

## Properties of Expectations

### Conditional Variance [Ross S7.5.4]

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

$$Var[X|Y] = E[X^2|Y] - (E[X|Y])^2$$

So,

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

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

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

Adding (35.1) to (35.2), we get

**Proposition 35.1** *Conditional Variance Formula:*

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

**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

$$X = \sum_{i=1}^N X_i$$
$$Y = N$$

then  $E \left[ \sum_{i=1}^N X_i \middle| N = n \right] = E \left[ \sum_{i=1}^n X_i \right]$  [Since  $N$  is ind. of the  $X_i$ ]

$= nE[X_1]$  [Since  $X_i$  are iid]

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

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

[Since  $N$  is ind. of the  $X_i$ ]

$$= nVar[X_1]$$

[Since  $X_i$  are iid]

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

By the conditional variance formula:

$$Var \left[ \sum_{i=1}^N X_i \right] = E[ NVar[X_1] ] + Var[ NE[X_1] ]$$
$$= E[N]Var[X_1] + (E[X_1])^2 Var[N]$$