# Numerical variables from a single sample

Chapter 11

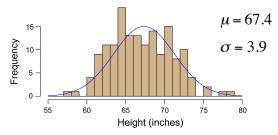
### Inference about means

Because  $\overline{Y}$  is normally distributed, we can convert its distribution to a standard normal distribution:

$$Z = \frac{\overline{Y} - \mu}{\sigma_{\overline{Y}}} = \frac{\overline{Y} - \mu}{\sigma / \sqrt{n}}$$

This would give a probability distribution of the difference between a sample mean and the population mean.

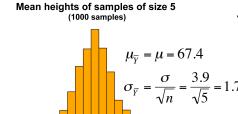
#### Heights of BIOL300 students (n = 157)



Frequency

55

60



70

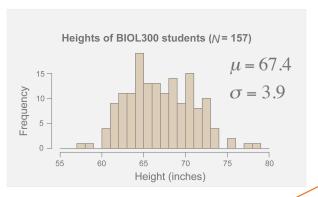
Mean height (inches)

 $\overline{Y}$  is normally distributed whenever: Y is normally distributed or n is large

## But... We don't know $\sigma$ ...

75

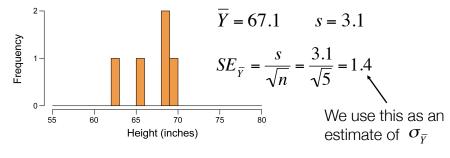
However, we do know s, the standard deviation of our sample. We can use that as an estimate of  $\sigma$ .



In most cases, we don't know the real population distribution.

We only have a sample.

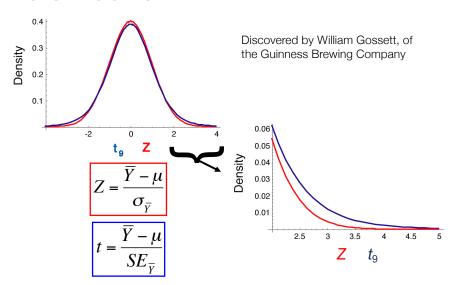
Heights of a sample of students (n = 5)



## A good approximation to the standard normal is then:

$$t = \frac{\overline{Y} - \mu}{SE_{\overline{Y}}} = \frac{\overline{Y} - \mu}{s/\sqrt{n}}$$

# t has a Student's t distribution



## Degrees of freedom

$$df = n - 1$$

## We use the *t*-distribution to calculate a confidence interval of the mean

$$-t_{\alpha(2),df} < \frac{\overline{Y} - \mu}{SE_{\overline{Y}}} < t_{\alpha(2),df}$$

We rearrange the above to generate:

$$\overline{Y} - t_{\alpha(2),df} SE_{\overline{Y}} < \mu < \overline{Y} + t_{\alpha(2),df} SE_{\overline{Y}}$$

Another way to express this is:  $\overline{Y} \pm SE_{\overline{Y}} t_{\alpha(2),df}$ 

https://www.youtube.com/watch?v=16aGSx9gFO4

# 95% confidence interval for a mean

Example:

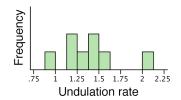
Paradise flying snakes



Undulation rates (in Hz)

0.9, 1.4, 1.2, 1.2, 1.3, 2.0, 1.4, 1.6

## Estimate the mean and standard deviation



$$\bar{Y} = 1.375$$

$$s = 0.324$$

$$n = 8$$

### Find the standard error

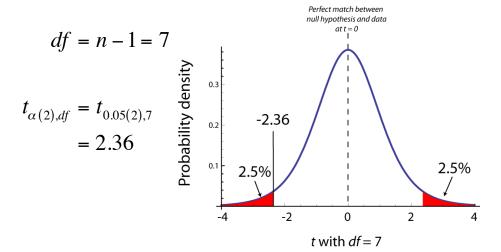
$$\overline{Y} \pm SE_{\overline{Y}} t_{\alpha(2),df}$$

