

MSA 8190 Statistical Foundations

Fall 2016

Sample Problems on Hypothesis Testing

Problem 1

The yield of a chemical process is being studied. From previous experience yield is known to be normally distributed and $\sigma = 3$. The past five days of plant operation have resulted in the following percent yields: 91.6, 88.75, 90.8, 89.95, and 91.3. Find a 95% two-sided confidence interval on the true mean yield.

Solution:

$$\bar{x} = \frac{91.6 + 88.75 + 90.8 + 89.95 + 91.3}{5} = 90.48$$

$$\alpha = 1 - 0.95 = 0.05$$

$$z_{\alpha/2} = \text{qnorm}(0.05/2, \text{lower.tail} = \text{FALSE}) = 1.959964$$

$$\left[\bar{x} - z_{\alpha/2} \frac{\sigma}{\sqrt{n}}, \bar{x} + z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \right] = \left[90.48 - 1.959964 \frac{3}{\sqrt{5}}, 90.48 + 1.959964 \frac{3}{\sqrt{5}} \right]$$

$$\Rightarrow [87.85043, 93.10957]$$

Problem 2

The heat evolved in calories per gram of a cement mixture is approximately normally distributed. The mean is thought to be 100 and the standard deviation is 2. We wish to test $H_0 : \mu = 100$ versus $H_1 : \mu \neq 100$ with a sample of $n = 9$ specimens.

- (a) If acceptance region is defined as $98.5 \leq \bar{x} \leq 101.5$, find the type I error probability α .
- (b) Find β for the case where the true mean heat evolved is 103.
- (c) Find β for the case where the true mean heat evolved is 105. This value of β is smaller than the one found in part (b) above. Why?

Solution:

$$\begin{aligned} \text{(a) } z_1 &= \frac{L - \mu_0}{\sigma/\sqrt{n}} = \frac{98.5 - 100}{2/\sqrt{9}} = -2.25 \\ z_2 &= \frac{U - \mu_0}{\sigma/\sqrt{n}} = \frac{101.5 - 100}{2/\sqrt{9}} = 2.25 \end{aligned}$$

$$\alpha = 1 - \mathbb{P}(98.5 \leq \bar{x} \leq 101.5 | \mu = 100) = \mathbb{P}(Z \leq -2.25) + \mathbb{P}(Z > 101.5) \\ \Rightarrow \alpha = \text{pnorm}(-2.25) + \text{pnorm}(2.25, \text{lower.tail} = \text{FALSE}) = 0.02444895$$

$$(b) \quad z_1 = \frac{L - \mu_1}{\sigma/\sqrt{n}} = \frac{101.5 - 103}{2/\sqrt{9}} = -2.25$$

$$z_2 = \frac{U - \mu_1}{\sigma/\sqrt{n}} = \frac{98.5 - 103}{2/\sqrt{9}} = -6.75$$

$$\beta = \mathbb{P}(98.5 \leq \bar{x} \leq 101.5 | \mu = 103) = \mathbb{P}(-6.75 \leq Z \leq -2.25) = \mathbb{P}(Z \leq -2.25) - \mathbb{P}(Z \leq -6.75) = \text{pnorm}(-2.25) - \text{pnorm}(-6.75) = 0.01222447$$

$$(c) \quad z_1 = \frac{L - \mu_2}{\sigma/\sqrt{n}} = \frac{101.5 - 105}{2/\sqrt{9}} = -5.25$$

$$z_2 = \frac{U - \mu_2}{\sigma/\sqrt{n}} = \frac{98.5 - 105}{2/\sqrt{9}} = -9.75$$

$$\beta = \mathbb{P}(98.5 \leq \bar{x} \leq 101.5 | \mu = 105) = \mathbb{P}(-9.75 \leq Z \leq -5.25) = \mathbb{P}(Z \leq -5.25) - \mathbb{P}(Z \leq -9.75) = \text{pnorm}(-5.25) - \text{pnorm}(-9.75) = 7.604961e - 08$$

The value of β is getting smaller because the true mean is getting farther away from the sample mean, so it becomes easier to detect that true mean is different than $\mu_0 = 100$.

