9.1.4 Conditional Expectation (MMSE)

Remember that the posterior distribution, $f_{X|Y}(x|y)$, contains all the knowledge that we have about the unknown quantity X. Therefore, to find a point estimate of X, we can just choose a summary statistic of the posterior such as its mean, median, or mode. If we choose the mode (the value of x that maximizes $f_{X|Y}(x|y)$), we obtain the MAP estimate of X. Another option would be to choose the posterior mean, i.e.,

$$\hat{x} = E[X|Y = y].$$

We will show that E[X|Y=y] will give us the best estimate of X in terms of the *mean* squared error. For this reason, the conditional expectation is called the *minimum mean* squared error (MMSE) estimate of X. It is also called the *least mean squares* (LMS) estimate or simply the Bayes' estimate of X.

Minimum Mean Squared Error (MMSE) Estimation

The **minimum mean squared error (MMSE)** estimate of the random variable X, given that we have observed Y = y, is given by

$$\hat{x}_M = E[X|Y=y].$$

Example 9.6

Let X be a continuous random variable with the following PDF

$$f_X(x) = \left\{egin{array}{ll} 2x & & ext{if } 0 \leq x \leq 1 \ & & & ext{otherwise} \end{array}
ight.$$

We also know that

$$f_{Y|X}(y|x) = \left\{ egin{array}{ll} 2xy - x + 1 & \quad ext{if } 0 \leq y \leq 1 \\ 0 & \quad ext{otherwise} \end{array}
ight.$$

Find the MMSE estimate of X, given Y = y is observed.

Solution

First we need to find the posterior density, $f_{X|Y}(x|y)$. We have

$$f_{X|Y}(x|y) = rac{f_{Y|X}(y|x)f_X(x)}{f_Y(y)}.$$

We can find $f_Y(y)$ as

$$egin{aligned} f_Y(y) &= \int_0^1 f_{Y|X}(y|x) f_X(x) dx \ &= \int_0^1 (2xy - x + 1) 2x dx \ &= rac{4}{3} y + rac{1}{3}, \quad ext{for } 0 \leq y \leq 1. \end{aligned}$$

We conclude

$$f_{X|Y}(x|y)=rac{6x(2xy-x+1)}{4y+1}, \quad ext{ for } 0\leq x\leq 1.$$

The MMSE estimate of X given Y = y is then given by

$$egin{aligned} \hat{x}_M &= E[X|Y=y] \ &= \int_0^1 x f_{X|Y}(x|y) dx \ &= rac{1}{4y+1} \int_0^1 6x^2 (2xy-x+1) dx \ &= rac{3y+rac{1}{2}}{4y+1}. \end{aligned}$$