

# Problem Set 4

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December 27, 2019

## Problem 1.

1. We have the log-likelihood function:

2. From previous result, if  $\lambda_0 \in L$ , we do not reject  $H_0$ . Otherwise, there is evidence to reject  $H_0$ .

## Problem 2.

1. Take derivative of Gaussian expression respect to  $\mu_1, \sigma_1, \mu_2, \sigma_2$  and set it equals to 0, we get:  
 $\hat{\mu}_1 = \bar{X}_{n1}, \hat{\mu}_2 = \bar{X}_{n2}, \hat{\sigma}_1^2 = \frac{1}{n1} \sum_{i=1}^{n1} (X_i - \hat{\mu}_1)^2$  and  $\hat{\sigma}_2^2 = \frac{1}{n2} \sum_{i=1}^{n2} (Y_i - \hat{\mu}_2)^2$ .

2. We have:  $\frac{n1\hat{\sigma}_1^2}{\sigma_1^2} = \frac{n1 \frac{1}{n1} \sum_{i=1}^{n1} (X_i - \mu_1)^2}{\sigma_1^2} = \sum_{i=1}^{n1} (\frac{X_i - \mu_1}{\sigma_1})^2$ . Since  $\frac{X_i - \mu_1}{\sigma_1} \sim N(0, 1)$ , hence  $\frac{n1\hat{\sigma}_1^2}{\sigma_1^2} \sim \chi_{n1}^2$ .

By similar method, we get  $\frac{n2\hat{\sigma}_2^2}{\sigma_2^2} \sim \chi_{n2}^2$

3. From previous result,  $\frac{n1\hat{\sigma}_1^2}{\sigma_1^2} + \frac{n2\hat{\sigma}_2^2}{\sigma_2^2} \sim \chi_{n1}^2 + \chi_{n2}^2 \sim \chi_{n1+n2}^2$ .

4. From CTT,  $\bar{X}_{n1} \sim N(\mu_1, \sigma_1^2)$ ,  $\bar{X}_{n2} \sim N(\mu_2, \sigma_2^2)$ . Hence,  $\Delta = \hat{\mu}_1 - \hat{\mu}_2 = \bar{X}_{n1} - \bar{X}_{n2} \sim N(\mu_1 - \mu_2, \sigma_1^2 + \sigma_2^2)$ .

5. From the previous question, we get  $\Delta = \hat{\mu}_1 - \hat{\mu}_2 = \bar{X}_{n1} - \bar{X}_{n2} \sim N(\mu_1 - \mu_2, \sigma_1^2 + \sigma_2^2)$ .

Under  $H_0$ :  $\Delta \sim N(0, \sqrt{\sigma_1^2 + \sigma_2^2})$ . Hence,  $\frac{\Delta}{\sigma_1^2 + \sigma_2^2} \rightarrow \frac{\Delta}{\sqrt{\sigma_1^2 + \sigma_2^2}} \sim N(0, 1)$ .

Let's denote test statistic  $T = \left| \frac{\Delta}{\sqrt{\sigma_1^2 + \sigma_2^2}} \right|$ . We now  $\mathbb{P}[T > c] = \alpha$ , hence  $c = 1 - \phi^{-1}(\frac{\alpha}{2})$ .

Conclusion: If  $T > c$ , we reject  $H_0$ , otherwise, we fail to reject  $H_0$ .

6. In this case, from the previous question, we have  $T = \left| \frac{\Delta}{\sqrt{\sigma_1^2 + \sigma_2^2}} \right| = \frac{8.43 - 8.07}{\sqrt{0.22^2 + 0.17^2}} = 1.295$ . Moreover,  $c = \phi^{-1}(1 - \frac{\alpha}{2}) = \phi^{-1}(0.975) = 1.96$ . Because  $T < c$ , we fail to reject  $H_0$  which means we can conclude two machines are significantly identical.

p-value =  $\mathbb{P}[|N(0, 1)| > 1.295] = 1 - \mathbb{P}[|N(0, 1)| \leq 1.295] = 1 - (2\mathbb{P}[N(0, 1) \leq 1.295] - 1) = 2 - 2\phi(1.295) = 0.3$ .

