2 \leq x < 3 \\ \frac{15}{16} \text{ for } 3 \leq x < 4 \\ 1 \text{ for } x \geq 4 \end{array} \right\}$$



Discrete Probability Distributions V

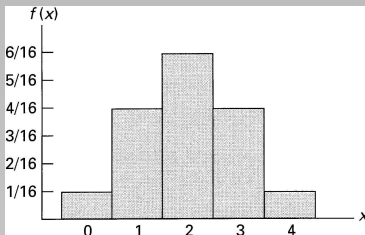
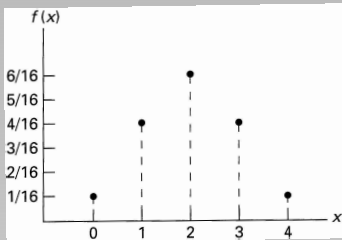
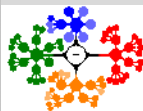


Figure: Bar chart and probability histogram

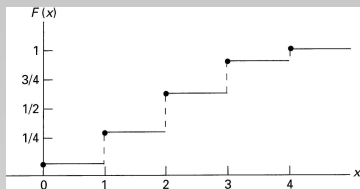


Figure: Discrete cumulative distribution.

Continuous Probability Distributions I

- A continuous random variable has a probability of zero of assuming exactly any of its values.

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

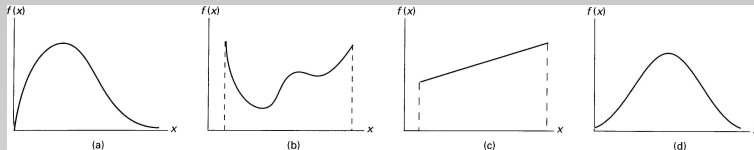
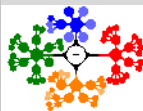


Figure: Typical density functions.

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 $P(X = 178 \text{ cm}) = 0$. No assuming exactly.

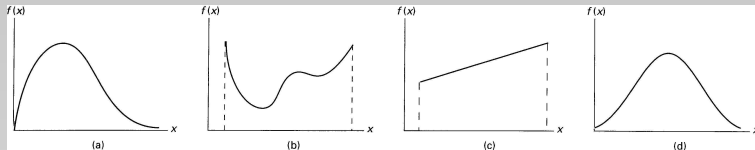
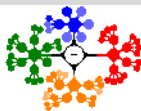


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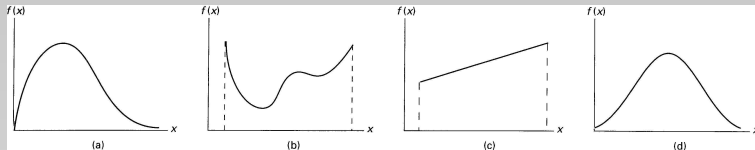
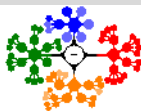


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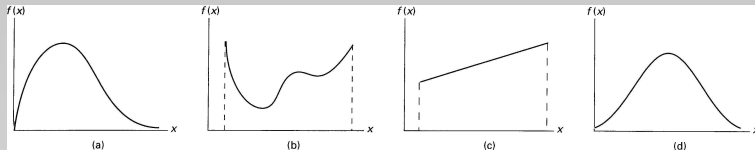
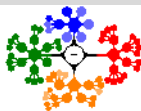


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- **Some of these functions are shown below:**

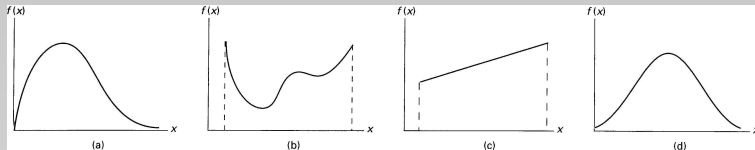
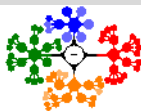
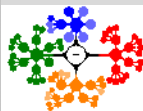