$$SE_{\bar{Y}} = \frac{s}{\sqrt{n}} = \frac{0.324}{\sqrt{8}} = 0.115$$

#### Table C: Student's t distribution

	α(2):	0.2	0.10	0.05	0.02	0.01	0.001	0.0001
df	α(1):	0.1	0.05	0.025	0.01	0.005	0.0005	0.00005
1		3.08	6.31	12.71	31.82	63.66	636.62	6366.20
2		1.89	2.92	4.30	6.96	9.92	31.60	99.99
3		1.64	2.35	3.18	4.54	5.84	12.92	28.00
4		1.53	2.13	2.78	3.75	4.60	8.61	15.54
5		1.48	2.02	2.57	3.36	4.03	6.87	11.18
6		1.44	1.94	2.45	3.14	3.71	5.96	9.08
7		1.41	1.89	2.36	3.00	3.50	5.41	7.88
8		1.40	1.86	2.31	2.90	3.36	5.04	7.12
9		1.38	1.83	2.26	2.82	3.25	4.78	6.59

### Find the critical value of t



### Finding the critical value of t in R

## Putting it all together...

$$\overline{Y} \pm SE_{\overline{Y}} t_{\alpha(2),df} = 1.375 \pm 0.115 (2.36)$$
  
= 1.375 \pm 0.271

$$1.10 < \mu < 1.65$$

(95% confidence interval)

# Confidence interval for mean in R

#### t.test(undulationRate)\$conf.int

[1] 1.104098 1.645902
attr(,"conf.level")
[1] 0.95

[1] 0.9740838 1.7759162
attr(,"conf.level")
[1] 0.99

## 99% confidence interval

$$t_{\alpha(2),df} = t_{0.01(2),7} = 3.50$$

$$\overline{Y} \pm SE_{\overline{Y}} \quad t_{\alpha(2),df} = 1.375 \pm 0.115 \quad (3.50)$$
  
= 1.375 \pm 0.403

$$0.97 < \mu < 1.78$$

#### Confidence interval for the variance

$$\frac{df \quad s^2}{\chi^2_{\frac{\alpha}{2}, df}} \le \sigma^2 \le \frac{df \quad s^2}{\chi^2_{1 - \frac{\alpha}{2}, df}}$$

 $\begin{array}{c} 2.5\% \\ \\ \\ \\ \chi^2 \\ \\ \\ \\ \chi^2_{1-\alpha/2} \\ \end{array}$ 

$$\chi_{\frac{\alpha}{2},df}^{2} = \chi_{0.025,7}^{2} = 16.01$$

$$\chi_{1-\frac{\alpha}{2},df}^{2} = \chi_{0.975,7}^{2} = 1.69$$

#### Table A

df										
X	0.999	0.995	0.99	0.975	0.95	0.05	0.025	0.01	0.005	0.001
1	1.6	3.9E-5	0.00016	0.00098	0.00393	3.84	5.02	6.63	7.88	10.83
	E-6									
2	0	0.01	0.02	0.05	0.1	5.99	7.38	9.21	10.6	13.82
3	0.02	0.07	0.11	0.22	0.35	7.81	9.35	11.34	12.84	16.27
4	0.09	0.21	0.3	0.48	0.71	9.49	11.14	13.28	14.86	18.47
5	0.21	0.41	0.55	0.83	1.15	11.07	12.83	15.09	16.75	20.52
6	0.38	0.68	0.87	1.24	1.64	12.59	14.45	16.81	18.55	22.46
7	0.6	0.99	1.24	1.69	2.17	14.07	16.01	18.48	20.28	24.32
8	0.86	1.34	1.65	2.18	2.73	15.51	17.53	20.09	21.95	26.12

## 95% confidence interval for the variance of flying snake undulation rate

$$\frac{df \quad s^2}{\chi^2_{\frac{\alpha}{2}, df}} \le \sigma^2 \le \frac{df \quad s^2}{\chi^2_{1 - \frac{\alpha}{2}, df}}$$

$$df = n - 1 = 7$$

$$s^2 = (0.324)^2 = 0.105$$

[1] 16.01276

> qchisq(0.975,df=7,lower.tail=FALSE)

