

1. A store sells "16-ounce" boxes of *Captain Crisp* cereal. A random sample of 9 boxes was taken and weighed. The results were the following (in ounces):

15.5    16.2    16.1    15.8    15.6    16.0    15.8    15.9    16.2

Assume the weight of cereal in a box is normally distributed.

- a) Compute the sample mean  $\bar{x}$  and the sample standard deviation  $s$ .

$$\sum x = 143.1, \quad \sum x^2 = 2275.79, \quad \sum (x - \bar{x})^2 = 0.50.$$

$$\bar{x} = \frac{\sum x_i}{n} = \frac{143.1}{9} = \mathbf{15.9}.$$

$$s^2 = \frac{\sum x_i^2 - (\sum x_i)^2 / n}{n-1} = \frac{2275.79 - (143.1)^2 / 9}{8} = \frac{0.5}{8} = 0.0625.$$

OR

$$s^2 = \frac{\sum (x_i - \bar{x})^2}{n-1} = \frac{0.5}{8} = 0.0625.$$

$$s = \sqrt{s^2} = \sqrt{0.0625} = \mathbf{0.25}.$$

```
> x = c(15.5, 16.2, 16.1, 15.8, 15.6, 16.0, 15.8, 15.9, 16.2)
```

```
> mean(x)
[1] 15.9
```

OR

```
> sum(x)/length(x)
[1] 15.9
```

```
> sd(x)
[1] 0.25
```

OR

```
> sqrt(var(x))
[1] 0.25
```

OR

```
> sqrt(sum((x-mean(x))^2)/(length(x)-1))
[1] 0.25
```

- b) Construct a 95% confidence interval for the overall average weight of boxes of *Captain Crisp* cereal.

$\sigma$  is unknown.  $n = 9$  - small. The confidence interval :  $\bar{X} \pm t_{\alpha/2} \frac{s}{\sqrt{n}}$ .

$$\alpha = 0.05 \quad \alpha/2 = 0.025.$$

$$\text{number of degrees of freedom} = n - 1 = 9 - 1 = 8. \quad t_{\alpha/2} = 2.306.$$

$$15.9 \pm 2.306 \cdot \frac{0.25}{\sqrt{9}} \quad \mathbf{15.9 \pm 0.192} \quad \mathbf{(15.708 ; 16.092)}$$

```
> t.test(x, alternative=c("two.sided"), conf.level=0.95)
```

One Sample t-test

```
data: x
t = 190.8, df = 8, p-value = 6.372e-16
alternative hypothesis: true mean is not equal to 0
95 percent confidence interval:
 15.70783 16.09217
sample estimates:
mean of x
 15.9
```

```
OR
> qt(0.975, 8)
[1] 2.306004
> mean(x) - qt(0.975, 8) * sd(x) / sqrt(9)
[1] 15.70783
> mean(x) + qt(0.975, 8) * sd(x) / sqrt(9)
[1] 16.09217
```

- c) The company that makes *Captain Crisp* cereal claims that the average weight of its box is at least 16 ounces. Use a 0.05 level of significance to test the company's claim. What is the p-value of this test?

Claim:  $\mu \geq 16$

$H_0 : \mu \geq 16$  vs.  $H_1 : \mu < 16$ .

Left - tailed.

$\sigma$  is unknown.  $n = 9$  – small.

$$T = \frac{\bar{X} - \mu_0}{s / \sqrt{n}} = \frac{15.9 - 16}{0.25 / \sqrt{9}} = -1.2.$$

Rejection Region:  $T < -t_{\alpha}$

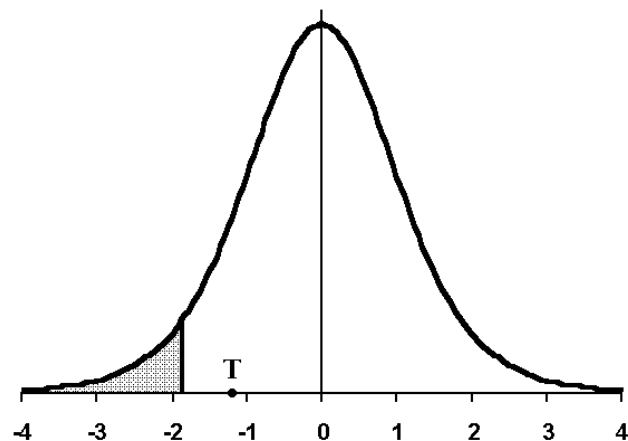
number of degrees of freedom

$$n - 1 = 9 - 1 = 8.$$

$$-t_{0.05} = -1.860.$$

The value of the test statistic is **not** in the Rejection Region.

