

Stat511(Section001): Homework #8

Due on Mar. 25, 2022 at 5:00pm

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Problem 1

We can see the hypothesis testing problem as:

$$H_0 : \lambda = 1000, H_1 : \lambda \neq 1000$$

Where λ is the parameter of exponential distribution. And let X_i be random variable of the i -th bulb's life time. We know under H_0 , $X_i \sim \exp(1000) = \text{Gamma}(1, 1000)$. Thus under H_0 , we have:

$$\sum_{i=1}^4 X_i \sim \text{Gamma}(4, 1000)$$

Then we use P-value to get the test, Since $\alpha = 0.05$, we have:

$$p\left(\sum_{i=1}^4 X_i > c_1\right) = 0.05/2 = 0.025$$

$$p\left(\sum_{i=1}^4 X_i < c_2\right) = 0.05/2 = 0.025$$

Then we need to find c_1, c_2 . It is easy to see that c_1 is 97.5-th percentile of $\text{Gamma}(4, 1000)$ and c_2 is 2.5th percentile of the distribution. Here, we can adjust Gamma distribution to get a chi-square distribution. However, since we can use computer here, we can direct get the c_1 is 8767.273, c_2 is 1089.865 by python and scipy. And we can get:

$$\sum_{i=1}^4 X_i = 4 \times \bar{X} = 4 \times 900 = 3600, 1089.865 < 3600 < 8767.273$$

Thus we fail to reject H_0 , which means we do not have significant evidence against the GE's claim. And this is our code for this part:

```
from scipy.stats import gamma
print("c1:", gamma.ppf(0.025, 4, scale=1000))
print("c2:", gamma.ppf(0.975, 4, scale=1000))
```

Problem 2

Now we have:

$$H_0 : \lambda = 1000, H_1 : \lambda = 900$$

Under H_1 , we have:

$$E_{\theta_1}(\phi(x)) = p_{\lambda=900}\left(\sum_{i=1}^4 X_i > 8767.273\right) + p_{\lambda=900}\left(\sum_{i=1}^4 X_i < 1089.865\right)$$

$$\sum_{i=1}^4 X_i \sim \text{Gamma}(4, 900)$$

By python and scipy, we can get the power of our test is $0.01248 + 0.03472 = 0.04720$. And our code for this part is:

```
from scipy.stats import gamma
print("p1:", 1 - gamma.cdf(8767.273, 4, scale=900))
print("p2:", gamma.cdf(1089.865, 4, scale=900))
```

Problem 3

(a)

we have $X_i \sim \exp(\mu_x)$, $Y_i \sim \exp(\mu_y)$. Then, Under H_0 , we have:

$$L_{H_0}(\mu_x|X, Y) = \mu_x^{-(n+m)} \exp\left(-\frac{1}{\mu_x} \left(\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i\right)\right)$$

since under H_0 , Y_i, X_i has same distribution. And we can see Y_i as X_{n+1} . Then we can get the MLE of them is $\hat{\mu}_x = \frac{1}{n+m} (\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i)$. Then we have:

$$\max L_{H_0}(\mu_x|X, Y) = \left(\frac{n+m}{\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i}\right)^{n+m} \exp(-(n+m))$$

Similarly, we have:

$$\max L_{H_1}(\mu_x, \mu_y|X, Y) = \mu_x^{-n} \mu_y^{-m} \exp\left(-\frac{1}{\mu_x} \sum_{i=1}^n X_i\right) \exp\left(-\frac{1}{\mu_y} \sum_{i=1}^m Y_i\right) = \left(\frac{n}{\sum_{i=1}^n X_i}\right)^n \left(\frac{m}{\sum_{i=1}^m Y_i}\right)^m \exp(-(n+m))$$

Then we have LRT statistic:

$$T(X, Y) = \frac{\left(\frac{n+m}{\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i}\right)^{n+m}}{\left(\frac{n}{\sum_{i=1}^n X_i}\right)^n \left(\frac{m}{\sum_{i=1}^m Y_i}\right)^m} = \left(\frac{(n+m) \sum_{i=1}^n X_i}{n(\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i)}\right)^n \left(\frac{(n+m) \sum_{i=1}^m Y_i}{m(\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i)}\right)^m$$

(b)

when $T = \frac{\sum_{i=1}^n X_i}{\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i}$, $1 - T = \frac{\sum_{i=1}^m Y_i}{\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i}$. Then we have:

$$T(X, Y) = \left(\frac{n+m}{n}\right)^n T^n \left(\frac{n+m}{m}\right)^m (1-T)^m$$

(c)

Under H_0 , we have:

$$\begin{aligned} \sum_{i=1}^n X_i &\sim \text{Gamma}(n, \mu_x) \rightarrow \frac{1}{\mu_x} \sum_{i=1}^n X_i \sim \text{Gamma}(n, 1) \\ \frac{1}{\mu_x} \left(\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i\right) &\sim \text{Gamma}(n+m, 1) \end{aligned}$$

Thus, we have:

$$T(X, Y) = \frac{\frac{1}{\mu_x} \sum_{i=1}^n X_i}{\frac{1}{\mu_x} (\sum_{i=1}^n X_i + \sum_{i=1}^m Y_i)} \sim \text{Beta}(n, m)$$

Then we know $\frac{L_0}{L_1}$ need to be less than some $k \in (0, 1)$ to reject H_0 . Hence we have:

$$P(T(X, Y) < k) = \alpha$$

Thus we can get k is the α -th percentile of $\text{Beta}(n, m)$. We will reject H_0 when the LRT statistic is less than k .

Problem 4

(a)

We have:

$$\max L_{H_0} = \begin{cases} 0 & \theta_0 < X_{(n)} \\ \left(\frac{1}{X_{(n)}}\right)^n & X_{(n)} \leq \theta_0 \end{cases}$$

Where $X_{(n)}$ the highest order statistic. For H_1 , we have:

$$L_{H_1} = \max L_{H_0 \cup H_1} = \left(\frac{1}{X_{(n)}}\right)^n$$

Thus we can get:

$$T(X) = \frac{L_{H_0}}{L_{H_1}} = \begin{cases} 0 & \theta_0 < X_{(n)} \\ 1 & X_{(n)} \leq \theta_0 \end{cases}$$

We will only reject H_0 when $\theta_0 < X_{(n)}$ for any level α .

(b)

we have $X_{(n)} = 22$ and $\theta_0 = 25$. Thus we have $\theta_0 > X_{(n)}$. And from the result in first part, we say we cannot reject the null hypothesis under level $\alpha = 0.05$.