

Conditional Expectations (Section 5.11)

Conditional distributions can be used in the same way unconditional distributions are used. For example, conditional distributions may be used to compute conditional expectations.

Definition 5.13: Conditional Expectations

If Y_1 and Y_2 are two random variables and $g(Y_1)$ is a function of Y_1 , then the **conditional expectation** of $g(Y_1)$ given that $Y_2 = y_2$ is:

$$E[g(Y_1) | Y_2 = y_2] = \begin{cases} \sum_{\text{all } y_1} g(y_1) \times p(y_1 | y_2) & ; \text{ if } Y_1 \text{ and } Y_2 \text{ are DISCRETE} \\ \int_{-\infty}^{\infty} g(y_1) \times f(y_1 | y_2) dy_1 & ; \text{ if } Y_1 \text{ and } Y_2 \text{ are CONTINUOUS} \end{cases}$$

In particular, the **conditional mean** of Y_1 given that $Y_2 = y_2$ is denoted by and computed using

$$\mu_{Y_1 | Y_2} = E(Y_1 | Y_2 = y_2)$$

and the **conditional variance** of Y_1 given that $Y_2 = y_2$ is denoted by and computed as

$$\text{Var}(Y_1 | Y_2 = y_2) = E[(Y_1 - \mu_{Y_1 | Y_2})^2 | Y_2 = y_2]$$

Further, the shortcut formula for $\text{Var}(Y_1 | Y_2 = y_2)$:

$$\text{Var}[Y_1 | Y_2 = y_2] = E[Y_1^2 | Y_2 = y_2] - (\mu_{Y_1 | Y_2})^2$$

Example 7: Consider the joint pdf

$$f_{Y_1, Y_2}(y_1, y_2) = e^{-y_2} ; \quad 0 < y_1 < y_2 < \infty$$

Previously, we showed the conditional distribution of $Y_1 | Y_2$ is

$$f(y_1 | y_2) = \begin{cases} \frac{1}{y_2} ; & 0 < y_1 < y_2 < \infty \\ 0 ; & \text{otherwise} \end{cases}$$

Find $E(Y_1 | Y_2 = y_2)$.

Here Y_1 is a function only Y_1 .

$$\begin{aligned} E(Y_1 | Y_2 = y_2) &= \int_{-\infty}^{\infty} y_1 f(y_1 | y_2) dy_1 \\ &= \int_0^{y_2} y_1 \left(\frac{1}{y_2}\right) dy_1 = \frac{y_2}{2} \end{aligned}$$

$$\Rightarrow E(Y_1 | Y_2 = 5) = \frac{5}{2}$$

$$\text{Let } E(Y_1^2 | Y_2 = y_2) = \int_{-\infty}^{\infty} y_1^2 f(y_1 | y_2) dy_1$$