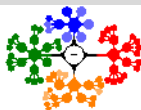
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The function $f(x)$ is a **probability density function** (or **density function, p.d.f**) for the continuous random variable X , defined over the set of real numbers R , if

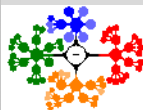




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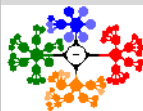
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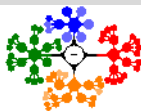
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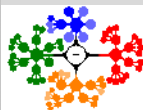
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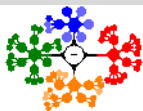


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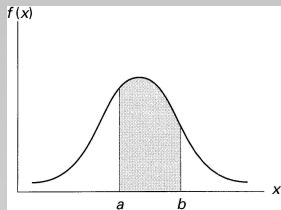
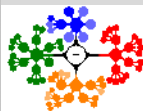


Figure: $P(a < X < b)$



- **Example 3.11:** Suppose that the error in reaction temperature in °C is a continuous random variable X having the probability density function

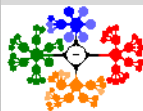
$$f(x) = \left\{ \begin{array}{l} \frac{x^2}{3} \text{ for } -1 < x < 2 \\ 0, \text{ elsewhere} \end{array} \right\}$$



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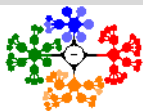


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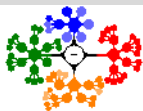
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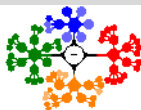
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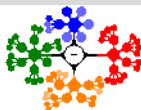
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The **cumulative function** $F(x)$ of a continuous random variable X with density function $f(x)$ is

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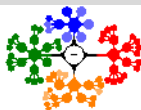


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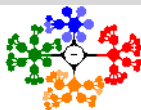


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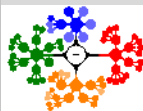
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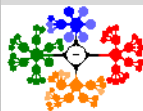
- An immediate consequence:
 - $P(a < X < b) = F(b) - F(a)$
 - $f(x) = \frac{dF(x)}{dx}$, if the derivative exists

Continuous Probability Distributions V

Example 3.12: For the density function of Example 3.6 find $F(x)$, and use it to evaluate $P(0 < X \leq 1)$.



Continuous Probability Distributions V



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For $-1 < x < 2$

$$F(x) = \int_{-\infty}^x f(t)dt = \int_{-\infty}^x \frac{t^2}{3} dt$$

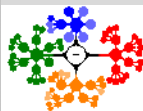
$$= \frac{t^3}{9} \Big|_{-\infty}^x = \frac{x^3 + 1}{9}$$

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

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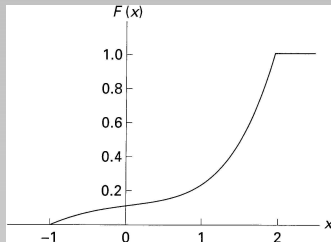
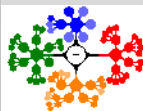


Figure: Continuous cumulative distribution function.

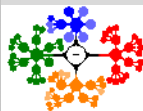
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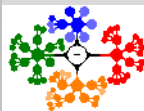
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- **Definition 3.8:**

The function $f(x, y)$ is a **joint probability distribution** (or **probability mass function**) of the discrete random variables X and Y if

For any region A in the xy -plane,

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Joint Probability Distribution I

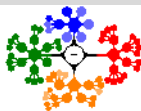
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Joint Probability Distribution I

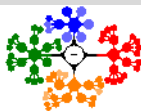
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Joint Probability Distribution I

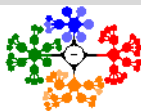
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- 3 $P(X = x, Y = y) = f(x, y)$

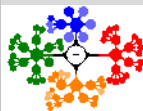
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Joint Probability Distribution II

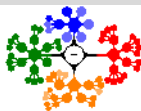
- **Example 3.14:** Two refills for a ballpoint pen are selected at random from a box that contains 3 blue refills, 2 red refills, and 3 green refills. If X is the number of blue refills and Y is the number of red refills selected, find