**Problem 3.** Let denote the function  $\mathbb{R}^2 \rightarrow \mathbb{R} : g(\mu, \sigma^2) = \mu - \sqrt{\sigma^2}$ . We have:

$$\frac{\partial g}{\partial \mu} = 1$$

$$\frac{\partial}{\partial \sigma^2} = -\frac{1}{2\sigma}$$

Let denote  $\hat{\mu} = \bar{X}_n$  and  $\hat{\sigma}^2 = \frac{1}{n} \sum (X_i - \hat{\mu})^2$ . Hence, by this result:

$$\sqrt{n}(\hat{\mu} - \mu) \rightarrow N(0, \sigma^2).$$

$$\sqrt{n}(\hat{\sigma}^2 - \sigma^2) \rightarrow N(0, 2\sigma^4)$$

Hence  $(\hat{\mu}, \hat{\sigma}^2)$  is asymptotically normal with covariance matrix:

$$\Sigma = \begin{bmatrix} \sigma^2 & 0 \\ 0 & 2\sigma^4 \end{bmatrix}$$

Using Delta method we have:

$$\sqrt{n}(g(\hat{\mu}, \hat{\sigma}^2) - g(\mu, \sigma^2)) \longrightarrow N(0, \nabla_{\mu, \sigma^2}^T \Sigma \nabla_{\mu, \sigma^2})$$

Where  $\nabla_{\mu, \sigma^2} = \nabla g(\mu, \sigma^2) = [1 - \frac{1}{2\sigma}]^T$ . Hence  $\nabla_{\mu, \sigma^2}^T \Sigma \nabla_{\mu, \sigma^2} = [1 - \frac{1}{2\sigma}] \begin{bmatrix} \sigma^2 & 0 \\ 0 & 2\sigma^4 \end{bmatrix} \begin{bmatrix} 1 \\ -\frac{1}{2\sigma} \end{bmatrix} = \frac{3}{2}\sigma^2$ .

Hence  $\sqrt{n}(g(\hat{\mu}, \hat{\sigma}^2) - g(\mu, \sigma^2)) \longrightarrow N(0, \frac{3}{2}\sigma^2)$  which equivalent to  $\sqrt{n}\frac{\sqrt{\frac{2}{3}}}{\sigma}(g(\hat{\mu}, \hat{\sigma}^2) - g(\mu, \sigma^2)) \longrightarrow N(0, 1)$ .  
Computing second order Delta method, we have:

$$ng(\hat{\mu}, \hat{\sigma}^2)^T \frac{2}{3\sigma^2} g(\hat{\mu}, \hat{\sigma}^2) \longrightarrow \chi_1^2$$

Let denote  $c$  is  $(1 - \alpha)$ -quantile of Kai-square with 1 degree of freedom. Because this is one-sided test statistic, if  $T > c$  we reject null hypothesis, othewise, we fail to reject null hypothesis.

Consider null hypothesis  $H_0 : \mu = \sigma$

In our case, under null hypothesis,  $T = ng(\hat{\mu}, \hat{\sigma}^2)^T \frac{2}{3\sigma^2} g(\hat{\mu}, \hat{\sigma}^2) = 100 \times (2.41 - \sqrt{5.2}) \times \frac{2}{3 \times 5.2} \times (2.41 - \sqrt{5.2}) = 0.215$  and  $c = 3.841$ . Since  $T < c$ , we fail to reject  $H_0 : \mu = \sigma$ . Hence, we also fail to reject  $H_0 : \mu < \sigma$ .

If the sample size is  $n = 100$ , the sample average is 3.28 and the sample variance is 15.95, by similar method, we have  $T = 2.13$ . At level 0.05 we know  $c = 3.841$ , hence  $T < c$  we fail to reject null hypothesis

At level 0.1, we know  $c = 2.706$ . Since  $T < c$ , we still fail to reject null hypothesis.