

35. Properties of Expectations

Conditional Variance [Ross S7.5.4]

So far, we have defined expectation, variance, and conditional expectation. We now define the **conditional variance**

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

So,

$$\begin{aligned} E[\text{Var}[X|Y]] &= E[E[X^2|Y]] - E[(E[X|Y])^2] \\ &= E[X^2] - E[(E[X|Y])^2] \end{aligned} \quad (35.1)$$

Also, $E[X|Y] = g(Y)$ for some function g , so

$$\begin{aligned} \text{Var}[g(Y)] &= E[(g(Y))^2] - (E[g(Y)])^2 \\ \text{Var}[E[X|Y]] &= E[(E[X|Y])^2] - (E[E[X|Y]])^2 \\ &= E[(E[X|Y])^2] - (E[X])^2 \end{aligned} \quad (35.2)$$

Adding (35.1) to (35.2), we get

Proposition 35.1 *Conditional Variance Formula:*

$$\begin{aligned} E[\text{Var}[X|Y]] + \text{Var}[E[X|Y]] &= E[X^2] - (E[X])^2 \\ &= \text{Var}[X] \end{aligned}$$

Example 35.1: Let X_1, X_2, \dots be iid and independent of the non-negative

integer random variable N . Let's compute

$$Var \left[\sum_{i=1}^N X_i \right]$$

by conditioning on N .

Using $Var[X] = E[Var[X|Y]] + Var[E[X|Y]]$ with

$$\begin{aligned} X &= \sum_{i=1}^N X_i \\ Y &= N \end{aligned}$$

$$\begin{aligned} \text{then } E \left[\sum_{i=1}^N X_i \middle| N = n \right] &= E \left[\sum_{i=1}^n X_i \right] && \text{[Since } N \text{ is ind. of the } X_i] \\ &= nE[X_1] && \text{[Since } X_i \text{ are iid]} \end{aligned}$$

$$\Rightarrow E \left[\sum_{i=1}^N X_i \middle| N \right] = NE[X_1]$$

$$\begin{aligned} Var \left[\sum_{i=1}^N X_i \middle| N = n \right] &= Var \left[\sum_{i=1}^n X_i \right] && \text{[Since } N \text{ is ind. of the } X_i] \\ &= nVar[X_1] && \text{[Since } X_i \text{ are iid]} \end{aligned}$$

$$\Rightarrow Var \left[\sum_{i=1}^N X_i \middle| N \right] = NVar[X_1]$$

By the conditional variance formula:

$$\begin{aligned} \text{Var} \left[\sum_{i=1}^N X_i \right] &= E [N \text{Var}[X_1]] + \text{Var} [N E[X_1]] \\ &= E[N] \text{Var}[X_1] + (E[X_1])^2 \text{Var}[N] \end{aligned}$$