School of Electronics Engineering



Example 7.1

Let *N* equals the number of bytes in email. A conditioning event might be the event *I* that the email contains an image.



Definition 7.1

Given the event B with P[B] > 0, the conditional CDF of X is

$$F_{X|B}(x) = P[X \le x \mid B]$$



Definition 7.2

Given the event B with P[B] > 0, the conditional PMF of X is

$$P_{X|B}(x) = P[X = x \mid B]$$



Definition 7.3

For a random variable X and an event B with P[B] > 0, the conditional PDF of X given B is

$$f_{X|B} = \frac{dF_{X|B}(x)}{dx}$$



Theorem 7.1

For a random variable X and an event $B \subset S_X$ with P[B] > 0, the conditional PDF of X given B is

Discrete:
$$P_{X|B}(x) = \begin{cases} \frac{P_X(x)}{P[B]}, & x \in B \\ 0, & \text{otherwise} \end{cases}$$

Continuous:
$$f_{X|B}(x) = \begin{cases} \frac{f_X(x)}{P[B]}, & x \in B \\ 0, & \text{otherwise} \end{cases}$$



Example 7.2

A website distributes videos, where the length of each video is modeled as a random variable X with the PMF of

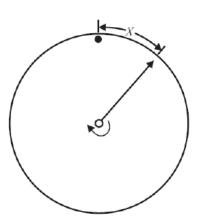
$$P_X(x) = \begin{cases} 0.15, & x = 1, 2, 3, 4 \\ 0.1, & x = 5, 6, 7, 8 \\ 0, & \text{otherwise} \end{cases}$$

Suppose the website has two servers, one for X < 5, the other one for $X \ge 5$. Find the PMF of video length in the 2nd server.



Example 7.3

For the pointer-spinning experiment of Example 4.1, find the conditional PDF of the pointer position for spins in which the pointer stops on the left side of the circle.





Example 7.4

Suppose *X*, the time in integer mins you wait for a bus, is modeled as a discrete (0, 20) uniform random variable. Suppose the bus has not arrived by the eight minute; what is the conditional PMF of your waiting time *X*?



Theorem 7.2

For random variable X resulting from an experiment with partition $B_1, \dots, B_{m'}$

Discrete: $P_X(x) = \sum_{i=1}^m P_{X|B_i}(x)P[B_i]$

Continuous: $f_X(x) = \sum_{i=1}^m f_{X|B_i}(x) P[B_i]$



Example 7.7

Random variable X is a voltage at the receiver of a modem. When symbol "0" is transmitted, X is the Gaussian (-5, 2) random variable. When symbol "1" is transmitted, X is the Gaussian (5, 2) random variable. What is the PDF of X?



Theorem 7.3

Discrete *X*:

- (a) For any $x \in B$, $P_{X|B}(x) \ge 0$
- (b) $\sum_{x \in B} P_{X|B}(x) = 1$
- (c) $P[C|B] = \sum_{x \in C} P_{X|B}(x)$

Continuous X:

- (a) For any $x \in B$, $f_{X|B}(x) \ge 0$
- (b) $\int_{B} f_{X|B}(x) dx = 1$
- (c) $P[C|B] = \int_C f_{X|B}(x) dx$



Definition 7.4

The conditional expected value of random variable *X* given condition B is

Discrete: $E[X|B] = \sum_{x \in B} x P_{X|B}(x)$

Continuous: $E[X|B] = \int_{-\infty}^{\infty} x f_{X|B}(x) dx$



Theorem 7.4

For a random variable X resulting from an experiment with partitions $B_1, ..., B_m$

$$E[X] = \sum_{i=1}^{m} E[X|B_i]P[B_i]$$



Theorem 7.5

The conditional expected value of Y = g(X) given condition B is

Discrete: $E[Y|B] = E[g(X)|B] = \sum_{x \in B} g(x)P_{X|B}(x)$

Continuous: $E[Y|B] = E[g(X)|B] = \int_{-\infty}^{\infty} g(x) f_{X|B}(x) dx$



Definition 7.5

The conditional variance of X given event B is

$$Var[X|B] = E[(X - \mu_{X|B})^2 | B] = E[X^2|B] - \mu_{X|B}^2$$



Definition 7.6

For discrete random variables X and Y and an event B with

P[B] > 0, the conditional joint PMF of X and Y given B is

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



Theorem 7.6

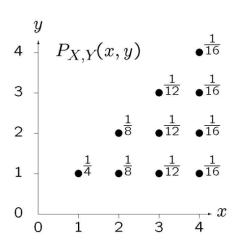
For any event B, a region of X, Y plane with P[B] > 0,