Joint Probability Distribution II

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 - the joint probability function $f(x, y)$

$$f(x, y) = \frac{\binom{3}{x} \binom{2}{y} \binom{3}{2-x-y}}{\binom{8}{2}}$$



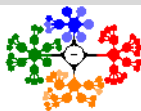
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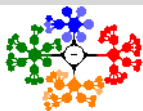
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- $P[(X, Y) \in A]$, where A is the region $\{(x, y) | x + y \leq 1\}$.

$$\begin{aligned} P[(X, Y) \in A] &= P(X + Y \leq 1) \\ &= f(0, 0) + f(0, 1) + f(1, 0) \\ &= \frac{3}{28} + \frac{3}{14} + \frac{9}{28} = \frac{9}{14} \end{aligned}$$



Joint Probability Distribution III



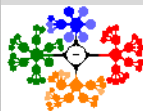
			x		Row Totals
$f(x, y)$		0	1	2	
	0	$\frac{3}{28}$	$\frac{9}{28}$	$\frac{3}{28}$	$\frac{15}{28}$
y	1	$\frac{3}{14}$	$\frac{3}{14}$		$\frac{3}{7}$
	2	$\frac{1}{28}$			$\frac{1}{28}$
Column Totals		$\frac{5}{14}$	$\frac{15}{28}$	$\frac{3}{28}$	1

Joint Probability Distribution IV

- **Definition 3.9:**

The function $f(x, y)$ is a **joint density function** of the continuous random variables X and Y if

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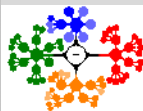


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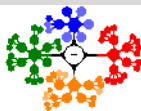
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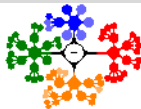


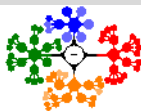
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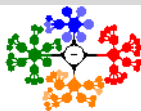
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- **Example 3.15:** A candy company distributes boxes of chocolates with a mixture of creams, toffees, and nuts coated in both light and dark chocolate.



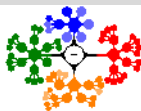
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- For randomly selected box, let X and Y , respectively, be the proportions of the light and dark chocolates that are creams.



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- 1 $f(x, y) \geq 0$, for all (x, y)
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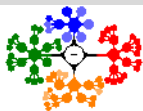
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- For randomly selected box, let X and Y , respectively, be the proportions of the light and dark chocolates that are creams.
- The joint density function is as follows:

$$f(x, y) = \left\{ \begin{array}{l} \frac{2}{5}(2x + 3y), 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0, \text{ elsewhere} \end{array} \right\}$$

Joint Probability Distribution V

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

$$\begin{aligned} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy &= \int_0^1 \int_0^1 \frac{2}{5} (2x + 3y) dx dy \\ &= \int_0^1 \left(\frac{2x^2}{5} + \frac{6xy}{5} \right) \Big|_{x=0}^{x=1} dy = \int_0^1 \left(\frac{2}{5} + \frac{6y}{5} \right) dy = \left(\frac{2y}{5} + \frac{3y^2}{5} \right) \Big|_0^1 \\ &= \frac{2}{5} + \frac{3}{5} = 1 \end{aligned}$$



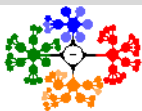
Joint Probability Distribution V

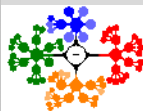
- Verify $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$

$$\begin{aligned}\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy &= \int_0^1 \int_0^1 \frac{2}{5} (2x + 3y) dx dy \\ &= \int_0^1 \left(\frac{2x^2}{5} + \frac{6xy}{5} \right) \Big|_{x=0}^{x=1} dy = \int_0^1 \left(\frac{2}{5} + \frac{6y}{5} \right) dy = \left(\frac{2y}{5} + \frac{3y^2}{5} \right) \Big|_0^1 \\ &= \frac{2}{5} + \frac{3}{5} = 1\end{aligned}$$

