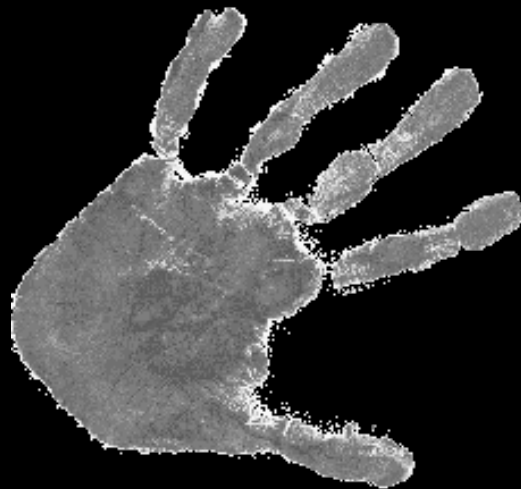


# hypothesis testing

Bisazza et al. (1996) tested the possibility of handedness in European toads, *Bufo bufo*, by sampling and measuring 18 toads from the wild.



Of the 18 toads tested, 14 were right-handed and 4 were left-handed. Are these results evidence of a predominance of one type of handedness in toads?



# stating the hypothesis

number of interest in population is the proportion that are right-handed

our null hypothesis should be that the two handedness types are equally frequent in the population

$$H_0: p=0.5$$

# stating the hypothesis

our alternative hypothesis should be that left- and right-handed toads are not equally frequent in the population

$H_A$ :  $p$  is not equal to 0.5

*this is a **two-sided hypothesis** because the alternative hypothesis includes values on both sides of the value specified by the null hypothesis*

# test statistic

quantity calculated from the data that is used to evaluate how compatible the data are with the result expected under the null hypothesis

# test statistic

on average, if the null hypothesis were correct, we would expect to observe nine right-handed toads (out of the 18)

instead, we observed 14 right-handed toads out of the 18 sampled



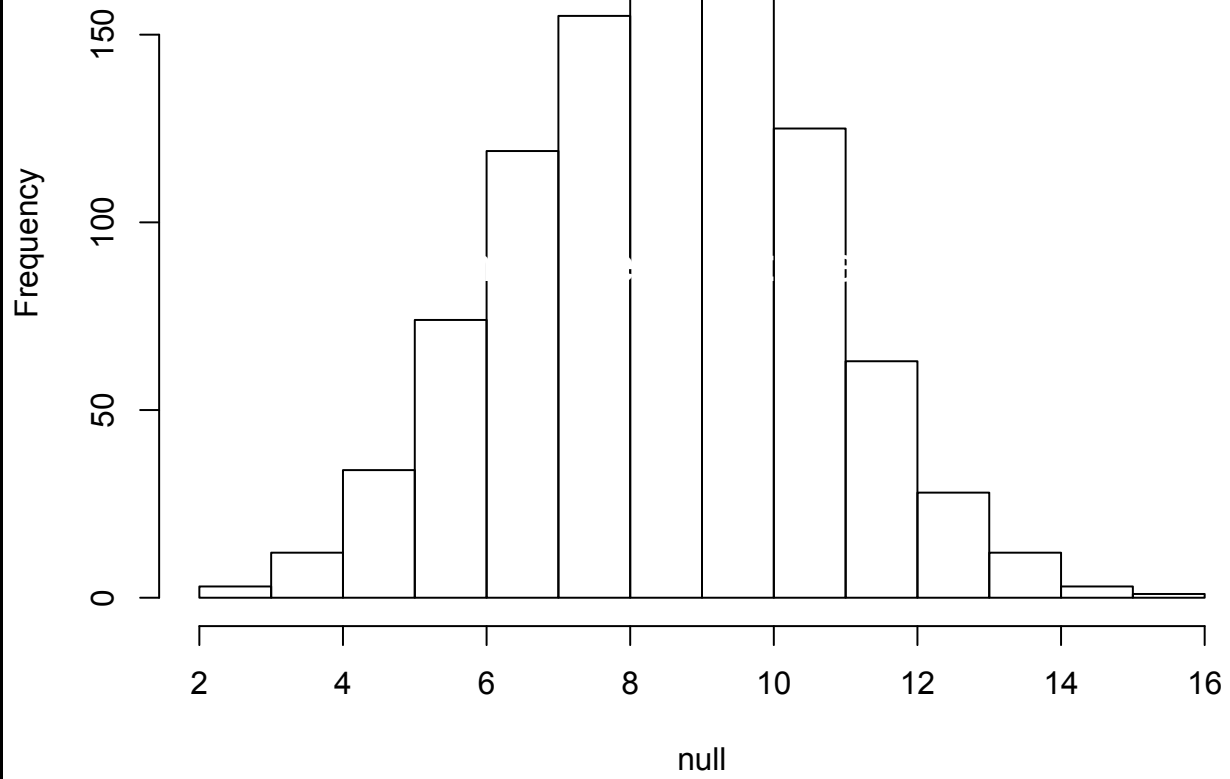
# null distribution

sampling distribution of outcomes for a test statistic under the assumption that the null hypothesis is true

sampling 18 toads under the null hypothesis is like tossing a coin in the air 18 times and counting the number of “heads” that come up (heads=right-handed)

