

LECTURE 13: Conditional expectation and variance revisited;

Application: Sum of a random number of independent r.v.'s

- A more abstract version of the conditional expectation
 - view it as a random variable
 - the law of iterated expectations
- A more abstract version of the conditional variance
 - view it as a random variable
 - the law of total variance
- Sum of a random number of independent r.v.'s
 - mean
 - variance

Conditional expectation as a random variable

- Function h
e.g., $h(x) = x^2$, for all x
- Random variable X ; what is $h(X)$?
 $= x^2$
- $h(X)$ is the r.v. that takes the value x^2 , if X happens to take the value x
- $\underline{g(y)} = \mathbf{E}[X \mid Y = y] = \sum_x x p_{X|Y}(x \mid y)$
(integral in continuous case)
- $g(Y)$: is the r.v. that takes the value $\mathbf{E}[X \mid Y = y]$, if Y happens to take the value y
- Remarks:
 - It is a function of Y
 - It is a random variable
 - Has a distribution, mean, variance, etc.

Definition: $\underline{\mathbf{E}[X|Y]} = g(Y)$

The mean of $E[X | Y]$: Law of iterated expectations

- $g(y) = E[X | Y = y]$

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

$$E[X | Y] \triangleq g(Y)$$

$$E[\underbrace{E[X | Y]}] = E[g(Y)]$$

$$= \sum_Y g(y) P_Y(y)$$

exp. value rule

$$= \sum_Y E[X | Y = y] P_Y(y)$$

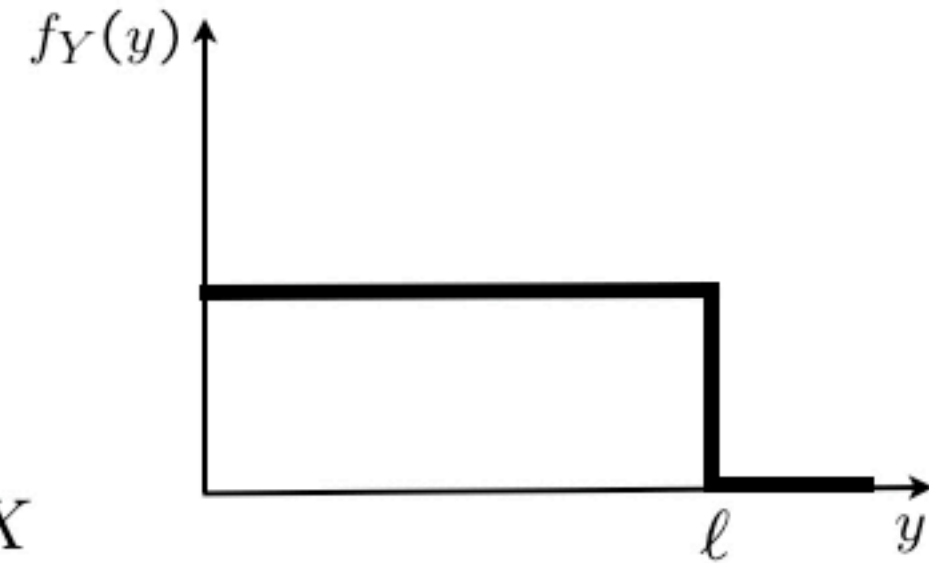
• total exp thm

$$= E[X]$$

Stick-breaking example