Problem 3

A bearing used in an automotive application is suppose to have a nominal inside diameter of 1.5 inches. A random sample of 25 bearings is selected and the average inside diameter of these bearings is 1.4975 inches. Bearing diameter is known to be normally distributed with standard deviation $\sigma = 0.01$ inch.

- (a) Test the hypotheses $H_0 : \mu = 1.5$ versus $H_1 : \mu \neq 1.5$ using $\alpha = 0.01$
- (b) Compute the power of the test if the true mean diameter is 1.495 inches.
- (c) What sample size would be required to detect a true mean diameter as low as 1.495 inches if we wanted the power of the test to be at least 0.9?
- (d) Explain how the question in part (a) could be answered by constructing a two-sided confidence interval on the mean diameter.

Solution:

$$(a) \quad z_0 = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} = \frac{1.4975 - 1.5}{0.01/\sqrt{25}} = -1.25$$

$$z_{\alpha/2} = \text{qnorm}(0.01/2, \text{lower.tail} = \text{FALSE}) = 2.575829$$

Since the $|z_0| < z_{\alpha/2}$ we can't reject H_0 . In other words, we accept $\mu = 1.5$

$$(b) \quad d = \frac{\mu - \mu_0}{\sigma} = \frac{1.495 - 1.5}{0.01} = -0.5$$

$$\text{pwr.norm.test}(-0.5, n = 25, \text{sig.level} = 0.01)$$

$$\Rightarrow \text{Power} = 0.4697776$$

(c) $\text{pwr.norm.test}(-0.5, \text{sig.level} = 0.01, \text{power} = 0.9)$

$$\Rightarrow n = 59.51753 \approx 60$$

(d) 99% CI : $\left[1.4975 - 2.575829 \frac{0.01}{\sqrt{25}}, 1.4975 + 2.575829 \frac{0.01}{\sqrt{25}} \right]$

$$\Rightarrow [1.492348, 1.502652]$$

Since $\mu_0 = 1.5$ is inside the CI, we can't reject H_0 and we accept $\mu = 1.5$.

Problem 4

If $X_1, X_2, \dots, X_n \sim NID(\mu_1, \sigma^2)$, $Y_1, Y_2, \dots, Y_m \sim NID(\mu_2, \sigma^2)$, and X 's and Y 's are independent, find the distribution of the following statistics. (If the distribution is normal, specify the mean and variance. If it has chi-square, t, and/or F distribution, specify the degrees of freedom.)

(a) $\bar{X} - \bar{Y}$

(b) $\frac{\sum_{i=1}^n (X_i - \mu_1) - \sum_{i=1}^m (Y_i - \mu_2)}{\sigma}$

(c) $\frac{\sum_{i=1}^n (X_i - \mu_1)^2}{\sigma^2}$

(d) $\frac{\sum_{i=1}^n (X_i - \bar{X})^2}{\sigma^2}$

(e) $\frac{n(\bar{X} - \mu_1)^2}{\sigma^2}$

(f) $\frac{S_x^2}{S_y^2}$

(g) $\frac{\sum_{i=1}^n (X_i - \bar{X})^2 + \sum_{i=1}^m (Y_i - \bar{Y})^2}{\sigma^2}$

(h) $\frac{\sum_{i=1}^n (X_i - \mu_1)^2 + \sum_{i=1}^m (Y_i - \mu_2)^2}{\sigma^2}$

(i) $\frac{\bar{X} - \mu_1}{S_x / \sqrt{n}}$

Solution:

(a) $\bar{X} \sim N(\mu_1, \frac{\sigma^2}{n})$, $\bar{Y} \sim N(\mu_2, \frac{\sigma^2}{m}) \Rightarrow \bar{X} - \bar{Y} \sim N(\mu_1 - \mu_2, \frac{\sigma^2}{n} + \frac{\sigma^2}{m})$

$$\begin{aligned}
\text{(b)} \quad & \frac{\sum_{i=1}^n (X_i - \mu_1) - \sum_{i=1}^m (Y_i - \mu_2)}{\sigma} = \frac{\sum_{i=1}^n (X_i - \mu_1)}{\sigma} - \frac{\sum_{i=1}^m (Y_i - \mu_2)}{\sigma} \\
& \frac{\sum_{i=1}^n (X_i - \mu_1)}{\sigma} \sim N(0, n), \quad \frac{\sum_{i=1}^m (Y_i - \mu_2)}{\sigma} \sim N(0, m) \\
& \Rightarrow \frac{\sum_{i=1}^n (X_i - \mu_1) - \sum_{i=1}^m (Y_i - \mu_2)}{\sigma} \sim N(0, n + m)
\end{aligned}$$

$$\begin{aligned}
\text{(c)} \quad & \frac{(X_i - \mu_1)^2}{\sigma^2} \sim \chi_{(1)}^2, \quad i \in \{1, 2, \dots, n\} \\
& \Rightarrow \frac{\sum_{i=1}^n (X_i - \mu_1)^2}{\sigma^2} \sim \chi_{(n)}^2
\end{aligned}$$

$$\text{(d)} \quad \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{\sigma^2} \sim \chi_{(n-1)}^2$$

$$\begin{aligned}
\text{(e)} \quad & \frac{\bar{X} - \mu_1}{\sigma/\sqrt{n}} \sim N(0, 1) \\
& \Rightarrow \left(\frac{\bar{X} - \mu_1}{\sigma/\sqrt{n}} \right)^2 = \frac{n(\bar{X} - \mu_1)^2}{\sigma^2} \sim \chi_{(1)}^2
\end{aligned}$$

$$\begin{aligned}
\text{(f)} \quad & \frac{(n-1)S_x^2}{\sigma^2} \sim \chi_{(n-1)}^2, \quad \frac{(m-1)S_y^2}{\sigma^2} \sim \chi_{(m-1)}^2 \\
& \Rightarrow \frac{S_x^2}{S_y^2} \sim \mathcal{F}(n-1, m-1)
\end{aligned}$$

$$\begin{aligned}
\text{(g)} \quad & \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{\sigma^2} \sim \chi_{(n-1)}^2 \\
& \frac{\sum_{i=1}^m (Y_i - \bar{Y})^2}{\sigma^2} \sim \chi_{(m-1)}^2 \\
& \Rightarrow \frac{\sum_{i=1}^n (X_i - \bar{X})^2 + \sum_{i=1}^m (Y_i - \bar{Y})^2}{\sigma^2} \sim \chi_{(n+m-2)}^2
\end{aligned}$$

$$\begin{aligned}
\text{(h)} \quad & \frac{\sum_{i=1}^n (X_i - \mu_1)^2}{\sigma^2} \sim \chi_{(n)}^2 \\
& \frac{\sum_{i=1}^m (Y_i - \mu_2)^2}{\sigma^2} \sim \chi_{(m)}^2 \\
& \Rightarrow \frac{\sum_{i=1}^n (X_i - \mu_1)^2 + \sum_{i=1}^m (Y_i - \mu_2)^2}{\sigma^2} \sim \chi_{(n+m)}^2
\end{aligned}$$

$$(i) \frac{\bar{X}-\mu_1}{\sigma/\sqrt{n}} \sim N(0,1)$$

$$\frac{(n-1)S_x^2}{\sigma^2} \sim \chi_{(n-1)}^2$$

$$\Rightarrow \frac{(\bar{X}-\mu_1)\sqrt{n}/\sigma}{\sqrt{S_x^2/\sigma^2}} = \frac{\bar{X}-\mu_1}{S_x/\sqrt{n}} \sim t_{(n-1)}$$