**Do NOT Reject  $H_0$**   
at  $\alpha = 0.05$ .



OR

df	Upper-tail Probability			
	...	0.25	0.10	...
8	...	0.706	1.397	...

$$T = -1.2.$$

$$0.10 < \text{p-value} < 0.25.$$

$$\text{p-value} > \alpha.$$

**Do NOT Reject  $H_0$  at  $\alpha = 0.05$ .**

The t Distribution

df	$t_{0.40}$	$t_{0.25}$	$t_{0.10}$	$t_{0.05}$	$t_{0.025}$	$t_{0.01}$	$t_{0.005}$
8	0.262	0.706	1.397	1.860	2.306	2.896	3.355

```
> t.test(x,mu=16,alternative=c("less"),conf.level=0.95)
```

One Sample t-test

```
data: x
t = -1.2, df = 8, p-value = 0.1322
alternative hypothesis: true mean is less than 16
95 percent confidence interval:
 -Inf 16.05496
sample estimates:
mean of x
15.9
```

```
OR > pt(sqrt(9)*(mean(x)-16)/sd(x),8)
[1] 0.1322336
```

p-value >  $\alpha$ .

**Do NOT Reject  $H_0$  at  $\alpha = 0.05$ .**

2. Assume that the distributions of  $X$  and  $Y$  are  $N(\mu_1, \sigma^2)$  and  $N(\mu_2, \sigma^2)$ , respectively. Given the  $n = 6$  observations of  $X$ ,

70, 82, 78, 74, 94, 82

and the  $m = 8$  observations of  $Y$ ,

64, 72, 60, 76, 72, 80, 84, 68

find the p-value for the test  $H_0: \mu_1 = \mu_2$  vs.  $H_1: \mu_1 > \mu_2$ .

$$\bar{x} = 80, \quad s_x^2 = 68.8, \quad \bar{y} = 72, \quad s_y^2 = 64.$$

$$s_{\text{pooled}}^2 = \frac{(6-1) \cdot 68.8 + (8-1) \cdot 64}{6+8-2} = 66 \quad s_{\text{pooled}} \approx 8.124$$

$$\text{Test Statistic: } T = \frac{(\bar{X} - \bar{Y}) - \delta_0}{s_{\text{pooled}} \cdot \sqrt{\frac{1}{n} + \frac{1}{m}}} = \frac{(80 - 72) - 0}{8.124 \cdot \sqrt{\frac{1}{6} + \frac{1}{8}}} \approx \mathbf{1.82337}.$$

$$n + m - 2 = 6 + 8 - 2 = \mathbf{12} \text{ d.f.}$$

	Upper-tail Probability			
<i>df</i>	...	0.05	0.025	...
12	...	1.782	2.179	...

T = 1.82337.

**0.025 < p-value < 0.05.**

```
> x = c(70, 82, 78, 74, 94, 82)
> y = c(64, 72, 60, 76, 72, 80, 84, 68)
```

```
> t.test(x, y, alternative=c("greater"), var.equal=TRUE)
```

Two Sample t-test

```
data: x and y
t = 1.8234, df = 12, p-value = 0.04662
alternative hypothesis: true difference in means is greater than 0
95 percent confidence interval:
 0.1802451      Inf
sample estimates:
mean of x mean of y
      80      72
```

**OR**

```
> Spooled2 = ((6-1)*var(x)+(8-1)*var(y))/(6+8-2)
> Spooled2
[1] 66
> test_stat = (mean(x)-mean(y))/sqrt(Spooled2*(1/6+1/8))
> test_stat
[1] 1.823369
> 1-pt(test_stat, 6+8-2)
[1] 0.04661961
```

3. Consider the model:

$X_{11}, X_{12}, \dots, X_{1n}$  are i.i.d.  $N(\mu_1, \sigma^2)$

$X_{21}, X_{22}, \dots, X_{2n}$  are i.i.d.  $N(\mu_2, \sigma^2)$

Assume that  $\mu_1 = 6$ ,  $\mu_2 = 5$ ,  $\sigma^2 = 4$ ,  $n = 25$ .

Let  $\bar{X}_1 = \frac{1}{n} \sum_{i=1}^n X_{1i}$ ,  $\bar{X}_2 = \frac{1}{n} \sum_{i=1}^n X_{2i}$ ,  $D = \bar{X}_1 - \bar{X}_2$ .

a) Find  $P(0 < D < 2)$ .

$$D = \bar{X}_1 - \bar{X}_2 \sim N(\mu_1 - \mu_2, \frac{\sigma^2}{n} + \frac{\sigma^2}{n}) = N(6 - 5, \frac{4}{25} + \frac{4}{25})$$

$$D \sim N(1, 0.32) \quad \frac{D - 1}{\sqrt{0.32}} = Z \sim N(0, 1)$$

$$P(0 < D < 2) \approx P(-1.77 < Z < 1.77) = 0.9616 - 0.0384 = \mathbf{0.9232}.$$

OR

```
> z = 1/sqrt(0.32)
> z
[1] 1.767767
> pnorm(z) - pnorm(-z)
[1] 0.9229001
```

b) Empirical distribution of D:

Generate  $S = 1000$  datasets for each of group 1 and group 2.

