

Lecture 6: Probabilities and distributions (Part 4), Statistical Independence of RV

Kateřina Staňková

Statistics (MAT1003)

April 24, 2012

Outline

- 1 **Conditional Probabilities**
- 2 **Conditional Probability Distributions**
 - Discrete
 - Continuous = Density
- 3 **Statistical Independence of RVs**

And now ...

- 1 **Conditional Probabilities**
- 2 **Conditional Probability Distributions**
 - Discrete
 - Continuous = Density
- 3 **Statistical Independence of RVs**

book: Section 2.6

Main Question

book: Section 2.6

Main Question

If we already know that an event A occurs, what is then the probability that an event B occurs?

book: Section 2.6

Main Question

If we already know that an event A occurs, what is then the probability that an event B occurs?

Example

$$S = \{1, 2, 3, 4\}, P(\{i\}) = \frac{i}{10}, i = 1, 2, 3, 4$$

book: Section 2.6

Main Question

If we already know that an event A occurs, what is then the probability that an event B occurs?

Example

$$S = \{1, 2, 3, 4\}, P(\{i\}) = \frac{i}{10}, i = 1, 2, 3, 4$$

$$A = \{1, 2\}, B = \{2, 3\}$$

book: Section 2.6

Main Question

If we already know that an event A occurs, what is then the probability that an event B occurs?

Example

$$S = \{1, 2, 3, 4\}, P(\{i\}) = \frac{i}{10}, i = 1, 2, 3, 4$$

$$A = \{1, 2\}, B = \{2, 3\}$$

$B|A$: The event B given A

book: Section 2.6

Main Question

If we already know that an event A occurs, what is then the probability that an event B occurs?

Example

$$S = \{1, 2, 3, 4\}, P(\{i\}) = \frac{i}{10}, i = 1, 2, 3, 4$$

$$A = \{1, 2\}, B = \{2, 3\}$$

$$B|A : \text{The event } B \text{ given } A = \{2\} = A \cap B$$

book: Section 2.6

Main Question

If we already know that an event A occurs, what is then the probability that an event B occurs?

Example

$$S = \{1, 2, 3, 4\}, P(\{i\}) = \frac{i}{10}, i = 1, 2, 3, 4$$

$$A = \{1, 2\}, B = \{2, 3\}$$

$$B|A : \text{The event } B \text{ given } A = \{2\} = A \cap B$$

But given that A occurs the sample space is actually reduced to $A = \{1, 2\}$

book: Section 2.6

Main Question

If we already know that an event A occurs, what is then the probability that an event B occurs?

Example

$$S = \{1, 2, 3, 4\}, P(\{i\}) = \frac{i}{10}, i = 1, 2, 3, 4$$

$$A = \{1, 2\}, B = \{2, 3\}$$

$B|A$: The event B given $A = \{2\} = A \cap B$

But given that A occurs the sample space is actually reduced to $A = \{1, 2\}$

$$P(B|A) = \frac{P(\{2\})}{P(\{1,2\})} = \frac{\frac{2}{10}}{\frac{3}{10}} = \frac{2}{3} \text{ (conditional probability)}$$

Conditional probability

Conditional probability

Probability of event B given A is

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

for any two events A and B , as long as $P(A) > 0$.

Conditional probability

Probability of event B given A is

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

for any two events A and B , as long as $P(A) > 0$.

Dependence of events

Conditional probability

Probability of event B given A is

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

for any two events A and B , as long as $P(A) > 0$.

Dependence of events

In the example we have $P(B) = \frac{1}{2}$, $P(B|A) = \frac{2}{3}$

Conditional probability

Probability of event B given A is

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

for any two events A and B , as long as $P(A) > 0$.

Dependence of events

In the example we have $P(B) = \frac{1}{2}$, $P(B|A) = \frac{2}{3}$

So apparently the probability that B occurs changes if we have the additional info that A occurs: $P(B) \neq P(B|A)$

Conditional probability

Probability of event B given A is

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

for any two events A and B , as long as $P(A) > 0$.