- $P[(X, Y) \in A]$, where A is the region $(x, y) | 0 < x < \frac{1}{2}, \frac{1}{4} < y < \frac{1}{2}$,

$$\begin{aligned}P[(X, Y) \in A] &= P\left(0 < X < \frac{1}{2}, \frac{1}{4} < Y < \frac{1}{2}\right) \\ &= \int_{\frac{1}{4}}^{\frac{1}{2}} \int_0^{\frac{1}{2}} \frac{2}{5} (2x + 3y) dx dy = \int_{\frac{1}{4}}^{\frac{1}{2}} \left(\frac{2x^2}{5} + \frac{6xy}{5} \right) \Big|_{x=0}^{x=\frac{1}{2}} dy \\ &= \int_{\frac{1}{4}}^{\frac{1}{2}} \left(\frac{1}{10} + \frac{3y}{5} \right) dy = \left(\frac{y}{10} + \frac{3y^2}{10} \right) \Big|_{\frac{1}{4}}^{\frac{1}{2}} = \frac{13}{160}\end{aligned}$$





- **Definition 3.10:**

The **marginal distributions** of X alone and of Y alone are

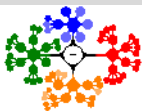
$$g(x) = \sum_y f(x, y) \text{ and } h(y) = \sum_x f(x, y)$$

for the discrete case

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

for the continuous case

Joint Probability Distribution VII



Example 3.16: Show that the column and row totals of the following table give the marginal distribution of X alone and of Y alone.

$f(x, y)$			x		Row Totals
	0	$\frac{3}{28}$	$\frac{9}{28}$	$\frac{3}{28}$	$\frac{15}{28}$
y	1	$\frac{3}{14}$	$\frac{3}{14}$		$\frac{3}{7}$
	2	$\frac{1}{28}$			$\frac{1}{28}$
Column Totals		$\frac{5}{14}$	$\frac{15}{28}$	$\frac{3}{28}$	1

Joint Probability Distribution VIII

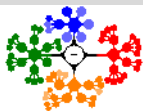
Solution:

$$\begin{aligned}P(X = 0) = g(0) &= \sum_{y=0}^2 f(0, y) = f(0, 0) + f(0, 1) + f(0, 2) \\ &= \frac{3}{28} + \frac{3}{14} + \frac{1}{28} = \frac{5}{14}\end{aligned}$$

$$\begin{aligned}P(X = 1) = g(1) &= \sum_{y=0}^2 f(1, y) = f(1, 0) + f(1, 1) + f(1, 2) \\ &= \frac{9}{28} + \frac{3}{14} + 0 = \frac{15}{28}\end{aligned}$$

$$\begin{aligned}P(X = 2) = g(2) &= \sum_{y=0}^2 f(2, y) = f(2, 0) + f(2, 1) + f(2, 2) \\ &= \frac{3}{28} + 0 + 0 = \frac{3}{28}\end{aligned}$$

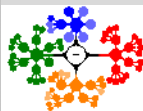
x	0	1	2
g(x)	$\frac{5}{14}$	$\frac{15}{28}$	$\frac{3}{28}$



Joint Probability Distribution IX

- **Example 3.17:** Find $g(x)$ and $h(y)$ for the following joint density function.

$$f(x, y) = \left\{ \begin{array}{l} \frac{2}{5}(2x + 3y), \quad 0 \leq x \leq 1, \quad 0 \leq y \leq 1 \\ 0, \quad \textit{elsewhere} \end{array} \right\}$$



Joint Probability Distribution IX

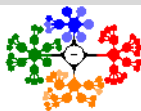
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- $g(x)$

$$\begin{aligned} &= \int_{-\infty}^{\infty} f(x, y) dy = \int_0^1 \frac{2}{5}(2x + 3y) dy \\ &= \left(\frac{4xy}{5} + \frac{6y^2}{10} \right) \Big|_{y=0}^{y=1} = \frac{4x + 3}{5} \end{aligned}$$

for $0 \leq x \leq 1$, $0 \leq y \leq 1$ and $g(x) = 0$, elsewhere



Joint Probability Distribution IX

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for $0 \leq x \leq 1$, $0 \leq y \leq 1$ and $g(x) = 0$, elsewhere

