

## **38. Normal Mean Tests (variance unknown)**

Simple Hypothesis Test / Example

Two-Sample Pooled- $t$  Hypothesis Test / Example

Two-Sample Approx- $t$  Hypothesis Test / Example

Two-Sample Paired- $t$  Hypothesis Test / Example

## Simple Hypothesis Test

Suppose  $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \text{Nor}(\mu, \sigma^2)$ , where  $\sigma^2$  is *unknown*.

Two-sided test:

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu \neq \mu_0$$

We'll use  $\bar{X}$  to estimate  $\mu$ . If  $\bar{X}$  is “significantly different” than  $\mu_0$ , then we'll reject  $H_0$ . To determine what “significantly different” means, we'll also need to estimate  $\sigma^2$ .

Define

$$T_0 \equiv \frac{\bar{X} - \mu_0}{S/\sqrt{n}},$$

where  $S^2$  is our old friend, the sample variance,

$$S^2 \equiv \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 \sim \frac{\sigma^2 \chi^2(n-1)}{n-1}.$$

### 9.38 Normal Mean Tests (var unknown)

If  $H_0$  is true, then

$$T_0 = \frac{\frac{\bar{X} - \mu_0}{\sqrt{\sigma^2/n}}}{\sqrt{\frac{S^2}{\sigma^2}}} \sim \frac{\text{Nor}(0, 1)}{\sqrt{\frac{\chi^2(n-1)}{n-1}}} \sim t(n-1).$$

So

$$\text{Reject } H_0 \quad \text{iff} \quad |T_0| > t_{\alpha/2, n-1}.$$

One-Sided Tests:

$$H_0 : \mu \leq \mu_0 \quad \text{vs.} \quad H_1 : \mu > \mu_0$$

$$\text{Reject } H_0 \quad \text{iff} \quad T_0 > t_{\alpha, n-1}.$$

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$$H_0 : \mu \geq \mu_0 \quad \text{vs.} \quad H_1 : \mu < \mu_0$$

$$\text{Reject } H_0 \quad \text{iff} \quad T_0 < -t_{\alpha, n-1}.$$

**Example:** Suppose we want to test at level 0.05 whether or not the mean of some process is 150.

We have  $n = 15$ ,  $\bar{X} = 152.18$ , and  $S^2 = 16.63$ .

Then

$$T_0 \equiv \frac{\bar{X} - \mu_0}{S/\sqrt{n}} = 2.07.$$

Since  $t_{\alpha/2, n-1} = t_{.025, 14} = 2.145$ , we *fail to reject*  $H_0$ .

## Two-Sample Normal Means Test when Variances are Unknown

Suppose we have the following set-up:

$$X_1, X_2, \dots, X_{n_x} \stackrel{\text{iid}}{\sim} \text{Nor}(\mu_x, \sigma_x^2) \quad \text{and}$$

$$Y_1, Y_2, \dots, Y_{n_y} \stackrel{\text{iid}}{\sim} \text{Nor}(\mu_y, \sigma_y^2),$$

where the samples are indep of each other, and  $\sigma_x^2$  and  $\sigma_y^2$  are *unknown*.

Which population has the larger mean?

## 9.38 Normal Mean Tests (var unknown)

Three Cases:

Pooled  $t$ -test:  $\sigma_x^2 = \sigma_y^2 = \sigma^2$ , say.

Approximate  $t$ -test:  $\sigma_x^2 \neq \sigma_y^2$ .

Paired  $t$ -test:  $(X_i, Y_i)$  observations paired.

## Pooled $t$ -Test

Suppose that  $\sigma_x^2 = \sigma_y^2 = \sigma^2$  (unknown).

We'll look at the two-sided test to see if the means are different.

$$H_0 : \mu_x = \mu_y$$

$$H_1 : \mu_x \neq \mu_y$$

### 9.38 Normal Mean Tests (var unknown)

Sample means and variances from the two popns:

$$\bar{X} \equiv \frac{1}{n_x} \sum_{i=1}^{n_x} X_i \quad \text{and} \quad \bar{Y} \equiv \frac{1}{n_y} \sum_{i=1}^{n_y} Y_i$$

$$S_x^2 = \frac{\sum_{i=1}^{n_x} (X_i - \bar{X})^2}{n_x - 1} \quad \text{and} \quad S_y^2 = \frac{\sum_{i=1}^{n_y} (Y_i - \bar{Y})^2}{n_y - 1}.$$

