

This homework is due on Wednesday, October 21 at 11:59pm. Submit your solutions in a pdf document on Moodle. Include your R code (which must be commented and properly indented) in the pdf file. Please name the pdf file <your last name>-HW1.pdf. Please also submit one text file with your R code, which must be commented and properly indented.

1. Suppose we can see a realization of X_1, \dots, X_n iid $N(\mu, \sigma^2)$. A $100(1-\alpha)\%$ random confidence interval for σ^2 is

$$\left(\frac{(n-1)S^2}{\chi_{1-\alpha/2, n-1}^2}, \frac{(n-1)S^2}{\chi_{\alpha/2, n-1}^2} \right), \quad (1)$$

where $\chi_{u, \nu}^2$ is the $100*uth$ percentile of the Chi-squared distribution with ν degrees of freedom, $S^2 = (n-1)^{-1} \sum_{i=1}^n (X_i - \bar{X})^2$, and $\bar{X} = n^{-1} \sum_{i=1}^n X_i$.

- (a) Perform a simulation study with 10000 replications to make a 99% *score* approximate confidence interval for the coverage probability of the random confidence interval for σ^2 defined in (1) when X_1, \dots, X_n are iid $N(\mu, \sigma^2)$. Set $\mu = 68, \sigma = 3$ and perform the study for $n = 10$ and $\alpha = 0.05$. Comment on the agreement between $1 - \alpha$ and the corresponding 99% *score* approximate confidence interval for the coverage probability.
 - (b) Perform a simulation study with 10000 replications to make a 99% *score* approximate confidence interval for the coverage probability of the random confidence interval for σ^2 defined in (1) when X_1, \dots, X_n are iid $\text{Exp}(1)$. Perform the study for each $(n, \alpha) \in \{10, 50, 500\} \times \{0.01, 0.05\}$. Comment on the agreement between $1 - \alpha$ and the corresponding 99% *score* approximate confidence interval for the coverage probability.
 - (c) Perform a simulation study with 10000 replications to make a 99% *score* approximate confidence interval for the coverage probability of the random confidence interval for σ^2 defined in (1) when $(X_1, \dots, X_n)'$ have the n -variate Normal distribution with mean vector $\mu \mathbf{1}_n$ and covariance matrix $\Sigma \in \mathbb{S}_0^n$. The (i, j) th entry of Σ is $\Sigma_{ij} = \sigma^2 \rho^{|i-j|}$. Set $\mu = 68, \sigma = 3, \rho = 0.7$ and perform the study for each $(n, \alpha) \in \{10, 50, 500\} \times \{0.01, 0.05\}$. Comment on the agreement between $1 - \alpha$ and the corresponding 99% *score* approximate confidence interval for the coverage probability.
2. In the two independent samples t-test model, we have access to a realization of X_1, \dots, X_{n_1} iid $N(\mu_1, \sigma^2)$ and access to a realization of the independent sequence Y_1, \dots, Y_{n_2} iid $N(\mu_2, \sigma^2)$. Let $\bar{X} = n_1^{-1} \sum_{i=1}^{n_1} X_i$, $\bar{Y} = n_2^{-1} \sum_{i=1}^{n_2} Y_i$, $S_1^2 = (n_1 - 1)^{-1} \sum_{i=1}^{n_1} (X_i - \bar{X})^2$, and $S_2^2 = (n_2 - 1)^{-1} \sum_{i=1}^{n_2} (Y_i - \bar{Y})^2$. To test

$$H_0 : \mu_1 - \mu_2 = 0$$

$$H_a : \mu_1 - \mu_2 \neq 0$$

we use the random test statistic

$$T = \frac{\bar{X} - \bar{Y} - 0}{S_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}},$$

where

$$S_p^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2}.$$

When H_0 is true, $T \sim \text{Tdist}(n_1 + n_2 - 2)$. The random p -value is

$$2\text{pt}(-|T|, n_1 + n_2 - 2),$$

where $\text{pt}(\cdot, \nu)$ is the cdf of the t -distribution with ν degrees of freedom.