For each of the  $s = 1 : 1000$  datasets, compute  $d_s = \bar{x}_{1s} - \bar{x}_{2s}$ .

Make a histogram for the 1000 values of  $d$ .

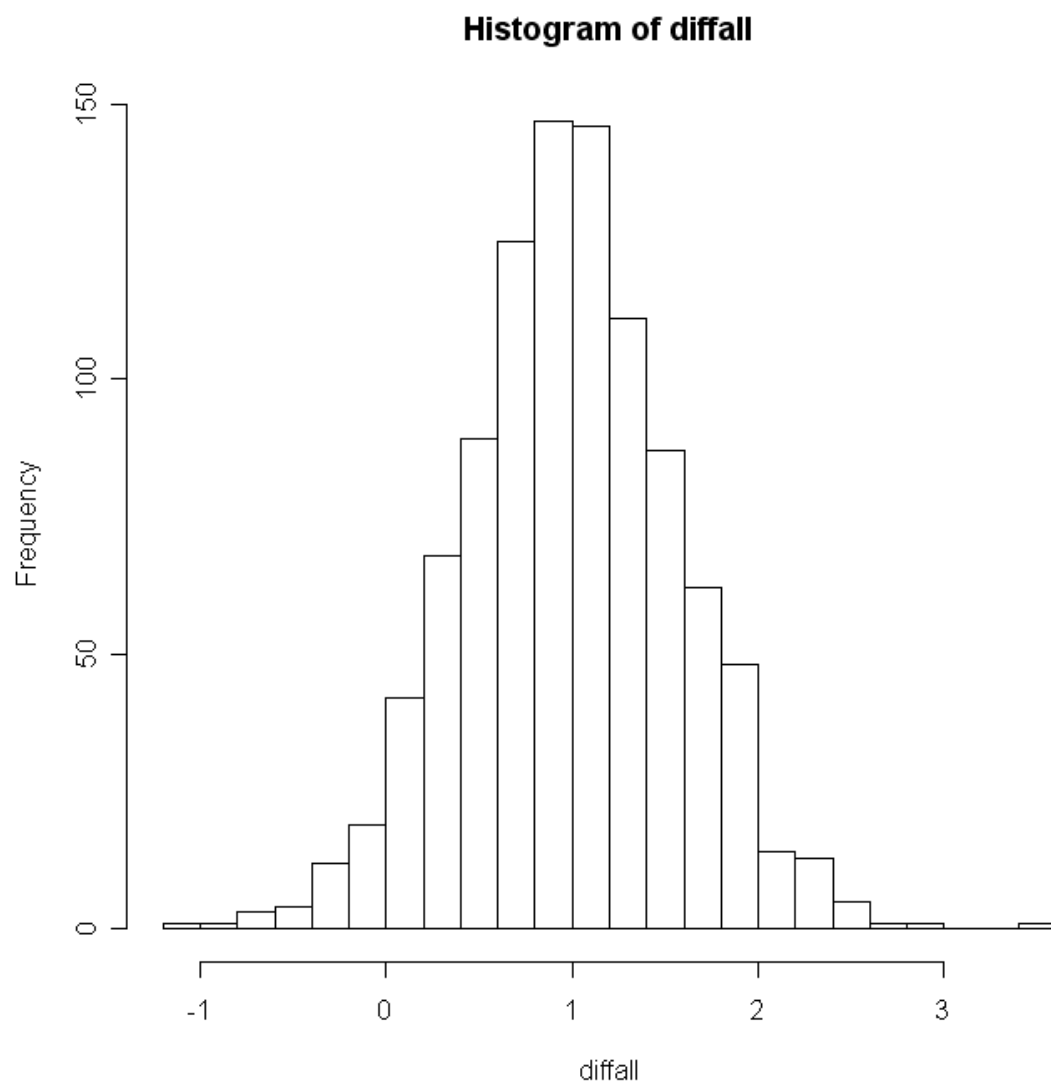
What is the proportion of values of  $d$  (among the 1000 values of  $d$  generated) that are between 0 and 2?

```
> N = 25
> mu1 = 6
> mu2 = 5
> std = 2
```

```

> S=1000
> count = 0
> diffall = c(1:S)
>
> for (i in 1:S){
+ x1 = rnorm(N, mu1, std)
+ x2 = rnorm(N, mu2, std)
+ diffall[i] = mean(x1) - mean(x2)
+ if ((diffall[i] > 0) & (diffall[i] < 2)){
+ count = count + 1}}
>
> count/S
[1] 0.925
>
> hist(diffall)

```



```
> qqnorm(difffall)  
> qqline(difffall)
```

