# One-sample t-test in R

#### Cheatsheet

## 2024-08-10

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# i Assumed knowledge

- You know how to install and load packages in R.
- You know how to import data into R.
- You recognise data frames and vectors.

# Data structure

The data should be in a **long format** (also known as tidy data), where each row is an observation and each column is a variable (Figure 1). If your data is not already structured this way, reshape it manually in a spreadsheet program or in R using the pivot\_longer() function from the tidyr package.

Sex	BW
F M	2.15 2.55
F	2.95
F	2.70
M	2.20
F M	1.85
M M	2.55 2.60

Figure 1: Data should be in long format (left) where each row is an observation and each column is a variable. This is the preferred format for most statistical software. Wide format (right) is also common, but may require additional steps to analyse or visualise in some instances.

# Data

For this cheatsheet we will use the entire possums dataset used in BIOL2022 labs.

#### **About**

The one-sample t-test is used to determine whether the mean of a single sample y is significantly different from a known or hypothesised population mean  $(\mu)$ . **Examples**:

- Is the mean weight of canned tuna significantly different from what was stated on the label (400 g)?
- Is the mean height of a sample of male students significantly different from the national average height (175.6 cm)?
- Is the mean number of kittens in a litter significantly different from 4?

# Modelling

Is the mean **body weight** of possums (BW) significantly different from 3.5 kg?

The **simplified model** for the mathematically-adverse individual is

body weight  $\sim 3.5$ 

which translates to "the body weight of possums is around 3.5 kg". The **statistical model** is

body weight = 
$$\beta_0 + \epsilon$$

where  $\beta_0$  is the hypothesised population mean and  $\epsilon$  is the error term.

## Preparing the data

Extract **only** the variable of interest from the dataset using **select()** from the **dplyr** package – BW. Assign the variable to a new object – bw in this case.

```
library(dplyr)
library(readxl)
possums <- read_excel("possums.xlsx", sheet = 2) # import
bw <- select(possums, BW) # select variable</pre>
```

Your own data should be in a similar format.

## **Analytical approaches**

The traditional approach to the one-sample t-test is to use the t.test() function in R, while the modern approach is to use a general linear model (GLM) with the lm() or glm() functions.

#### t.test() function

#### Methods reporting

A one-sample t-test was used to determine whether the mean body weight of possums was significantly different from 3.5 kg. This was computed using the t.test() function in R version 4.4.0 (R Core Team, 2024).

## Perform the analysis

```
t.test(bw, mu = 3.5)
```

## Check assumption(s)

## **Normality**

Any combination of one or more of the following checks can be used to assess normality:

Histogram: hist(bw\$BW)Q-Q plot: qqnorm(bw\$BW)

• Shapiro-Wilk test: shapiro.test(bw\$BW)

Include the appropriate description in your methods section.

The normality of body weight was assessed using [insert method(s)].

#### How to report results

The mean body weight of possums was significantly different from 3.5 kg ( $t_{19}$  = -10.3, 95% CI [2.3, 2.7], p < 0.001).

#### lm() function

## Methods reporting

A general linear model was used to determine whether the mean body weight of possums was significantly different from 3.5 kg. This was computed using the lm() function in R version 4.4.0 (R Core Team, 2024).

#### Perform the analysis

For a one-sample t-test, the formula needs to be specified as  $y - \mu \sim 1$  where y is the variable of interest and  $\mu$  is the hypothesised value that is being tested. The 1 indicates that the model has an intercept only i.e. we are testing whether the mean difference is significantly different from 0.

```
fit <- lm((BW - 3.5) ~ 1, data = bw)
summary(fit)</pre>
```

## Check assumption(s)

## Normality

With a GLM, normality can be assessed using the residuals of the model. The following checks can be used:

- Histogram: hist(residuals(fit))
- Q-Q plot: qqnorm(residuals(fit))
- Shapiro-Wilk test: shapiro.test(residuals(fit))

## How to report results

There is evidence to suggest that the mean body weight of possums was significantly different from 3.5 kg (GLM,  $t_{19}=-10.3$ , p < 0.001).

# Exercise(s)

Download the penguins dataset (from below if you are reading this in HTML), or load the dataset from the palmerpenguins package. Perform a one-sample t-test to determine whether the mean flipper length of penguins is significantly different from 200 mm.