Define the *pooled* variance estimator by

$$S_p^2 = \frac{(n_x - 1)S_x^2 + (n_y - 1)S_y^2}{n_x + n_y - 2}.$$

### 9.38 Normal Mean Tests (var unknown)

If  $H_0$  is true, it can be shown that

$$S_p^2 \sim \frac{\sigma^2 \chi^2(n_x + n_y - 2)}{n_x + n_y - 2},$$

and then the test statistic

$$T_0 \equiv \frac{\bar{X} - \bar{Y}}{S_p \sqrt{\frac{1}{n_x} + \frac{1}{n_y}}} \sim t(n_x + n_y - 2).$$

Thus,

$$\text{Reject } H_0 \quad \text{iff} \quad |T_0| > t_{\alpha/2, n_x + n_y - 2}.$$

One-Sided Tests:

$$H_0 : \mu_x \leq \mu_y \quad \text{vs.} \quad H_1 : \mu_x > \mu_y$$

$$\text{Reject } H_0 \quad \text{iff} \quad T_0 > t_{\alpha, n_x + n_y - 2}.$$

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$$H_0 : \mu_x \geq \mu_y \quad \text{vs.} \quad H_1 : \mu_x < \mu_y$$

$$\text{Reject } H_0 \quad \text{iff} \quad T_0 < -t_{\alpha, n_x + n_y - 2}.$$

**Example:** Catalyst X is currently used by a certain chemical process. If catalyst Y gives higher mean yield, we'll use it instead.

Thus, we want to test  $H_0 : \mu_x \geq \mu_y$  vs.  $H_1 : \mu_x < \mu_y$ .

Suppose we have the following data:

$$n_x = 8, \quad \bar{X} = 91.73, \quad S_x^2 = 3.89$$

$$n_y = 8, \quad \bar{Y} = 93.75, \quad S_y^2 = 4.02$$

### 9.38 Normal Mean Tests (var unknown)

$S_x^2$  is pretty close to  $S_y^2$ , so we'll assume  $\sigma_x^2 \approx \sigma_y^2$ .

This justifies the use of the pooled variance estimator

$$S_p^2 = \frac{(n_x - 1)S_x^2 + (n_y - 1)S_y^2}{n_x + n_y - 2} = 3.955,$$

so that

$$T_0 = \frac{\bar{X} - \bar{Y}}{S_p \sqrt{\frac{1}{n_x} + \frac{1}{n_y}}} = -2.03.$$

### 9.38 Normal Mean Tests (var unknown)

Let's test at level  $\alpha = 0.05$ . Then

$$t_{\alpha, n_x + n_y - 2} = t_{.05, 14} = 1.761.$$

Since  $T_0 < -t_{\alpha, n_x + n_y - 2}$ , we *reject*  $H_0$ .

Thus, we should probably use catalyst  $Y$ .

## Approximate $t$ -Test

Suppose that  $\sigma_x^2 \neq \sigma_y^2$  (both unknown). As with our work with CI's, define

$$T_0^* \equiv \frac{\bar{X} - \bar{Y}}{\sqrt{\frac{S_x^2}{n_x} + \frac{S_y^2}{n_y}}} \approx t(\nu) \quad (\text{if } H_0 \text{ true}),$$

where the approximate degrees of freedom is given by

$$\nu \equiv \frac{\left( S_x^2/n_x + S_y^2/n_y \right)^2}{\frac{(S_x^2/n_x)^2}{n_x+1} + \frac{(S_y^2/n_y)^2}{n_y+1}} - 2.$$

9.38 Normal Mean Tests (var unknown)

2-sided	$H_0 : \mu_x = \mu_y$ $H_1 : \mu_x \neq \mu_y$	Reject $H_0$ iff $ T_0^*  > t_{\alpha/2, \nu}$
1-sided	$H_0 : \mu_x \leq \mu_y$ $H_1 : \mu_x > \mu_y$	Reject $H_0$ iff $T_0^* > t_{\alpha, \nu}$
1-sided	$H_0 : \mu_x \geq \mu_y$ $H_1 : \mu_x < \mu_y$	Reject $H_0$ iff $T_0^* < -t_{\alpha, \nu}$

**Example:** Let's test  $H_0 : \mu_x = \mu_y$  vs.  $H_1 : \mu_x \neq \mu_y$  at level  $\alpha = 0.10$ .

Suppose we have the following data:

$$n_x = 15, \bar{X} = 24.2, S_x^2 = 10$$

$$n_y = 10, \bar{Y} = 23.9, S_y^2 = 20$$

$S_x^2$  isn't very close to  $S_y^2$ , so we'll assume  $\sigma_x^2 \neq \sigma_y^2$ .