Dependence of events

In the example we have $P(B) = \frac{1}{2}$, $P(B|A) = \frac{2}{3}$

So apparently the probability that B occurs changes if we have the additional info that A occurs: $P(B) \neq P(B|A)$

A and B are **dependent**

Independence of RVs

Independence of RVs

A and B are called independent if

(a) $P(A|B) = P(A)$ or

(b) $P(B|A) = P(B)$

Independence of RVs

A and B are called independent if

(a) $P(A|B) = P(A)$ or

(b) $P(B|A) = P(B)$

(a) and (b) are equivalent if $P(A) > 0$, $P(B) > 0$

Independence of RVs

A and B are called independent if

(a) $P(A|B) = P(A)$ or

(b) $P(B|A) = P(B)$

(a) and (b) are equivalent if $P(A) > 0$, $P(B) > 0$

Since

- $P(A \cap B) = P(A) \cdot P(B|A)$ and

- $P(A \cap B) = P(B) \cdot P(A|B)$

A and B are independent $\Leftrightarrow P(A \cap B) = P(A) \cdot P(B)$

Independence of RVs

A and B are called independent if

(a) $P(A|B) = P(A)$ or

(b) $P(B|A) = P(B)$

(a) and (b) are equivalent if $P(A) > 0$, $P(B) > 0$

Since

- $P(A \cap B) = P(A) \cdot P(B|A)$ and

- $P(A \cap B) = P(B) \cdot P(A|B)$

A and B are independent $\Leftrightarrow P(A \cap B) = P(A) \cdot P(B)$

Next: Conditional Probability Distributions (book: Section 3.4)

And now ...

- 1 Conditional Probabilities
- 2 Conditional Probability Distributions**
 - Discrete
 - Continuous = Density
- 3 Statistical Independence of RVs



Conditional Probability Distribution

Let X and Y be discrete RVs with joint probability distribution f . The conditional probability of X given that Y has the value y is

$$f(x|y) = P(X = x|Y = y) = \frac{P(X=x, Y=y)}{P(Y=y)} = \frac{f(x,y)}{h(y)}$$

Conditional Probability Distribution

Let X and Y be discrete RVs with joint probability distribution f .
The conditional probability of X given that Y has the value y is

$$f(x|y) = P(X = x|Y = y) = \frac{P(X=x, Y=y)}{P(Y=y)} = \frac{f(x,y)}{h(y)}$$

Example

Throwing 2 dice, X : max, Y : min.

What is conditional probability distribution of X given $Y = 2$?

Conditional Probability Distribution

Let X and Y be discrete RVs with joint probability distribution f . The conditional probability of X given that Y has the value y is

$$f(x|y) = P(X = x|Y = y) = \frac{P(X=x, Y=y)}{P(Y=y)} = \frac{f(x,y)}{h(y)}$$

Example

Throwing 2 dice, X : max, Y : min.

What is conditional probability distribution of X given $Y = 2$?

(Check at home) $P(Y = 2) = \frac{1}{4}$,

$x = 3, 4, 5, 6 : f(x, y) = \frac{1}{18}$; $x = 2 : f(x, y) = \frac{1}{36}$

Conditional Probability Distribution

Let X and Y be discrete RVs with joint probability distribution f . The conditional probability of X given that Y has the value y is

$$f(x|y) = P(X = x|Y = y) = \frac{P(X=x, Y=y)}{P(Y=y)} = \frac{f(x,y)}{h(y)}$$

Example

Throwing 2 dice, X : max, Y : min.

What is conditional probability distribution of X given $Y = 2$?