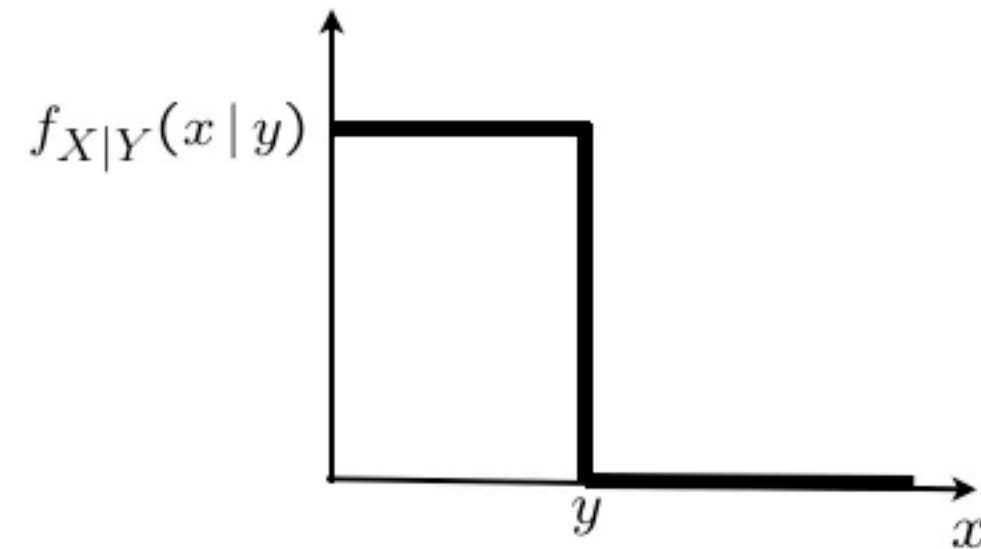
- Stick example: stick of length ℓ
break at uniformly chosen point Y

break what is left at uniformly chosen point X



- $E[X | Y = y] = y/2$

- $E[X | Y] = Y/2$



$$E[X] = E[E[X|Y]] = E[Y/2] = \frac{1}{2} E[Y] = \frac{1}{2} \cdot \frac{\ell}{2} = \frac{\ell}{4}$$

Forecast revisions

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

- Suppose forecasts are made by calculating expected value, given any available information

- X : February sales



- Forecast in the beginning of the year: $E[X]$
- End of January: will get new information, value y of Y

Revised forecast: $E[X | Y = y]$ $E[X | Y]$

- Law of iterated expectations:

$$E[\text{revised forecast}] = E[X] = \text{original forecast}$$

The conditional variance as a random variable

$$\text{var}(X) = \mathbf{E}[(X - \mathbf{E}[X])^2]$$

$$\text{var}(X \mid Y = \underline{y}) = \mathbf{E}[(X - \underline{\mathbf{E}[X \mid Y = y]})^2 \mid Y = y]$$

$\text{var}(X \mid Y)$ is the r.v. that takes the value $\text{var}(\bar{X} \mid Y = y)$, when $Y = y$

- Example: X uniform on $[0, Y]$

$$\text{var}(X \mid Y = y) = y^2/12$$

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Law of total variance: $\text{var}(X) = \mathbf{E}[\text{var}(X \mid Y)] + \text{var}(\mathbf{E}[X \mid Y])$

Derivation of the law of total variance

$$\text{var}(X) = \mathbf{E}[\text{var}(X | Y)] + \text{var}(\mathbf{E}[X | Y])$$

