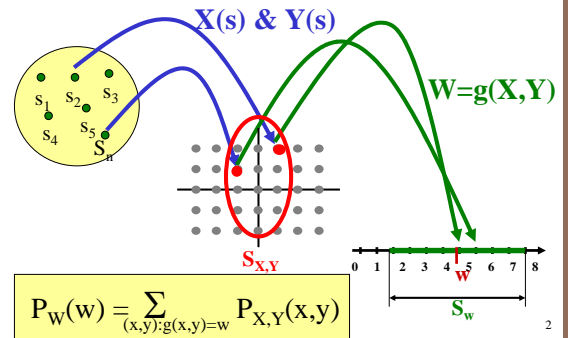


Multiple Discrete Random Variables (2)

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Derived Random Variable Functions of 2 RVs



Example

- Fax Sending (text-40sec & graphics-60sec)

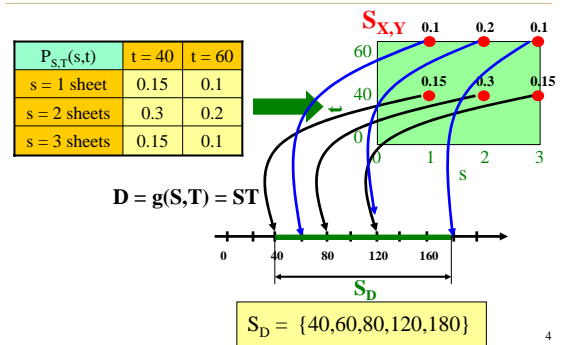
$P_{S,T}(s,t)$	$t = 40$	$t = 60$
$s = 1$ sheet	0.15	0.1
$s = 2$ sheets	0.3	0.2
$s = 3$ sheets	0.15	0.1

Let $D =$ duration for sending one fax
 $= g(S,T)$
 $= ST$

Find $P_D(d)$, S_D , and $E[D]$

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Example



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Example

$S_D = \{40, 60, 80, 120, 180\}$

$$P_D(d) = \sum_{(s,t):g(s,t)=d} P_{S,T}(s,t)$$

$P_D(d) =$	0.15	$d = 40$	$E[D] = \sum_{d \in S_D} P_D(d)$ $= 96 \text{ sec}$
	0.1	$d = 60$	
	0.3	$d = 80$	
	$0.15 + 0.2$	$d = 120$	
	0.1	$d = 180$	
	0	Otherwise	

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Expectations

- $E[W]$ for $W = g(X, Y)$
- $E[X+Y]$
- $\text{Var}[X+Y]$ (Variance)
- $\text{Cov}[X, Y]$ (Covariance)
- $r_{X,Y}$ (Correlation)
- $\rho_{X,Y}$ (Correlation Coefficient)

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Expected Value of $g(X,Y)$

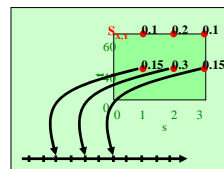
Theorem: for $W = g(X,Y)$

$$E[W] = \sum_{x \in S_X} \sum_{y \in S_Y} g(X,Y) P_{X,Y}(x,y)$$

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From Last Example

To find the $E[D]$, $D = g(S,T) = ST$



0.15	$d = 40$
0.1	$d = 60$
0.3	$d = 80$
0.15 + 0.2	$d = 120$
0.1	$d = 180$
0	Otherwise

$$E[D] = \sum_{d \in S_D} P_D(d)$$

Map $g(S,T) \rightarrow D$

Find $P_D(d)$

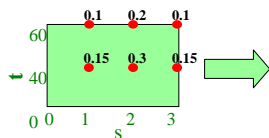
Find $E[D]$

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From Last Example

With the theorem, we can directly find $E[D]$

$$E[D] = \sum_{s=1}^3 \sum_{t=40,60} st P_{S,T}(s,t)$$



$$E[D] = 1*40*0.15 + 1*60*0.1 + 2*40*0.3 + 2*60*0.2 + 3*40*0.15 + 3*60*0.1 = 96 \text{ sec}$$

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For any 2 RVs

Theorem:

$$E[X + Y] = E[X] + E[Y]$$

- Find $E[X]$ and $E[Y]$
→ Marginal PMF

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Var[X+Y]

Definition: $\text{Var}[X] = E[(X - \mu_x)^2]$

$$\begin{aligned} \text{Var}[X+Y] &= E[((X+Y) - \mu_{x+y})^2] \\ &= E[((X+Y) - (\mu_x + \mu_y))^2] \\ &= E[(X - \mu_x + Y - \mu_y)^2] \\ &= E[(X - \mu_x)^2 + 2(X - \mu_x)(Y - \mu_y) + (Y - \mu_y)^2] \\ &= E[(X - \mu_x)^2] + 2E[(X - \mu_x)(Y - \mu_y)] + E[(Y - \mu_y)^2] \end{aligned}$$

Theorem:

$$\text{Var}[X+Y] = \text{Var}[X] + \text{Var}[Y] + 2E[(X - \mu_x)(Y - \mu_y)]$$

Covariance

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Covariance of X and Y

Definition: $\text{Cov}[X,Y] = E[(X - \mu_x)(Y - \mu_y)]$

$$\begin{aligned} \text{Cov}[X,Y] &= E[XY - \mu_x Y - \mu_y X + \mu_x \mu_y] \\ &= E[XY] - \mu_x E[Y] - \mu_y E[X] + \mu_x \mu_y \\ &= E[XY] - \mu_x \mu_y - \mu_x \mu_y + \mu_x \mu_y \end{aligned}$$

