Conditional expectations

- ► Let X and Y be random variables such that E(X) and E(Y) exist and are finite.
- The conditional expectation (conditional mean) of Y given that X = x is defined as the expected value of the conditional distribution of Y given that X = x.

$$E(Y | X = x) = \int_{-\infty}^{\infty} yg_2(y | x)dy \quad \text{continuous case}$$
$$E(Y | X = x) = \sum_{\text{All } y} yg_2(y | x) \quad \text{discrete case}$$

Recall

$$g_2(y \mid x) = \frac{f(x, y)}{f_1(x)}$$
$$g_2(y \mid x) = \frac{\Pr(X = x, Y = y)}{\Pr(X = x)}$$

continuous case

discrete case

Conditional expectations

Similarly

$$E(X | Y = y) = \int_{-\infty}^{\infty} xg_1(x | y) dx \quad \text{continuous case}$$
$$E(X | \mathbf{x} = \mathbf{y}) = \sum_{\text{All } x} xg_1(x | y) \quad \text{discrete case}$$

Recall

$$g_2(x \mid y) = \frac{f(x, y)}{f_2(y)} \quad \text{continuous case}$$
$$g_2(x \mid y) = \frac{\Pr(X = x, Y = y)}{\Pr(Y = y)} \quad \text{discrete case}$$

Suppose that a point X is chosen in accordance with the uniform distribution on the interval [0, 1]. Also, suppose that after the value X = x has been observed (0 < x < 1), a point Y is chosen in accordance with a uniform distribution on the interval [x, 1]. Determine the value of E(Y).

$$X \sim \bigcup ni \text{ form } (0, 1) \qquad f_1(x) = \begin{cases} 1 & x \in (0, 1) \\ 0 & 0/w \end{cases}$$

$$Y \mid X = x \qquad \sim \bigcup ni \text{ form } (x, 1) \qquad g_2(y \mid x) = \begin{cases} \frac{1}{1-x} & x < y < 1 \\ 0 & 0/w \end{cases}$$

$$f(x_1 y) = f(x) \quad g_2(y \mid x) = \begin{cases} \frac{1}{1-x} & 0 < x < y < 1 \\ 0 & 0/w \end{cases}$$

$$E(y) = \int_{-\infty}^{\infty} y \cdot f_2(y) \, dy$$

$$\begin{aligned} f_{z}(y) &= \int_{-\infty}^{\infty} f(x_{1}y) dx &= \int_{0}^{y} \frac{1}{1-x} dx = -\log(1-x) \Big|_{0}^{y} & \log f(y) \\ &= -\log(1-y) = \log \frac{1}{1-y} & y \in (0,1) \\ &= \int_{-\infty}^{\infty} y f_{z}(y) dy = \int_{0}^{1} \frac{y}{1-y} \log \frac{1}{1-y} dy = \frac{3}{4} (\text{please checl.} !) \end{aligned}$$



Consider a clinical trial in which a number of patients will be treated and each patient will have one of two possible outcomes: success or failure. Let P be the proportion of successes in a very large collection of patients, and let $X_i = 1$ if the ith patient is a success and $X_i = 0$ if not. Assume that the random variables X_1, X_2, \ldots are conditionally independent given P = p with

$$\Pr(X_i = 1 | P = p) = p$$

Let $X = X_1 + ... + X_n$, which is the number of patients out of the first *n* who are successes. Find the conditional mean of X given P = p.

$$E(X | P=P) = E(X_1 + X_2 + ... + X_n | P=P)$$

= $E(X_1 | P=P) + E(X_2 | P=P) + ... + E(X_n | P=P)$
= $P + P + ... + P = ^n P$

Suppose that X and Y have a continuous joint distribution for which the joint pdf is as follows:

$$f(x,y) = \begin{cases} x+y & \text{if } 0 \le x \le 1 \text{ } 0 \le y \le 1, \\ 0 & \text{otherwise} \end{cases}$$