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



Example 7.9

If $B = \{X + Y \le 4\}$, find the conditional PMF $P_{X,Y|B}(x,y)$





Definition 7.7

Given an event B with P[B] > 0, the conditional joint PDF of X, Y is

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



Example 7.10

X and Y are random variables with joint PDF

$$f_{X,Y}(x,y) = \begin{cases} \frac{1}{15}, & 0 \le x \le 5, 0 \le y \le 3\\ 0, & \text{otherwise} \end{cases}$$

Find the conditional PDF of X, Y given the event $B = \{X + Y \ge 4\}$



Theorem 7.7

For random variables X and Y and an event B with P[B] > 0, the conditional expected value of W = g(X,Y) given B is

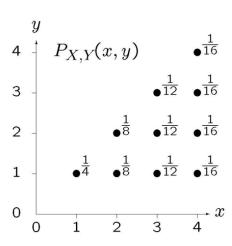
Discrete: $E[W|B] = \sum_{x \in S_X} \sum_{y \in S_Y} g(x, y) P_{X,Y|B}(x, y)$

Continuous: $E[W|B] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y) f_{X,Y|B}(x,y) dxdy$



Example 7.11

Find the conditional expected value and variance of W = X + Y given the event $B = \{X + Y \le 4\}$.





Definition 7.8

For any event Y = y such that $P_Y(y) > 0$, the conditional PMF of X given Y = y is

$$P_{X|Y}(x|y) = P[X = x \mid Y = y]$$



Theorem 7.8

For discrete random variables X and Y with joint PMF $P_{X,Y}(x,y)$, and x and y such that $P_X(x) > 0$ and $P_Y(y) > 0$,

$$P_{X|Y}(x|y) = \frac{P_{X,Y}(x,y)}{P_{Y}(y)}, \qquad P_{Y|X}(y|x) = \frac{P_{X,Y}(x,y)}{P_{X}(x)}$$



Definition 7.9

For y such that $f_Y(y) > 0$, the conditional PDF of X given $\{Y = y\}$ is

$$f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)}$$



Example 7.13

Random variables X and Y have joint PDF

$$f_{X,Y}(x,y) = \begin{cases} 2, & 0 \le y \le x < 1 \\ 0, & \text{otherwise} \end{cases}$$

For $0 \le x \le 1$, find the conditional PDF $f_{Y|X}(y|x)$



Theorem 7.9

For discrete random variables X and Y with joint PDF $P_{X,Y}(x,y)$, and x and y such that $P_X(x) > 0$ and $P_Y(y) > 0$, $P_{X,Y}(x,y) = P_{Y|X}(y|x)P_X(x) = P_{X|Y}(x|y)P_Y(y)$



Theorem 7.10

For continuous random variables X and Y with joint PDF $f_{X,Y}(x,y)$, and x and y such that $f_X(x) > 0$ and $f_Y(y) > 0$, $f_{X,Y}(x,y) = f_{Y|X}(y|x)f_X(x) = f_{X|Y}(x|y)f_Y(y)$



Example 7.14

Let R be the uniform (0,1) random variable. Given R=r, X is the uniform (0,r) random variable. Find the conditional PDF of R given X



Theorem 7.11

If X and Y are independent,

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

Continuous: $f_{X|Y}(x|y) = P_X(x)$, $f_{Y|X}(y|x) = f_Y(y)$



Definition 7.10

For any $y \in S_Y$, the conditional expected value of g(X,Y) given Y = y is

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

Continuous: $E[g(X,Y)|Y=y] = \int_{-\infty}^{\infty} g(x,y) f_{X|Y}(x|y) dx$



Theorem 7.12

For independent random variables X and Y,

(a)
$$E[X|Y = y] = E[X]$$
 for all $y \in S_Y$

(b)
$$E[Y|X = x] = E[Y]$$
 for all $x \in S_X$



Definition 7.11

The conditional expected value E[X|Y] is a function of random variable Y such that if Y = y, then E[X|Y] = E[X|Y = y]



Theorem 7.13 (Iterated expectation)

$$E\big[E[X|Y]\big] = E[X]$$

