

Math 281A HW1

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Problem 1. Let X_i be IID (μ, σ^2) . Let τ be the skewness and κ be the kurtosis. Show that

$$\tau(\bar{X}) = \tau/\sqrt{n}, \quad \kappa(\bar{X}) = \kappa/n$$

Proof. Without loss of generality, set $\mu = 0$. First calculate $E(\bar{X})^3$:

$$E(\bar{X})^3 = \frac{1}{n^3} E\left(\sum_i X_i\right)^3 = \frac{1}{n^2} \mu_3$$

The last equation is due to the fact that all the components, except those third moments, have forms of $X_i^2 X_j$ or $X_i X_j X_k$, where all of them have expectation 0. Note that \bar{X} has variance σ/n , and therefore

$$\tau(\bar{X}) = \frac{\mu_3/n^2}{(\sigma^2/n)^{3/2}} = \frac{\mu_3}{\sigma^3 \sqrt{n}} = \tau/\sqrt{n}.$$

Similarly, by the same argument,

$$E(\bar{X})^4 = \frac{1}{n^4} E\left(\sum_i X_i\right)^4 = \frac{1}{n^3} \mu_4 + \frac{6n(n-1)}{n^4} \sigma^4 = \left(\frac{\kappa+3}{n^3} + \frac{6n(n-1)}{n^4}\right) \sigma^4.$$

Therefore

$$\kappa(\bar{X}) = E(\bar{X})^4 / (\sigma^4/n^2) - 3 = \frac{\sigma^4(\kappa+3n)}{n^3} \frac{n^2}{\sigma^4} - 3 = \kappa/n.$$

□

Problem 2. Show that

$$\text{var}\left(\frac{1}{n-1} \sum_i (X_i - \bar{X})^2\right) = \frac{\mu_4}{n} - \frac{n-3}{n(n-1)} \sigma^4.$$

Proof. We can assume that all X_i have mean zero here. Use the formula of the variance of quadratic form:

$$\text{var}(X'AX) = (\mu_4 - 3\mu_2^2)a'a + 2\mu_2^2 \text{tr}(A^2) + 4\mu_2 \theta' A^2 \theta + 2\mu_3 \theta' A a$$

where $a = \text{tr}(A)$. Plug in $A = I - II'/n$ and $X'AX = \sum_i (X_i - \bar{X})^2$. Calculation will finish the proof.

□

Problem 3. Let X_i be uncorrelated random variables with common expectation θ and variances. Then among all estimator $\sum_i \alpha_i X_i$ of θ that $\sum_i \alpha_i = 1$, the mean \bar{X} has the least variance.

Proof. It is equivalent to $\min \sum_i \alpha_i^2$ with respect to $\sum_i \alpha_i = 1$. A Lagrangian multiplier will do the job. Also this can be shown by assuming there exists α_i and α_j that not equal, averaging these two coefficients will produce an estimator that has less variance. \square

Problem 4. Minimize the variance of $\sum_i \alpha_i X_i$ under the following situations: (i) the variance of X_i is σ^2/a_i (a_i known); (ii) X_i have common variance σ^2 but correlated with common correlation coefficient ρ .

Solution . (i) The target function becomes $\min \alpha^2/a$ with respect to $\sum_i \alpha_i = 1$. Use the Lagrangian multiplier we have

$$2\alpha_i/a_i = \lambda$$

and this gives

$$2\alpha_i = \lambda a_i.$$

Using $\sum_i \alpha_i = 1$, we get $\lambda = 2/\sum_i a_i$. Therefore $\alpha = a_i/\sum_i a_i$.

(ii) Similar to (3), one may show that $\alpha = 1/n$ by using symmetry or Lagrangian multiplier.

Problem 5. Let X, Y have common mean θ , variance σ^2 and τ^2 , and correlation ρ . Determine the condition that (i) $\text{var}(X) < \text{var}[(X + Y)/2]$; (ii) The value that minimizes $\text{var}[\alpha X + (1 - \alpha)Y]$ is negative.

Solution . (i) Direct calculation shows that $\text{var}[(X + Y)/2] = (\sigma^2 + \tau^2 + 2\rho\sigma\tau)/4 > \sigma^2$, which is equivalent to $\tau^2 + 2\rho\sigma\tau > 3\sigma^2$. This is interpreted as the contribution of variance from Y should be larger than the piece from X , with correlations counted.

(ii) By calculating $\text{var}[\alpha X + (1 - \alpha)Y] = \alpha^2\sigma^2 + (1 - \alpha)^2\tau^2 + 2\alpha(1 - \alpha)\rho\sigma\tau$, we may see that when

$$\alpha = \frac{\tau^2 - \rho\sigma\tau}{\sigma^2 + \tau^2 - 2\rho\sigma\tau},$$

the variance is minimized. It is clear that the condition is $\tau - \rho\sigma < 0$. This means including one piece of Y will decrease the total variance compared to the contribution from the correlation of X .