- (a) Perform a simulation study to estimate the distribution of the random p -value when $n_1 = n_2 = 25$, $\mu_1 = \mu_2 = 68$, and $\sigma = 3$. What is the actual distribution of the random p -value in this case? Explain.
- (b) Fix $n_1 = n_2 = 50$, $\mu_1 = 68$, $\sigma = 3$, and significance level $\alpha = 0.05$. Produce a plot of a simulated estimate of the power curve of this test formed by increasing μ_2 (from 68). Select the sequence of values for μ_2 so that the simulated estimate of the power increases from roughly 5% to roughly 99%.
- (c) Fix $\mu_1 = 68$, $\mu_2 = 68.5$, $\sigma = 3$, and significance level $\alpha = 0.05$. Use simulation to find a value of $n_1 = n_2$ such that the simulated estimate of the power is roughly 95%.
- (d) Suppose that we perform this test with X_1, \dots, X_{n_1} iid $N(\mu_1, \sigma_1^2)$ independent of Y_1, \dots, Y_{n_2} iid $N(\mu_2, \sigma_2^2)$. We will see what happens when $\sigma_1 \neq \sigma_2$. Perform a simulation study to estimate the distribution of the random p -value in the following cases:
 - i. $n_1 = n_2 = 100$, $\mu_1 = \mu_2 = 68$, $\sigma_1 = 3$ and $\sigma_2 = 6$
 - ii. $n_1 = 20$, $n_2 = 100$, $\mu_1 = \mu_2 = 68$, $\sigma_1 = 3$ and $\sigma_2 = 6$
 - iii. $n_1 = 100$, $n_2 = 20$, $\mu_1 = \mu_2 = 68$, $\sigma_1 = 3$ and $\sigma_2 = 6$.

Since H_0 is true in all cases, we hope that the random p -value is approximately $\text{Unif}(0, 1)$. Create a QQ-plot for each case that compares the data percentiles (of the realizations of the random p -value) to the corresponding percentiles of $\text{Unif}(0, 1)$. Comment on the agreement. Is the 0.01 data quantile approximately equal to 0.01? Is the 0.05 data quantile approximately equal to 0.05?

- (e) Suppose that $n_1 = n_2$, and $(X_1, Y_1), \dots, (X_{n_1}, Y_{n_1})$ are iid $N_2((\mu_1, \mu_2)', \Sigma)$, where $\Sigma \in \mathbb{S}_0^2$. The two-independent samples t -test requires that $\Sigma = \sigma^2 I$ so we will see what happens when $\Sigma \neq \sigma^2 I$. In particular, suppose that Σ has (i, j) th entry $\Sigma_{ij} = \sigma^2 \{1(i = j) + \rho 1(i \neq j)\}$. Perform a simulation study to estimate the distribution of the random p -value in the following cases:
 - i. $n_1 = 20$, $\mu_1 = \mu_2 = 68$, $\sigma = 3$ and $\rho = 0.01$.
 - ii. $n_1 = 20$, $\mu_1 = \mu_2 = 68$, $\sigma = 3$ and $\rho = 0.1$.

Since H_0 is true in both cases, we hope that the random p -value is approximately $\text{Unif}(0, 1)$. Create a QQ-plot for each case that compares the data percentiles (of the realizations of the random p -value) to the corresponding percentiles of $\text{Unif}(0, 1)$. Comment on the agreement. Is the 0.01 data quantile approximately equal to 0.01? Is the 0.05 data quantile approximately equal to 0.05?

3. Suppose that we can see a realization of X_1, \dots, X_n iid $\text{Exp}(\mu)$ where μ is the unknown distribution mean. The density for $\text{Exp}(\mu)$ evaluated at x is

$$f(x; \mu) = \frac{1}{\mu} e^{-x/\mu} 1(x > 0).$$

- (a) Show that $\bar{X} = n^{-1} \sum_{i=1}^n X_i$ is the maximum likelihood estimator of μ .
- (b) Consider the competing *shrinkage* estimator $\hat{a}\bar{X}$, where

$$\hat{a} = \arg \min_{a \in \mathbb{R}_+} E\{(a\bar{X} - \mu)^2\}$$

and $\mathbb{R}_+ = \{b \in \mathbb{R} : b > 0\}$. Derive a closed-form expression for \hat{a} . Is the estimator $\hat{a}\bar{X}$ unbiased for μ ?

- (c) Report closed-form expressions for $E\{(\bar{X} - \mu)^2\}$ and $E\{(\hat{a}\bar{X} - \mu)^2\}$. Which estimator has the smaller mean-squared error?
- (d) For each $(n, \mu) \in \{5, 10, 50\} \times \{0.5, 1, 10\}$, perform a simulation study that compares the performance of \bar{X} and $\hat{a}\bar{X}$ as estimators of μ : report simulated estimates of $E\{(\bar{X} - \mu)^2\}$ and $E\{(\hat{a}\bar{X} - \mu)^2\}$. Compare these simulated estimates to their formulas derived in part 3c.