Theorem: $\text{Cov}[X,Y] = E[XY] - \mu_x \mu_y$

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Covariance of X and Y

Theorem: $\text{Cov}[X, Y] = E[XY] - \mu_x \mu_y$

Correlation

If $X = Y \rightarrow \text{Cov}[X, X] = E[XX] - \mu_x \mu_x$
 $= E[X^2] - \mu_x^2$
 $= E[X^2 - 2\mu_x X + \mu_x^2]$
 $= E[(X - \mu_x)^2]$
 $= \text{Var}[X]$

If $\mu_x = 0$ or $\mu_y = 0 \rightarrow \text{Cov}[X, Y] = E[XY]$

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Correlation

Definition: The correlation of X and Y is $r_{X,Y}$
 $r_{X,Y} = E[XY]$

Theorem: $\text{Cov}[X, Y] = r_{X,Y} - \mu_x \mu_y$

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More Definition

Definition 1:

X and Y are **Orthogonal** if $r_{X,Y} = 0$; $E[XY] = 0$

Definition 2:

X and Y are **Uncorrelated** if $\text{Cov}[X, Y] = 0$

Definition 3:

Correlation Coefficient of X and Y is

$$\rho_{X,Y} = \frac{\text{Cov}[X, Y]}{\sqrt{\text{Var}[X]\text{Var}[Y]}} = [-1, 1]$$

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Correlation Coefficient

- $\rho_{X,Y}$
 - Describes the info about X by observing Y
- $\rho_{X,Y} > 0$
 - If $X \uparrow$ (relative to mean) $\rightarrow Y \uparrow$
 - If $X \downarrow$ (relative to mean) $\rightarrow Y \downarrow$
- $\rho_{X,Y} < 0$
 - If $X \uparrow$ (relative to mean) $\rightarrow Y \downarrow$
 - If $X \downarrow$ (relative to mean) $\rightarrow Y \uparrow$
- Example:
 - X = student's height, Y = student's weight $\rho_{X,Y} > 0$
 - X = cell phone distance, Y = Rx signal strength $\rho_{X,Y} < 0$

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Uncorrelated

If X and Y are **Independent**, then

$\rightarrow \text{Cov}[X, Y] = 0 \rightarrow \rho_{X,Y} = 0$

\rightarrow X and Y are **Uncorrelated**

Note:

If X and Y are **Uncorrelated**,

\rightarrow X and Y **may or may not Independent**

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Conditional Joint PMF by an Event

$$P_{X,Y|B}(x,y) = \frac{P[(X=x, Y=y) \cap B]}{P[B]}$$

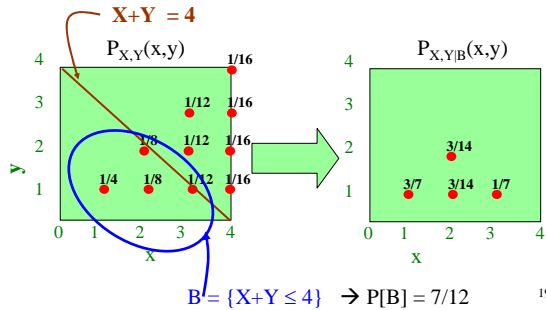
If $(X=x, Y=y) \in B \rightarrow (X=x, Y=y) \cap B = (X=x, Y=y)$

$$P_{X,Y|B}(x,y) = \begin{cases} \frac{P[(X=x, Y=y)]}{P[B]} & (x,y) \in B \\ 0 & \text{Otherwise} \end{cases}$$

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Example $P_{X,Y|B}(x,y)$

Let $B = \{X+Y \leq 4\}$ Find $P_{X,Y|B}(x,y)$



Conditional PMF

- Special case of Conditional Joint PMF by an Event
→ the Event is $X=x$ or $Y=y$
- $P_{X,Y|B}(x,y)$ when $B = \{Y=y\}$
→ $P_{X,Y|Y=y}(x,y) = P_{X|Y}(x|y)$

Definition: $P_{X|Y}(x|y) = P[X=x | Y=y]$

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Conditional PMF

$$\begin{aligned} P_{X|Y}(x|y) &= P[X=x | Y=y] \\ &= \frac{P[X=x, Y=y]}{P[Y=y]} \\ &= \frac{P_{X,Y}(x,y)}{P_Y(y)} \end{aligned}$$

Theorem:

$$P_{X,Y}(x,y) = P_{X|Y}(x|y)P_Y(y) = P_{Y|X}(y|x)P_X(x)$$

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Independent RVs

- From the independent definition
A and B are independent iff $P[AB] = P[A]P[B]$
- X and Y are independent RVs if and only if
 $\{X=x\}$ and $\{Y=y\}$ are independent for all x,y in $S_{X,Y}$

Definition: $P_{X,Y}(x,y) = P_X(x)P_Y(y)$

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Independent RVs

Theorem:

- $r_{X,Y} = E[XY] = E[X]E[Y]$
- $E[X|Y=y] = E[X]$ for all $y \in S_Y$
- $E[Y|X=x] = E[Y]$ for all $x \in S_X$
- $\text{Var}[X+Y] = \text{Var}[X] + \text{Var}[Y]$
- $\text{Cov}[X,Y] = \rho_{X,Y} = 0$

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More than 2 RVs

Definition: Joint PMF of discrete RV X_1, \dots, X_N is

$$P_{X_1, \dots, X_N}(x_1, \dots, x_N) = P[X_1=x_1, \dots, X_N=x_N]$$

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Homework

- | | |
|----------|----------|
| 1) 2.6.5 | 1) 3.1.6 |
| 2) 2.7.7 | 2) 3.2.5 |
| 3) 2.8.6 | 3) 3.3.5 |
| 4) 2.9.6 | 4) 3.4.6 |
| | 5) 3.5.2 |
| | 6) 3.6.3 |
| | 7) 3.6.6 |
| | 8) 3.7.8 |
| | 9) 3.8.3 |

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