(Check at home) $P(Y = 2) = \frac{1}{4}$,

$x = 3, 4, 5, 6 : f(x, y) = \frac{1}{18}$; $x = 2 : f(x, y) = \frac{1}{36}$

x	2	3, 4, 5, 6
$f(x, 2) = P(X = x Y = 2)$	$\frac{\frac{1}{36}}{\frac{1}{4}} = \frac{1}{9}$	$\frac{\frac{1}{18}}{\frac{1}{4}} = \frac{2}{9}$



Conditional Probability Distribution

Let X and Y be continuous RV's with joint probability density f .
The conditional probability of X given that Y has the value y is

Conditional Probability Distribution

Let X and Y be continuous RV's with joint probability density f . The conditional probability of X given that Y has the value y is

$$\begin{aligned}
 f(x|y) &= \lim_{(\delta, \epsilon) \downarrow (0, 0)} \frac{P(x \leq X \leq x + \delta | y \leq Y \leq y + \epsilon)}{\delta} \\
 &= \lim_{(\delta, \epsilon) \downarrow (0, 0)} \frac{P(x \leq X \leq x + \delta, y \leq Y \leq y + \epsilon)}{\delta \cdot P(y \leq Y \leq y + \epsilon)} \\
 &= \lim_{(\delta, \epsilon) \downarrow (0, 0)} \frac{P(x \leq X \leq x + \delta, y \leq Y \leq y + \epsilon)}{\delta \epsilon} \\
 &\quad \cdot \frac{\epsilon}{P(y \leq Y \leq y + \epsilon)} \\
 &= f(x, y) \cdot \frac{1}{h(y)} = \frac{f(x, y)}{h(y)}, \quad \text{provided } h(y) > 0
 \end{aligned}$$

Conditional Probability Distribution

Let X and Y be continuous RV's with joint probability density f .
The conditional probability of X given that Y has the value y is

$$\begin{aligned}
 f(x|y) &= \lim_{(\delta, \epsilon) \downarrow (0, 0)} \frac{P(x \leq X \leq x + \delta | y \leq Y \leq y + \epsilon)}{\delta} \\
 &= \lim_{(\delta, \epsilon) \downarrow (0, 0)} \frac{P(x \leq X \leq x + \delta, y \leq Y \leq y + \epsilon)}{\delta \cdot P(y \leq Y \leq y + \epsilon)} \\
 &= \lim_{(\delta, \epsilon) \downarrow (0, 0)} \frac{P(x \leq X \leq x + \delta, y \leq Y \leq y + \epsilon)}{\delta \epsilon} \\
 &\quad \cdot \frac{\epsilon}{P(y \leq Y \leq y + \epsilon)} \\
 &= f(x, y) \cdot \frac{1}{h(y)} = \frac{f(x, y)}{h(y)}, \quad \text{provided } h(y) > 0
 \end{aligned}$$

Same formula as for discrete case!

Example - from before

Let X and Y be continuous RV's with joint probability density f

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

Calculate conditional density of X , given $Y = \frac{1}{2}$ and calculate $P(X > \frac{1}{3} | Y = \frac{1}{2})$

Example - from before

Let X and Y be continuous RV's with joint probability density f

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

Calculate conditional density of X , given $Y = \frac{1}{2}$ and calculate $P(X > \frac{1}{3} | Y = \frac{1}{2})$

(Check at home) $g(x) = \frac{10}{3}x - \frac{10}{3}x^4, 0 < x < 1$ (0 elsewhere)

$h(y) = 5y^4, 0 < y < 1$ (0 elsewhere)

Example - from before

Let X and Y be continuous RV's with joint probability density f

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

Calculate conditional density of X , given $Y = \frac{1}{2}$ and calculate $P(X > \frac{1}{3} | Y = \frac{1}{2})$

(Check at home) $g(x) = \frac{10}{3}x - \frac{10}{3}x^4, 0 < x < 1$ (0 elsewhere)

$h(y) = 5y^4, 0 < y < 1$ (0 elsewhere)

$f(x|y) = \frac{10xy^2}{5y^4} = \frac{2x}{y^2}, 0 < x < y < 1$ (0 elsewhere)

Example - from before

Let X and Y be continuous RV's with joint probability density f

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

Calculate conditional density of X , given $Y = \frac{1}{2}$ and calculate $P(X > \frac{1}{3} | Y = \frac{1}{2})$

(Check at home) $g(x) = \frac{10}{3}x - \frac{10}{3}x^4, 0 < x < 1$ (0 elsewhere)

$h(y) = 5y^4, 0 < y < 1$ (0 elsewhere)