- $h(y)$

$$\begin{aligned} &= \int_{-\infty}^{\infty} f(x, y) dx = \int_0^1 \frac{2}{5}(2x + 3y) dx \\ &= \left(\frac{2x^2}{5} + \frac{6yx}{5} \right) \Big|_{x=0}^{x=1} = \frac{2 + 6y}{5} \end{aligned}$$

for $0 \leq x \leq 1$, $0 \leq y \leq 1$ and $h(y) = 0$, elsewhere



Joint Probability Distribution X

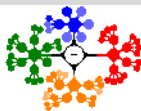
- **Definition 3.11:**

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

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

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

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



Joint Probability Distribution X

- **Definition 3.11:**

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

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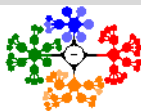
Similarly, the conditional distribution of the random variable X , given that $Y = y$, is

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

- Evaluate the probability that X falls between a and b given that Y is known.

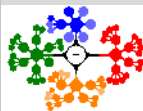
$$P(a < X < b | Y = y) = \sum_x f(x|y), \text{ for the discrete case}$$

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



Joint Probability Distribution XI

- **Example 3.18:** Referring to Example 3.14, find the conditional distribution of X , given that $Y = 1$, and use it to determine $P(X = 0|Y = 1)$.



Joint Probability Distribution XI

- **Example 3.18:** Referring to Example 3.14, find the conditional distribution of X , given that $Y = 1$, and use it to determine $P(X = 0|Y = 1)$.
- **Solution:**

$$h(y = 1) = \sum_{x=0}^2 f(x, 1) = \frac{3}{14} + \frac{3}{14} + 0 = \frac{3}{7}$$

$$f(x|1) = \frac{f(x, 1)}{h(1)} = \frac{7}{3}f(x, 1), x = 0, 1, 2$$

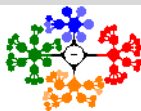
$$f(0|1) = \frac{7}{3}f(0, 1) = \frac{7}{3} * \frac{3}{14} = \frac{1}{2}$$

$$f(1|1) = \frac{7}{3}f(1, 1) = \frac{7}{3} * \frac{3}{14} = \frac{1}{2}$$

$$f(2|1) = \frac{7}{3}f(2, 1) = \frac{7}{3} * 0 = 0$$

$$\implies P(X = 0|Y = 1) = f(0|1) = \frac{1}{2}$$

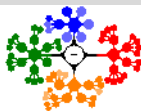
x	0	1	2
f(x 1)	$\frac{1}{2}$	$\frac{1}{2}$	0



Joint Probability Distribution XII

- **Example 3.19:** The joint density for the random variables (X, Y) , where X is the unit temperature change and Y is the proportion of spectrum shift that a certain atomic particle produces is

$$f(x, y) = \begin{cases} 10xy^2, & 0 < x < y < 1 \\ 0, & \text{elsewhere} \end{cases}$$



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$$f(x, y) = \begin{cases} 10xy^2, & 0 < x < y < 1 \\ 0, & \text{elsewhere} \end{cases}$$

- Find the marginal densities $g(x)$, $h(y)$, and the conditional density $f(y|x)$.

$$g(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_x^1 10xy^2 dy = \frac{10x(1-x^3)}{3}$$

$$h(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_0^y 10xy^2 dx$$

$$f(y|x) = \frac{f(x, y)}{g(x)} = \frac{10xy^2}{\frac{10x(1-x^3)}{3}} = \frac{3y^2}{(1-x^3)}$$



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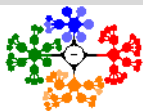
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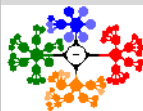
$$h(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_0^y 10xy^2 dx$$

$$f(y|x) = \frac{f(x, y)}{g(x)} = \frac{10xy^2}{\frac{10x(1-x^3)}{3}} = \frac{3y^2}{(1-x^3)}$$

- Find the probability that the spectrum shifts more than half of the total observations, given the temperature is increased to 0.25 unit.

$$P(Y > \frac{1}{2} | X = 0.25) = \int_{1/2}^1 f(y|0.25) dy = \int_{1/2}^1 \frac{3y^2}{(1 - 0.25^3)} dy = \frac{8}{9}$$

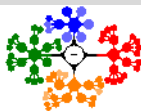




- **Definition 3.12:**

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

$$f(x, y) = g(x)h(y), \text{ for all } (x, y) \text{ within their range}$$



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- **Example 3.21:** Show that the random variables of Example 3.14 are not statistically independent.