[1] 1.689869

## 95% confidence interval for the variance of flying snake undulation rate

$$\frac{df \ s^2}{\chi^2_{\frac{\alpha}{2}, df}} \le \sigma^2 \le \frac{df \ s^2}{\chi^2_{1 - \frac{\alpha}{2}, df}}$$

$$\frac{7 (0.324)^2}{16.01} \le \sigma^2 \le \frac{7 (0.324)^2}{1.69}$$

$$0.0459 \le \sigma^2 \le 0.435$$

## One-sample *t*-test

The one-sample *t*-test compares the mean of a random sample from a normal population with the population mean proposed in a null hypothesis.

## Test statistic for one-sample *t*-test

$$t = \frac{\overline{Y} - \mu_0}{s / \sqrt{n}}$$

 $\mu_0$  is the mean value proposed by  $H_0$ 

## Hypotheses for one-sample *t*-tests

 $H_0$ : The mean of the population is  $\mu_0$ .

 $H_A$ : The mean of the population is not  $\mu_0$ .

### Example: Human body temperature



H<sub>0</sub>: Mean healthy human body temperature is 98.6°F.

H<sub>A</sub>: Mean healthy human body temperature is not 98.6°F.

## Human body temperature

$$n = 24$$

$$\overline{Y} = 98.28$$

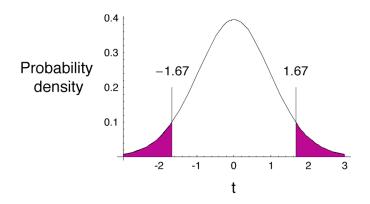
$$s = 0.940$$

$$t = \frac{\overline{Y} - \mu_0}{s / \sqrt{n}} = \frac{98.28 - 98.6}{0.940 / \sqrt{24}} = -1.67$$

## Degrees of freedom

$$df = n - 1 = 23$$

# Comparing *t* to its distribution to find the *P*-value

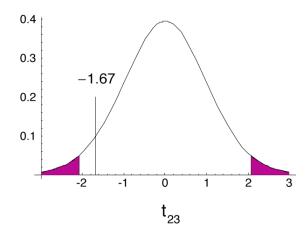


## A portion of the t table

df	α(1)	α(1)	α(1)	α(1)	α(1)
	=0.1	=0.05	=0.025	=0.01	=0.005
	$\alpha(2)=0.2$	$\alpha(2)=0.10$	$\alpha(2)=0.05$	$\alpha(2)=0.02$	$\alpha(2)=0.01$
20	1.33	1.72	2.09	2.53	2.85
21	1.32	1.72	2.08	2.52	2.83
22	1.32	1.72	2.07	2.51	2.82
23	1.32	1.71	2.07	2.5	2.81
24	1.32	1.71	2.06	2.49	2.8
25	1.32	1.71	2.06	2.49	2.79

-1.67 is closer to 0 than -2.07, so P > 0.05.

With these data, we cannot reject the null hypothesis that the mean human body temperature is 98.6.



```
t.test(bodyTempSmallData$temperature, mu =
98.6)

One Sample t-test

data: bodyTempSmallData$temperature
t = -0.56065, df = 24, p-value = 0.5802
alternative hypothesis: true mean is not equal
to 98.6
95 percent confidence interval:
98.24422 98.80378
sample estimates:
mean of x
98.524
```

## Body temperature revisited: n = 130

$$n = 130$$

$$\overline{Y} = 98.25$$

$$s = 0.733$$

$$t = \frac{\overline{Y} - \mu_0}{s / \sqrt{n}} = \frac{98.25 - 98.6}{0.733 / \sqrt{130}} = -5.44$$

## One-sample *t*-test: Assumptions

- 1. The variable is normally distributed.
- 2. The sample is a random sample.

## Body temperature revisited: n = 130

$$t = -5.44$$

$$t_{0.05(2),129} = \pm 1.98$$

t is further out in the tail than the critical value, so we could reject the null hypothesis. Human body temperature is not 98.6°F.  $P = 2.4 \times 10^{-7}$