$f(x|y) = \frac{10xy^2}{5y^4} = \frac{2x}{y^2}, 0 < x < y < 1$ (0 elsewhere)

$f(x|\frac{1}{2}) = P(X = x, Y = \frac{1}{2}) = \left[\frac{2x}{y^2} \right]_{y=\frac{1}{2}} = 8x, 0 < x < \frac{1}{2}$ (0

elsewhere)

Example - from before

Let X and Y be continuous RV's with joint probability density f

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

Calculate conditional density of X , given $Y = \frac{1}{2}$ and calculate $P(X > \frac{1}{3} | Y = \frac{1}{2})$

(Check at home) $g(x) = \frac{10}{3}x - \frac{10}{3}x^4$, $0 < x < 1$ (0 elsewhere)

$h(y) = 5y^4$, $0 < y < 1$ (0 elsewhere)

$f(x|y) = \frac{10xy^2}{5y^4} = \frac{2x}{y^2}$, $0 < x < y < 1$ (0 elsewhere)

$f(x|\frac{1}{2}) = P(X = x, Y = \frac{1}{2}) = \left[\frac{2x}{y^2} \right]_{y=\frac{1}{2}} = 8x$, $0 < x < \frac{1}{2}$ (0

elsewhere)

$P(X > \frac{1}{3} | Y = \frac{1}{2}) = \int_{\frac{1}{3}}^{\frac{1}{2}} 8x dx = \left[4x^2 \right]_{x=\frac{1}{3}}^{\frac{1}{2}} = \frac{5}{9}$ (0 elsewhere)

And now ...

- 1 Conditional Probabilities
- 2 Conditional Probability Distributions
 - Discrete
 - Continuous = Density
- 3 **Statistical Independence of RVs**

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Example 1

X : maximum of eyes on 2 dice, Y : minimum

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Example 1

X : maximum of eyes on 2 dice, Y : minimum

- $f(3, 2) = P(X = 3, Y = 2) = \frac{1}{18}$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Example 1

X : maximum of eyes on 2 dice, Y : minimum

- $f(3, 2) = P(X = 3, Y = 2) = \frac{1}{18}$
- $g(3) = \frac{5}{36}$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Example 1

X : maximum of eyes on 2 dice, Y : minimum

- $f(3, 2) = P(X = 3, Y = 2) = \frac{1}{18}$
- $g(3) = \frac{5}{36}$
- $h(2) = \frac{9}{36} = \frac{1}{4}$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Example 1

X : maximum of eyes on 2 dice, Y : minimum

- $f(3, 2) = P(X = 3, Y = 2) = \frac{1}{18}$
- $g(3) = \frac{5}{36}$
- $h(2) = \frac{9}{36} = \frac{1}{4}$
- $f(3, 2) \neq g(3) \cdot h(2) \Rightarrow X$ and Y are dependent

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Example 2

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

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Example 2

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

- $g(x) = \frac{1}{x^2}, \quad \frac{1}{2} < x < 1$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Example 2

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

- $g(x) = \frac{1}{x^2}, \quad \frac{1}{2} < x < 1$
- $h(y) = \frac{y}{2}, \quad 0 < y < 2$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = g(x) \cdot h(y)$$

Example 2

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

- $g(x) = \frac{1}{x^2}, \quad \frac{1}{2} < x < 1$
- $h(y) = \frac{y}{2}, \quad 0 < y < 2$
- $f(x, y) = g(x) \cdot h(y) \Rightarrow X$ and Y are independent

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = h(x) \cdot g(y)$$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = h(x) \cdot g(y)$$

Example 3

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

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = h(x) \cdot g(y)$$

Example 3

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

- $g(x) = \frac{10}{3}x - \frac{10}{3}x^4, \quad 0 < x < 1$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = h(x) \cdot g(y)$$

Example 3

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

- $g(x) = \frac{10}{3}x - \frac{10}{3}x^4, \quad 0 < x < 1$
- $h(y) = 5y^4, \quad 0 < y < 1$