$$\bullet \text{var}(X) = \mathbf{E}[X^2] - (\mathbf{E}[X])^2$$

$$\text{var}(X | Y = y) = E[X^2 | Y = y] - (E[X | Y = y])^2 \text{ for all } y$$

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

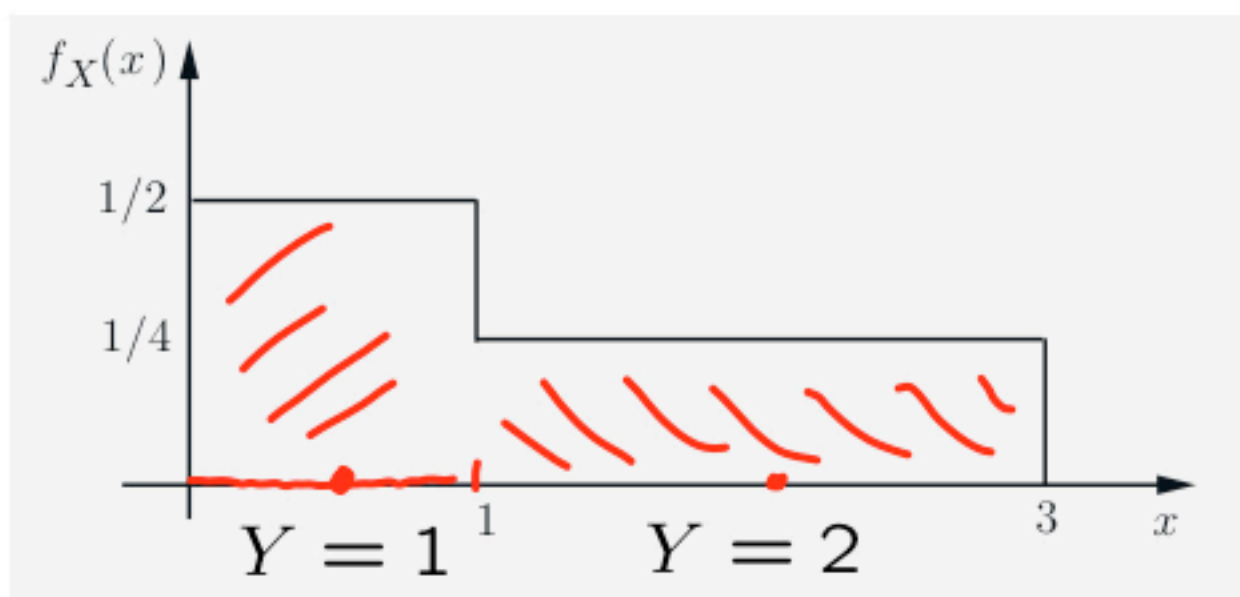
$$\mathbf{E}[\text{var}(X | Y)] = E[X^2] - E[(E[X | Y])^2]$$

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

A simple example

$$\text{var}(X) = \mathbf{E}[\text{var}(X | Y)] + \text{var}(\mathbf{E}[X | Y]) = \frac{37}{48}$$

$$= \frac{5}{24} + \frac{9}{16}$$



$$\text{var}(X | Y) = \begin{cases} \frac{1}{12} & \text{var}(X | Y = 1) = \frac{1}{12} \\ \frac{1}{12} & \text{var}(X | Y = 2) = \frac{2^2}{12} = \frac{4}{12} \end{cases}$$

$$\mathbf{E}[\text{var}(X | Y)] = \frac{1}{2} \cdot \frac{1}{12} + \frac{1}{2} \cdot \frac{4}{12} = \frac{5}{24}$$

$$\mathbf{E}[X | Y] = \begin{cases} \frac{1}{2} & \mathbf{E}[X | Y = 1] = \frac{1}{2} \\ 2 & \mathbf{E}[X | Y = 2] = 2 \end{cases}$$

$$\text{var}(\mathbf{E}[X | Y]) = \frac{1}{2} \left(\frac{1}{2} - \frac{5}{4} \right)^2 + \frac{1}{2} \left(2 - \frac{5}{4} \right)^2 = \frac{9}{16}$$

$$\mathbf{E}[\mathbf{E}[X | Y]] = \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot 2 = \frac{5}{4} = \mathbf{E}[X]$$

Section means and variances

- Two sections of a class: $y = 1$ (10 students); $y = 2$ (20 students)

x_i : score of student i

- Experiment: pick a student at random (uniformly)

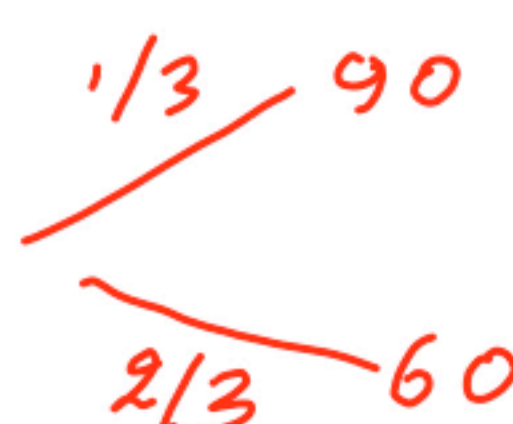
random variables: X and Y

- Data: $y = 1 : \frac{1}{10} \sum_{i=1}^{10} x_i = 90$ $y = 2 : \frac{1}{20} \sum_{i=11}^{30} x_i = 60$