Find E(Y | X = x) and Var(Y | X = x).

$$E(Y|X=x) = \int_{-\infty}^{\infty} y \cdot g_{1}(y|x) dy$$

$$g_{2}(y|x) = \frac{f(x,y)}{f_{1}(x)} = \frac{x \cdot y}{x \cdot \frac{1}{2}} \quad y \in [o_{1}]$$

$$f_{1}(x) = \int_{-\infty}^{\infty} f(x,y) dy = \int_{0}^{1} (x \cdot y) dy = x \cdot \frac{1}{2}$$

$$E(Y|X=x) = \int_{0}^{1} y \cdot \frac{x+y}{x+\frac{1}{2}} dy = \int_{0}^{1} \frac{x}{x+\frac{1}{2}} \cdot y + \frac{1}{x+\frac{1}{2}} \cdot y^{2} dy = \frac{x}{x+\frac{1}{2}} \cdot \frac{y^{2}}{2} \Big|_{0}^{1} + \frac{1}{x+\frac{1}{2}} \cdot \frac{y^{3}}{2} \Big|_{0}^{1}$$
$$= \frac{x}{x+\frac{1}{2}} \cdot \frac{1}{2} + \frac{1}{x+\frac{1}{2}} \cdot \frac{1}{3} = A$$

$$Var(Y|X=x) = Vaniance of the conditional distribution of Y given $X=x$
= $E(Y^2|X=x) - E(Y|X=x)^2 = B - A^2$$$

$$E(\gamma^{2}|X=x) = \int_{0}^{1} y^{2} \cdot \frac{x_{f}y}{x_{f}\frac{1}{2}} \, dy = \frac{x}{x_{f}\frac{1}{2}} \cdot \frac{y^{3}}{3} \Big|_{0}^{1} + \frac{1}{x_{f}\frac{1}{2}} \cdot \frac{y^{4}}{4} \Big|_{0}^{1} = \frac{x}{x_{f}\frac{1}{2}} \cdot \frac{1}{3} + \frac{1}{x_{f}\frac{1}{2}} \cdot \frac{1}{4} = B$$

Functions of two or more RV

Assume that Z = r(X, Y) is a random variable where the joint distribution of X, Y is described via the joint pmf/pdf f(x, y). One can evaluate the conditional expectation E(Z | X = x) in the following way

$$E(Z | X = x) = E(r(X, Y) | X = x)$$

= $E(r(x, Y) | X = x)$
= $\int_{-\infty}^{\infty} r(x, y)g_2(y | x) dy$

Suppose that the joint distribution of (X, Y) is uniform over the unit circle. Let $Z = X^2 + Y^2$. Find E(Z | Y = -0.2).

$$E(2|Y=y) = E(X^{2}+Y^{2}|Y=y)$$
$$= \int_{-\infty}^{\infty} (x^{2}+y^{2}) g_{2}(x|y) dx$$
$$= \int_{-\infty}^{\infty} (x,y) dx$$

$$g_{z}(x|y) = \frac{f(x,y)}{f_{z}(y)}$$

$$f(x_{1}y) = \int_{-\infty}^{\infty} f(x_{1}y) = \begin{cases} \frac{1}{11} & x^{\frac{1}{2}y^{\frac{1}{2}} \leq 1} & (\text{uniform} \Rightarrow \frac{1}{\alpha x_{n}}) \\ & & & \\ &$$

Prediction

Suppose X, Y are random variables and the goal is to predict one of them, say Y. Let d be the predicted value.

 If no information is available, the prediction that minimizes the mean squared error (MSE)

$$MSE = E((Y - d)^2)$$

among all possible d values, is d = E(Y).

The prediction that minimizes the mean absolute error (MAE)

$$MAE = E(|\mathbf{Y} - d|) = E(|\mathbf{Y} - d|)$$

among all possible d values is $d = \text{median}(\mathbf{y})$.

median (Y)







choose the smallest x : F(x) = g