Definition of statistical independence

Let X and Y be RV's with joint PDF f and let g and h be the marginal distributions of X and Y , respectively. Then X and Y are called independent if

$$f(x, y) = h(x) \cdot g(y)$$

Example 3

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

- $g(x) = \frac{10}{3}x - \frac{10}{3}x^4, \quad 0 < x < 1$
- $h(y) = 5y^4, \quad 0 < y < 1$
- $f(x, y) \neq g(x) \cdot h(y) \Rightarrow X$ and Y are dependent

Example 4

Example 4

Bag 1: 3 red, 2 blue marbles,

Example 4

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

Example 4

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

E: Pick 2 marbles from bag 1 and 4 marbles from bag 2.

Example 4

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

E: Pick 2 marbles from bag 1 and 4 marbles from bag 2.

RVs: X : # reds from bag 1, Y : # reds from bag 2

Example 4

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

E: Pick 2 marbles from bag 1 and 4 marbles from bag 2.

RVs: X : # reds from bag 1, Y : # reds from bag 2

Are X and Y dependent?

Example 4

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

E: Pick 2 marbles from bag 1 and 4 marbles from bag 2.

RVs: X : # reds from bag 1, Y : # reds from bag 2

Are X and Y dependent? **No!** # reds drawn from bag 1 does in no way influence the # reds drawn from bag 2 and vice versa

Example 4

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

E: Pick 2 marbles from bag 1 and 4 marbles from bag 2.

RVs: X : # reds from bag 1, Y : # reds from bag 2

Are X and Y dependent? **No!** # reds drawn from bag 1 does in no way influence the # reds drawn from bag 2 and vice versa

$$g(x) = P(X = x) = \frac{\binom{3}{x} \binom{2}{2-x}}{\binom{5}{2}}, \quad x = 0, 1, 2$$

Example 4

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

E: Pick 2 marbles from bag 1 and 4 marbles from bag 2.

RVs: X : # reds from bag 1, Y : # reds from bag 2

Are X and Y dependent? **No!** # reds drawn from bag 1 does in no way influence the # reds drawn from bag 2 and vice versa

$$g(x) = P(X = x) = \frac{\binom{3}{x} \binom{2}{2-x}}{\binom{5}{2}}, \quad x = 0, 1, 2$$

$$h(y) = P(Y = y) = \frac{\binom{4}{y} \binom{5}{4-y}}{\binom{9}{4}}, \quad y = 0, 1, 2, 3, 4$$

Example 4 (cont.)

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

RVs: X : # reds from bag 1, Y : # reds from bag 2

Example 4 (cont.)

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

RVs: X : # reds from bag 1, Y : # reds from bag 2

$$g(x) = P(X = x) = \frac{\binom{3}{x} \binom{2}{2-x}}{\binom{5}{2}}, \quad x = 0, 1, 2$$

Example 4 (cont.)

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

RVs: X : # reds from bag 1, Y : # reds from bag 2

$$g(x) = P(X = x) = \frac{\binom{3}{x} \binom{2}{2-x}}{\binom{5}{2}}, \quad x = 0, 1, 2$$

$$h(y) = P(Y = y) = \frac{\binom{4}{y} \binom{5}{4-y}}{\binom{9}{4}}, \quad y = 0, 1, 2, 3, 4$$

Example 4 (cont.)

Bag 1: 3 red, 2 blue marbles, bag 2: 4 red, 5 blue marbles

RVs: X : # reds from bag 1, Y : # reds from bag 2

$$g(x) = P(X = x) = \frac{\binom{3}{x} \binom{2}{2-x}}{\binom{5}{2}}, \quad x = 0, 1, 2$$

$$h(y) = P(Y = y) = \frac{\binom{4}{y} \binom{5}{4-y}}{\binom{9}{4}}, \quad y = 0, 1, 2, 3, 4$$

$$f(x, y) = g(x) \cdot h(y) = \frac{\binom{3}{x} \binom{2}{2-x} \binom{4}{y} \binom{5}{4-y}}{\binom{5}{2} \binom{9}{4}}, \quad x = 0, 1, 2,$$

$$y = 0, 1, 2, 3, 4$$