

Lecture 22

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October 31, 2022

Calculating expectation and variance by conditioning

Defn:

$$\mathbb{E}_X X = \mathbb{E}_Y(\mathbb{E}_{X|Y}(X | Y))$$

Also,

$$\mathbb{E}_X g(X) = \mathbb{E}_Y(\mathbb{E}_{X|Y}(g(X) | Y))$$

$$\stackrel{g(X)=X^2}{\implies} \mathbb{E}_X X^2 = \mathbb{E}_Y(\mathbb{E}_{X|Y}(X^2 | Y))$$

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

in stats, called ANOVA.

Defn: Compound model: $S = X_1 + X_2 + \dots + X_N$, N = the number of events $\in \{0, 1, 2, \dots\}$, X_i are iid and independent of N .

$$\begin{aligned} \mathbb{E}S &= \mathbb{E}(\mathbb{E}(S | N)) \\ &= \mathbb{E}(\mathbb{E}(X_1 + \dots + X_N | N)) \\ &= \mathbb{E}(N \cdot \mathbb{E}X) \\ &= \mathbb{E}N \cdot \mathbb{E}X \end{aligned}$$

$$\begin{aligned} \text{Var } S &= \mathbb{E}(\text{Var}(S | N)) + \text{Var}(\mathbb{E}(S | N)) \\ &= \mathbb{E}(\text{Var}(X_1 + \dots + X_N | N)) + \text{Var}(\mathbb{E}(S | N)) \\ &= \mathbb{E}(N \cdot \text{Var}(X)) + \text{Var}(N \cdot \mathbb{E}(X)) \\ &= \mathbb{E}N \cdot \text{Var } X + (\mathbb{E}X)^2 \cdot \text{Var } N \end{aligned}$$

$$\text{Suppose } N = \mathbb{E}N \implies \text{Var } S = \mathbb{E}N \cdot \text{Var } X$$

$$\text{Suppose } X = \mathbb{E}X \implies S = \mathbb{E}X \cdot N \implies \text{Var } S = (\mathbb{E}X)^2 \text{Var } N$$

E.g.: Special case: $N \sim \text{Poisson}(\lambda)$; then S follows the compound Poisson model

$$\mathbb{E}N = \lambda = \text{Var } N$$

$$\implies \mathbb{E}S = \mathbb{E}N \cdot \mathbb{E}X = \lambda \mathbb{E}X$$

$$\begin{aligned} \text{Var } S &= \mathbb{E}N \cdot \text{Var } X + (\mathbb{E}X)^2 \text{Var } N \\ &= \lambda \text{Var } X + \lambda (\mathbb{E}X)^2 \\ &= \lambda \mathbb{E}(X^2) \end{aligned}$$

If λ is large, then we often use the normal distribution to approximate the compound Poisson.

$$\mathbb{P}(S \leq s) = \mathbb{P}\left(\frac{S - \mathbb{E}S}{\sqrt{\text{Var } S}} \leq \frac{s - \mathbb{E}S}{\sqrt{\text{Var } S}}\right)$$

$$\begin{aligned} \lambda \uparrow \implies \mathbb{P}(S \leq s) &\approx \mathbb{P}\left(Z \leq \frac{s - \mathbb{E}S}{\sqrt{\text{Var } S}}\right) \\ &= \Phi\left(\frac{s - \lambda \mathbb{E}X}{\sqrt{\lambda \mathbb{E}(X^2)}}\right) \end{aligned}$$

E.g.: Exs (not the compound model)

Last time Ex.3.11 mean of the geometric distribution, we computed it by conditioning on $Y = 1$ if first experiment was a success; 0, otherwise. $\implies \mathbb{E}N = \frac{1}{p}$

Ex3.19: Variance of the geometric:

Independent trials, each with probability p of success, N is the number of trials until 1st success. Calculate $\text{Var } N$.

Solution:

$$\begin{aligned}\text{Var } N &= \mathbb{E}(N^2) - (\mathbb{E}N)^2 \\ &= \mathbb{E}(N^2) - \frac{1}{p^2}\end{aligned}$$

$$\begin{aligned}\mathbb{E}(N^2) &= \mathbb{E}(\mathbb{E}(N^2 \mid Y)) \\ &= \mathbb{E}(N^2 \mid Y = 1)\mathbb{P}(Y = 1) + \mathbb{E}(N^2 \mid Y = 0)\mathbb{P}(Y = 0) \\ &= 1^2 \cdot p + \mathbb{E}((1 + N)^2)(1 - p) \\ &= p + \mathbb{E}(1 + 2N + N^2)(1 - p) \\ &= p + (1 + 2 \cdot \frac{1}{p} + \mathbb{E}(N^2))(1 - p)\end{aligned}$$

$$\begin{aligned}p\mathbb{E}(N^2) &= p + (1 - p) \left(1 + \frac{2}{p}\right) \\ \mathbb{E}(N^2) &= 1 - \left(1 - \frac{1}{p}\right) \left(1 + \frac{2}{p}\right) \\ &= 1 - \left(1 + \frac{1}{p} - \frac{2}{p^2}\right) \\ &= -\frac{1}{p} + \frac{2}{p^2}\end{aligned}$$

$$\text{Var } N = -\frac{1}{p} + \frac{2}{p^2} - \frac{1}{p^2} = \frac{1}{p^2} - \frac{1}{p} = \frac{1-p}{p^2}$$