- $E[X] = \frac{1}{30} \sum_{i=1}^{30} x_i = \frac{1}{30} (90 \cdot 10 + 60 \cdot 20) = 70$

$E[X | Y = 1] = 90$

$E[X | Y = 2] = 60$

$E[X | Y] =$ 

- $E[E[X | Y]] = \frac{1}{3} \cdot 90 + \frac{2}{3} \cdot 60 = 70$

Section means and variances (ctd.)

$$\mathbf{E}[X | Y] = \begin{cases} 90, & \text{w.p. } 1/3 \\ 60, & \text{w.p. } 2/3 \end{cases} \quad \mathbf{E}[\mathbf{E}[X | Y]] = 70 = \mathbf{E}[X]$$
$$\text{var}(\mathbf{E}[X | Y]) = \frac{1}{3} (90 - 70)^2 + \frac{2}{3} (60 - 70)^2 = 200$$

- More data: $\frac{1}{10} \sum_{i=1}^{10} (x_i - 90)^2 = 10$ $\frac{1}{20} \sum_{i=11}^{30} (x_i - 60)^2 = 20$

$$\text{var}(X | Y = 1) = 10$$

$$\text{var}(X | Y) = \frac{1/3}{2/3} \begin{matrix} 10 \\ 20 \end{matrix}$$

$$\text{var}(X | Y = 2) = 20$$

$$\mathbf{E}[\text{var}(X | Y)] = \frac{1}{3} \cdot 10 + \frac{2}{3} \cdot 20 = \frac{50}{3}$$

$$\text{var}(X) = \mathbf{E}[\text{var}(X | Y)] + \text{var}(\mathbf{E}[X | Y]) = 50/3 + 200$$

$\text{var}(X) = (\text{average variability **within** sections}) + (\text{variability **between** sections})$

Sum of a random number of independent r.v.'s

$$\mathbf{E}[Y] = \mathbf{E}[N] \cdot \mathbf{E}[X]$$

- N : number of stores visited
(N is a nonnegative integer r.v.)
- Let $Y = X_1 + \dots + X_N$
- X_i : money spent in store i
 - X_i independent, identically distributed
 - independent of N

$$\begin{aligned}\mathbf{E}[Y | N = n] &= E[X_1 + \dots + X_n | N = n] = E[X_1 + \dots + X_n | N = n] \\ &= E[X_1 + \dots + X_n] = n E[X]\end{aligned}$$

$\hookrightarrow E[Y|N] = NE[X]$

- Total expectation theorem:

$$\mathbf{E}[Y] = \sum_n p_N(n) \mathbf{E}[Y | N = n] = \sum_n p_N(n) n E[X] = E[N] E[X]$$

- Law of iterated expectations:

$$\mathbf{E}[Y] = \mathbf{E}[\mathbf{E}[Y | N]] = E[N E[X]] = E[N] E[X]$$

Variance of sum of a random number of independent r.v.'s

$$Y = X_1 + \cdots + X_N$$

•

$$\text{var}(Y) = \mathbf{E}[\text{var}(Y | N)] + \text{var}(\mathbf{E}[Y | N])$$

- $\mathbf{E}[Y | N] = N \mathbf{E}[X]$

$$\text{var}(Y) = \mathbf{E}[N] \text{var}(X) + (\mathbf{E}[X])^2 \text{var}(N)$$

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

- $\text{var}(Y | N = n) = \text{var}(X_1 + \cdots + X_n | N = n) = \text{var}(X_1 + \cdots + X_n) = n \text{var}(X)$
 $\text{var}(Y | N) = N \text{var}(X)$

- $\mathbf{E}[\text{var}(Y | N)] = \mathbf{E}[N \text{var}(X)] = \mathbf{E}[N] \text{var}(X)$