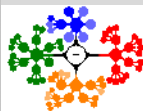
$$f(0, 1) = \frac{3}{14}, g(0) = \sum_{y=0}^2 f(0, y) = \frac{5}{14}, h(1) = \sum_{x=0}^2 f(x, 1) = \frac{3}{7}$$

$$\implies f(0, 1) \neq g(0) * h(1)$$

therefore X and Y are not statistically independent.

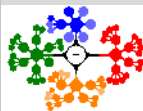
Joint Probability Distribution XIV

- **Example:** In a binary communications channel, let X denote the bit sent by the transmitter and let Y denote the bit received at the other end of the channel. Due to noise in the channel we do not always have $Y = X$. A joint probability distribution is given as



Joint Probability Distribution XIV

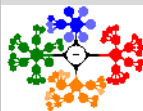
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		x		
		0	1	$h(y)$
y	0	0.45	0.03	0.48
	1	0.05	0.47	0.52
	$g(x)$	0.5	0.5	

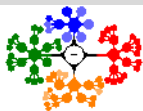


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Joint Probability Distribution XIV

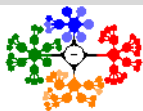
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$$f(0,0) \neq g(0)h(0) \implies 0.45 \neq 0.5 * 0.48$$



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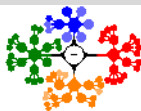
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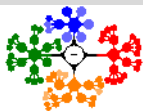
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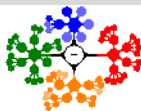
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- So $g(0) = P[X = 0]$
 $= P[(X = 0) \cap (Y = 0)] + P[(X = 0) \cap (Y = 1)] = f(0,0) + f(0,1)$



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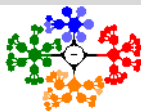
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y	0	$f(0,0)$	$f(1,0)$	$h(0)$
	1	$f(0,1)$	$f(1,1)$	$h(1)$
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$$f(0,0) \neq g(0)h(0) \implies 0.45 \neq 0.5 * 0.48$$

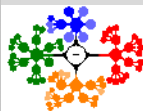
- $P(X = x, Y = y) = P[(X = x) \cap (Y = y)]$: it is the probability that $X = x$ and $Y = y$ simultaneously.
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- So $g(0) = P[X = 0]$
 $= P[(X = 0) \cap (Y = 0)] + P[(X = 0) \cap (Y = 1)] = f(0,0) + f(0,1)$
- $\implies P[Y = 0|X = 0] = \frac{P[(X=0) \cap (Y=0)]}{P[X=0]} = \frac{f(0,0)}{g(0)}$



Joint Probability Distribution XV

- Sent 0 & Received 0: NO error.

$$P[Y = 0|X = 0] = \frac{f(0, 0)}{g(0)} = \frac{0.45}{0.9} = 0.9$$



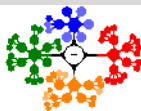
Joint Probability Distribution XV

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$$P[Y = 0|X = 0] = \frac{f(0, 0)}{g(0)} = \frac{0.45}{0.9} = 0.9$$

- Sent 1 & Received 0: ERROR

$$P[Y = 0|X = 1] = \frac{f(1, 0)}{g(1)} = \frac{0.03}{0.5} = 0.06$$



Joint Probability Distribution XV

- Sent 0 & Received 0: NO error.

