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4. LLMS estimation with random sums

Problem 4. LLMS estimation with random sums

4/4 points (graded)

Let N be a random variable with mean $\mathbf{E}[N]=m$, and $\mathrm{Var}(N)=v$; let A_1,A_2,\ldots be a sequence of i.i.d random variables, all independent of N, with mean 1 and variance 1; let B_1,B_2,\ldots be another sequence of i.i.d. random variables, all independent of N and of A_1,A_2,\ldots , also with mean 1 and variance 1. Let $A=\sum_{i=1}^N A_i$ and $B=\sum_{i=1}^N B_i$.

1. Find the following expectations using the law of iterated expectations. Express each answer in terms of m and v, using standard notation.

$$\mathbf{E}[AB] =$$

$$v+m^2$$

$$\mathbf{E}[NA] =$$

$$v+m^2$$

$$v+m^2$$

$$v+m^2$$

$$v+m^2$$

$$v+m^2$$
Answer: m^2+v

2. Let $\hat{N}=c_1A+c_2$ be the LLMS estimator of N given A. Find c_1 and c_2 in terms of m and v.

$$c_1 = \boxed{\begin{array}{c} v/(m+v) \\ \hline \frac{v}{m+v} \end{array}}$$
 Answer: $v/(m+v)$

$$c_2 = \boxed{\begin{array}{c} (m^2)/(m+v) \\ \hline \frac{m^2}{m+v} \end{array}}$$
 Answer: $m^2/(m+v)$

STANDARD NOTATION

Solution:

1. We begin by finding $\mathbf{E}[AB]$.

$$\begin{split} \mathbf{E}[AB] &= \mathbf{E}[(A_1 + \dots + A_N)(B_1 + \dots + B_N)] \\ &= \mathbf{E}[\mathbf{E}[(A_1 + \dots + A_N)(B_1 + \dots + B_N)midN]] \\ &= \mathbf{E}[\mathbf{E}[(A_1 + \dots + A_N)midN]\mathbf{E}[(B_1 + \dots + B_N)midN]] \\ &= \mathbf{E}[N\mathbf{E}[A_1]N\mathbf{E}[B_1]] \\ &= \mathbf{E}[N^2] \\ &= \mathbf{Var}(N) + (\mathbf{E}[N])^2 \\ &= m^2 + v. \end{split}$$

Similarly,

$$egin{aligned} \mathbf{E}[NA] &= \mathbf{E}[\mathbf{E}[N(A_1 + \cdots + A_N)midN]] \ &= \mathbf{E}[N\mathbf{E}[A_1 + \cdots + A_NmidN]] \ &= \mathbf{E}[N(N\mathbf{E}[A_1])] \ &= \mathbf{E}[N^2] \ &= m^2 + v. \end{aligned}$$

2. A is the sum of a random number, N, of independent and identically distributed random variables A_1, \ldots, A_N . Therefore,

$$\mathbf{E}[A] = \mathbf{E}[\mathbf{E}[AmiaN]] = \mathbf{E}[\mathbf{E}[A_1]N] = m,$$
 and $\mathbf{Var}(A) = \mathbf{Var}(A_i)\mathbf{E}[N] + (\mathbf{E}[A_i])^2\mathbf{Var}(N) = m + v.$ Law of total variance: $\mathsf{var}(X) = \mathsf{E}[\mathsf{var}(X|Y)] + \mathsf{var}(\mathsf{E}[X|Y)]$ 不适用!因为variance的来源还有A的数量 Similarly, $\mathbf{E}[B] = m$, and $\mathbf{Var}(B) = m + v$. Furthermore, $\mathsf{cov}(N,A) = \mathbf{E}[NA] - \mathbf{E}[N]\mathbf{E}[A] = (m^2 + v) - m^2 = v.$

Finally,

$$egin{align} \hat{N} &= \mathbf{E}[N] + rac{\mathrm{cov}(N,A)}{\mathsf{Var}(A)}(A - \mathbf{E}[A]) \ &= m + rac{v}{m+v}(A-m) \ &= rac{m^2}{m+v} + rac{v}{m+v}A. \end{split}$$