

Statistics 523, Problem Set 4 Solutions

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1. Exercise 7.5.4, PfS, page 146: Suppose that X_1, \dots, X_n are i.i.d. with continuous distribution function F . Show that with probability 1 all the observations are distinct. [Hint: use corollary 2 to Theorem 5.1.3.]

Solution: Now if $i \neq j$, it follows that

$$P(X_i = X_j) = EE(1_{[X_i=X_j]}|X_j) = E\{\Delta F(X_j)\} = E\{0\} = 0$$

since $\Delta F(x) \equiv F(x) - F(x-) = 0$ for all x by continuity of F . Then

$$P(\cup_{1 \leq i < j \leq n} [X_i = X_j]) \leq \sum_{1 \leq i < j \leq n} P(X_i = X_j) = \sum_{1 \leq i < j \leq n} 0 = 0.$$

2. Suppose that $\{b_i\}_{i=1}^N$ and $\{c_i\}_{i=1}^N$ are two sequences of real numbers, and write $c(i) \equiv c_i$. Suppose that $\underline{R} = (R_1, \dots, R_N)$ is distributed uniformly over Π_N , the collection of all permutations of $\{1, \dots, N\}$; i.e. $P(\underline{R} = \underline{r}) = 1/N!$ for all $\underline{r} \in \Pi_N$. Let $S \equiv S_N \equiv \sum_{j=1}^N b_j c(R_j)$. Show that $Var(S) = (N-1)^{-1} B_N^2 \cdot C_N^2$ where $B_N^2 = \sum_{j=1}^N (b_j - \bar{b}_N)^2$ and $C_N^2 = \sum_{j=1}^N (c_j - \bar{c}_N)^2$.

Solution: As I showed in class,

$$Var(c(R_i)) = N^{-1} \sum_{j=1}^N (c_j - \bar{c})^2$$

and

$$Cov(c(R_i), c(R_j)) = -\frac{1}{N(N-1)} \sum_{j=1}^N (c_j - \bar{c})^2.$$

Thus

$$\begin{aligned}
\text{Var}(S_N) &= \sum_{i=1}^N b_i^2 \text{Var}(c(R_i)) + \sum_{i \neq j} b_i b_j \text{Cov}(c(R_i), c(R_j)) \\
&= \frac{1}{N(N-1)} \sum_{j=1}^N (c_j - \bar{c})^2 \left\{ (N-1) \sum_{i=1}^N b_i^2 - \sum_{i \neq j} b_i b_j \right\} \\
&= \frac{1}{N(N-1)} \sum_{j=1}^N (c_j - \bar{c})^2 N \sum_{i=1}^N (b_i - \bar{b})^2 \\
&= \frac{1}{N-1} \sum_{j=1}^N (c_j - \bar{c})^2 \sum_{i=1}^N (b_i - \bar{b})^2 \\
&= \frac{1}{N-1} B_N^2 C_N^2,
\end{aligned}$$

where we used the identity

$$(N-1) \sum_{i=1}^N b_i^2 - \sum_{i \neq j} b_i b_j = N \sum_{i=1}^N b_i^2 - N^2 \bar{b}^2 = N \sum_{i=1}^N (b_i - \bar{b})^2.$$

3. Suppose that X_1, \dots, X_n are the numbers resulting from sampling without replacement from an urn consisting of balls with the numbers a_1, \dots, a_N on the N balls. Let $\bar{a}_N \equiv \bar{a} \equiv N^{-1} \sum_{i=1}^N a_i$ and $\sigma_a^2 \equiv N^{-1} \sum_{i=1}^N (a_i - \bar{a})^2$. Let $T_n \equiv X_1 + \dots + X_n$. Verify that for $j \neq k, j, k \in \{1, \dots, N\}$,

$$\text{Cov}[X_j, X_k] = \text{Cov}[X_1, X_2] = -\frac{\sigma_a^2}{N-1}$$

and that

$$\text{Var}(T_n/n) = \frac{\sigma_a^2}{n} \left(1 - \frac{n-1}{N-1} \right).$$

The factor $(1 - (n-1)/(N-1))$ is sometimes called the *finite-sampling correction factor*; note that the variance of the mean is *smaller* than the variance of the mean under sampling with replacement (namely $n^{-1}\sigma_a^2$).

Solution: As discussed in class $T_n \stackrel{d}{=} S$ as in Problem 2 with $c_i = a_i$ for $1 \leq i \leq N$ and $b_i = 1_{\{1, \dots, n\}}(i)$ for $1 \leq i \leq N$. Now $\bar{c}_N = \bar{a}_N$, $\bar{b}_N = n/N$. Furthermore $C_N^2 = \sum_{i=1}^N (a_i - \bar{a})^2$ while

$$\begin{aligned}
B_N^2 &= \sum_{i=1}^N (b_i - \bar{b})^2 = \sum_{i=1}^N b_i^2 - N\bar{b}^2 \\
&= n - N(n/N)^2 = \frac{n}{N}(N-n).
\end{aligned}$$

Thus it follows from the calculation in Problem 2 that

$$\begin{aligned} \text{Cov}(X_i, X_j) &= \text{Cov}(a(R_i), a(R_j)) = -\frac{1}{N(N-1)} \sum_{i=1}^N (a_i - \bar{a})^2 \\ &= -\frac{\sigma_a^2}{N-1} \end{aligned}$$

for $i \neq j$ since $\sigma_a^2 = N^{-1} \sum_{i=1}^N (a_i - \bar{a})^2$, and

$$\begin{aligned} \text{Var}(T_n/n) &= \frac{1}{n^2} \cdot \frac{1}{N-1} \frac{n}{N} (N-n) \sum_{i=1}^N (a_i - \bar{a})^2 \\ &= \frac{\sigma_a^2}{n} \cdot \frac{N-n}{N-1} = \frac{\sigma_a^2}{n} \cdot \left(1 - \frac{n-1}{N-1}\right). \end{aligned}$$

4. Suppose that Y_1, Y_2, \dots are i.i.d. with distribution function G and characteristic function $\varphi(t) = E \exp(itY_1)$. Let N_λ a random variable with Poisson(λ) distribution and assume that N_λ is independent of the $\{Y_i\}$'s. Let $S \equiv S_\lambda \equiv \sum_{j=1}^{N_\lambda} Y_j$. Find the characteristic function ϕ_S of S .

Solution: By conditioning on N_λ we find that

$$\begin{aligned} \phi_S(t) &= E \exp(itS) = E[E(e^{it \sum_{j=1}^{N_\lambda} Y_j} | N_\lambda)] \\ &= E(\varphi(t)^{N_\lambda}) = \sum_{k=0}^{\infty} \varphi(t)^k e^{-\lambda} \frac{\lambda^k}{k!} \\ &= e^{-\lambda} \sum_{k=0}^{\infty} \frac{(\lambda \varphi(t))^k}{k!} = e^{\lambda(\varphi(t)-1)}. \end{aligned}$$