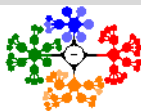
$$P[Y = 0|X = 0] = \frac{f(0, 0)}{g(0)} = \frac{0.45}{0.9} = 0.9$$

- Sent 1 & Received 0: ERROR

$$P[Y = 0|X = 1] = \frac{f(1, 0)}{g(1)} = \frac{0.03}{0.5} = 0.06$$

- Sent 0 & Received 1: ERROR

$$P[Y = 1|X = 0] = \frac{f(0, 1)}{g(0)} = \frac{0.05}{0.9} = 0.055$$



Joint Probability Distribution XV

- Sent 0 & Received 0: NO error.

$$P[Y = 0|X = 0] = \frac{f(0, 0)}{g(0)} = \frac{0.45}{0.9} = 0.9$$

- Sent 1 & Received 0: ERROR

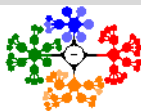
$$P[Y = 0|X = 1] = \frac{f(1, 0)}{g(1)} = \frac{0.03}{0.5} = 0.06$$

- Sent 0 & Received 1: ERROR

$$P[Y = 1|X = 0] = \frac{f(0, 1)}{g(0)} = \frac{0.05}{0.9} = 0.055$$

- Sent 1 & Received 1: NO error.

$$P[Y = 1|X = 1] = \frac{f(1, 1)}{g(1)} = \frac{0.47}{0.5} = 0.94$$



Joint Probability Distribution XV

- Sent 0 & Received 0: NO error.

$$P[Y = 0|X = 0] = \frac{f(0, 0)}{g(0)} = \frac{0.45}{0.9} = 0.9$$

- Sent 1 & Received 0: ERROR

$$P[Y = 0|X = 1] = \frac{f(1, 0)}{g(1)} = \frac{0.03}{0.5} = 0.06$$

- Sent 0 & Received 1: ERROR

$$P[Y = 1|X = 0] = \frac{f(0, 1)}{g(0)} = \frac{0.05}{0.9} = 0.0556$$

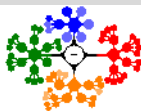
- Sent 1 & Received 1: NO error.

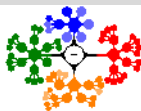
$$P[Y = 1|X = 1] = \frac{f(1, 1)}{g(1)} = \frac{0.47}{0.5} = 0.94$$

- Notice that

$$P[Y = 0|X = 0] + P[Y = 1|X = 0] = 1$$

$$P[Y = 0|X = 1] + P[Y = 1|X = 1] = 1$$





Definition 3.13:

Let X_1, X_2, \dots, X_n be n random variables, discrete or continuous, with joint probability distribution $f(x_1, x_2, \dots, x_n)$ and marginal distributions $f(x_1), f(x_2), \dots, f(x_n)$, respectively. The random variables X_1, X_2, \dots, X_n are said to be **mutually statistically independent** if and only if

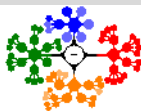
$$f(x_1, x_2, \dots) = f_1(x_1)f_2(x_2) \dots f_n(x_n)$$

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

Joint Probability Distribution XVII

- **Example 3.22:** Suppose that the shelf life, in years, of a certain perishable food product packaged in cardboard containers is a random variable whose probability density function is given by

$$f(x, y) = \left\{ \begin{array}{l} e^{-x}, x > 0 \\ 0, \textit{ elsewhere} \end{array} \right\}$$

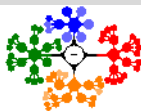


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- Let X_1, X_2, \dots, X_n represent the shelf lives for three of these containers selected independently and find $P(X_1 < 2, 1 < X_2 < 3, X_3 > 2)$



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

$$f(x_1, x_2, x_3) = f(x_1)f(x_2)f(x_3) = e^{-x_1 - x_2 - x_3}$$

for $x_1, x_2, x_3 > 0$ and $f(x_1, x_2, x_3) = 0$ elsewhere

$$\begin{aligned} P(X_1 < 2, 1 < X_2 < 3, X_3 > 2) &= \int_2^{\infty} \int_1^3 \int_0^2 e^{-x_1 - x_2 - x_3} dx_1 dx_2 dx_3 \\ &= (1 - e^{-2})(e^{-1} - e^{-3})e^{-2} = 0.0372 \end{aligned}$$