## 9.38 Normal Mean Tests (var unknown)

Plug-and-chug to get

$$T_0^* = 0.184,$$

$$\nu = 16.2 \rightarrow 16,$$

$$t_{\alpha/2, \nu} = t_{.05, 16} = 1.746.$$

Since  $|T_0^*| < t_{\alpha/2, \nu}$ , we *fail to reject*  $H_0$ .

(Actually,  $T_0^*$  was so small, we didn't need the tables.)

## Paired $t$ -Test

Again consider two competing normal pop'ns. Suppose we collect obs'ns from the two pop'ns in *pairs*.

The RV's within *different* pairs are *indep*. The two obs'ns within the *same* pair may *not* be indep — in fact, it's good for them to be positively correlated!

Example: One twin takes a new drug, the other takes a placebo.

## 9.38 Normal Mean Tests (var unknown)

In symbols,

$$\text{indep} \left\{ \begin{array}{l} \text{Pair 1 : } (X_1, Y_1) \\ \text{Pair 2 : } (X_2, Y_2) \\ \quad \quad \quad \vdots \\ \text{Pair } n : \underbrace{(X_n, Y_n)}_{\text{not indep}} \end{array} \right.$$

Define the pair-wise differences,

$$D_i \equiv X_i - Y_i, \quad i = 1, 2, \dots, n.$$

Note that  $D_1, D_2, \dots, D_n \stackrel{\text{iid}}{\sim} \text{Nor}(\mu_d, \sigma_d^2)$ , where

$$\mu_d \equiv \mu_x - \mu_y \quad \text{and}$$

$$\sigma_d^2 \equiv \sigma_x^2 + \sigma_y^2 - 2\text{Cov}(X_i, Y_i).$$

Define the sample mean and var of the differences,

$$\bar{D} \equiv \sum_{i=1}^n D_i/n \sim \text{Nor}(\mu_d, \sigma_d^2/n)$$

$$S_d^2 \equiv \frac{1}{n-1} \sum_{i=1}^n (D_i - \bar{D})^2 \sim \frac{\sigma_d^2 \chi^2(n-1)}{n-1}.$$

### 9.38 Normal Mean Tests (var unknown)

Then the test statistic is (assuming  $\mu_d = 0$ )

$$T_0 \equiv \frac{\bar{D}}{\sqrt{S_d^2/n}} \sim t(n-1).$$

Using the exact same manipulations as in the single-sample normal mean problem with unknown variance, we get the following...

## 9.38 Normal Mean Tests (var unknown)

2-sided	$H_0 : \mu_d = 0$ $H_1 : \mu_d \neq 0$	Reject $H_0$ iff $ T_0  > t_{\alpha/2, n-1}$
1-sided	$H_0 : \mu_d \leq 0$ $H_1 : \mu_d > 0$	Reject $H_0$ iff $T_0 > t_{\alpha, n-1}$
1-sided	$H_0 : \mu_d \geq 0$ $H_1 : \mu_d < 0$	Reject $H_0$ iff $T_0 < -t_{\alpha, n-1}$

Recall  $\mu_d = \mu_x - \mu_y$ , so that, e.g.,  $\mu_d = 0$  iff  $\mu_x = \mu_y$ .

9.38 Normal Mean Tests (var unknown)

**Example:** Times for people to parallel park two cars.

Person	Park Honda	Park Cadillac	Difference
1	10	20	-10
2	25	40	-15
3	5	5	0
4	20	35	-15
5	15	20	-5

Let's test  $H_0 : \mu_h = \mu_c$  at level  $\alpha = 0.10$ .

### 9.38 Normal Mean Tests (var unknown)

We see that  $n = 5$ ,  $\bar{D} = 9$ ,  $S_d^2 = 42.5$ .

This gives  $T_0 = 3.087$ .

Meanwhile,  $t_{.05,4} = 2.13$ , so we *reject*  $H_0$ .

Thus, we conclude that  $\mu_h \neq \mu_c$  (and it's probably the case that Hondas are easier to park).