

Intro Prob Lecture Notes

William Sun

April 26, 2017

Conditional Expectation

- Law of total expectation $E(X) = E(E(X|Y))$
 - Example: random variable N $p(N = n) = p_n$ for $n = 0, 1, 2, 3, \dots$ x_1, x_2, x_3, \dots i.i.d. and independent of N

$$* \rightarrow E(S) = \mu_x \mu_N$$

$$\begin{aligned} - \text{Var}(S) &= E(S^2) - (E(S))^2 = E(S^2) - \mu_x^2 \mu_N^2 \\ * \left(\sum_{i=1}^N x_i \right) \left(\sum_{j=1}^N x_j \right) &= \sum_{i=1}^N \sum_{j=1}^N x_i x_j = \sum_{i=1}^N x_i^2 \sum_{i \neq j}^N x_i x_j \end{aligned}$$

—

$$\begin{aligned} E(S^2) &= E(E(S^2|N)) \\ &= E\left(\sum_{i=1}^N x_i^2 \sum_{i \neq j}^N x_i x_j \mid N = n\right) \\ &= E\left(\sum_{i=1}^N x_i^2 \sum_{i \neq j}^N x_i x_j\right) \\ &= nE(x_1^2) + n(n-1)E(x_1 x_2) \\ &= n(\sigma_X^2 + \mu_X^2) + n(n-1)E(x_1)E(x_2) \\ &= n\sigma_X^2 + n\mu_X^2 + n(n-1)\mu_x^2 \\ &= n\sigma_X^2 + n^2\mu_X^2 \end{aligned}$$

—

$$\begin{aligned}
E(S^2) &= E(E(S^2|N)) \\
&= E(n\sigma_X^2 + n^2\mu_X^2) \\
&= \sum_{n=0}^{\infty} (n\sigma_X^2 + n^2\mu_X^2)P_N(n) &= \sigma_X^2 \sum_{n=0}^{\infty} nP_N(n) + \mu_X^2 \sum_{n=0}^{\infty} n^2P_N(n) \\
&= \sigma_X^2\mu_N + \sigma_X^2E(N^2)
\end{aligned}$$

—

$$Var(S) = \sigma_X^2\mu_N + \mu_X^2E(N^2) - \mu_X^2\mu_N^2 = \sigma^2\mu_N + \mu_X^2\sigma_N^2$$

Function with another variable

- Let X, Y r.v.'s and h is any function. Then

$$E(Xh(Y)|Y) = h(Y)E(X|Y)$$

or

$$E(Xh(Y)|Y = y) = h(y)E(X|Y = y)$$

- Proof:
-

$$\begin{aligned}
E(Xh(Y)|Y = y) &= \int_{-\infty}^{\infty} xh(y)f_{X|Y}(x, y)dx \\
&= h(y) \int_{-\infty}^{\infty} xf_{X|Y}(x|y)dx \\
&= h(y)E(X|Y = y)
\end{aligned}$$

•

$$\begin{aligned} E(g(X)h(Y)) &= E(E(g(X)h(Y)|Y)) \\ &= E(h(Y)E(g(X)|Y)) \end{aligned}$$

- Ex: $X \sim \text{geometric}(p)$

- Let Y be the outcome on the first toss. $f_Y(y) = \begin{cases} p & y = 1 \\ 1 - p & y = 0 \end{cases}$
- $E(X) = E(X|Y = 0) + E(X|Y = 1) = (1 + E(X))(1 - p) + 1(p) = 1 \cdot p + (1 - p)(1 + \mu) = \frac{1}{p}$
-

$$\text{Var}(X) = E(X^2) - (E(X))^2$$

$$\begin{aligned} E(X^2) &= E(E(X^2|Y)) \\ &= E(X^2Y = 1)p + E(X^2|Y = 0)(1 - p) \\ &= 1 \cdot p + E((1 + X)^2)(1 - p) \\ &= p + E(1 + 2X + X^2)(1 - p) \\ &= p + (1 - p) + (1 - p)\frac{2}{p} + E(X^2)(1 - p) \\ pE(X^2) &= 1 + \frac{2(1 - p)}{p} \\ E(X^2) &= \frac{1}{p} + \frac{2(1 - p)}{p^2} \end{aligned}$$

* So

$$\text{Var}(X) = \frac{1}{p} + \frac{2(1 - p)}{p^2} - \left(\frac{1}{p}\right)^2 = \frac{1 - p}{p^2}$$

Conditional Variance

Moment Generating Functions