# R Exercise: Generate Null Distribution

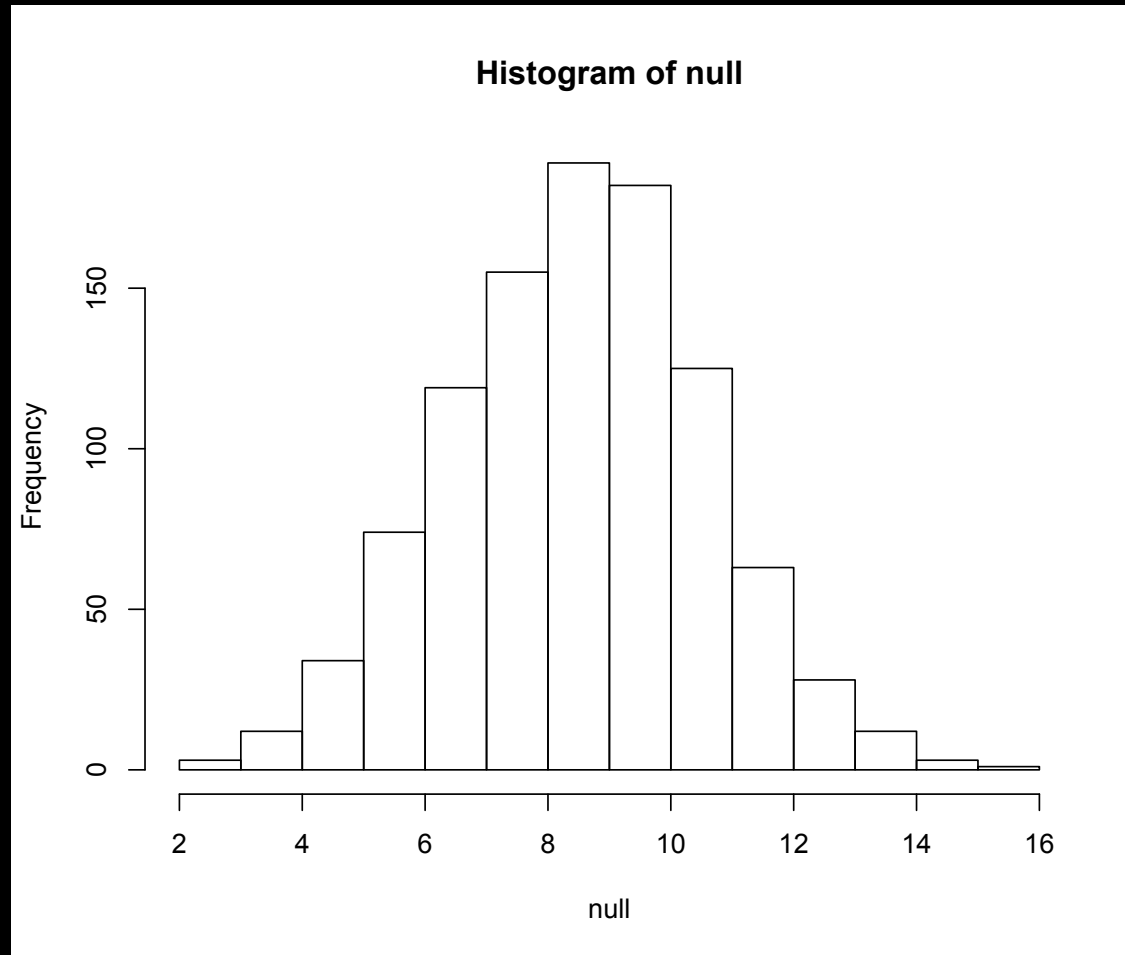
Histogram of null



# quantifying uncertainty: the $P$ -value

the  $P$ -value is the probability of obtaining the data (or data showing as great or greater difference from the null hypothesis) if the null hypothesis were true

# quantifying uncertainty: the *P*-value



# quantifying uncertainty: the $P$ -value

don't worry about the calculation of the  $P$ -value at the moment—we'll get to that next week

our  $P$ -value is around 0.031

# statistical significance

significance level ( $\alpha$ ): probability used as a criterion for rejecting the null hypothesis. If the  $P$ -value for a test is less than or equal to  $\alpha$ , then the null hypothesis is not rejected

a widely used significance level is  $\alpha=0.05$

# interpreting non-significant results

can never conclude that the null hypothesis is true

always possible

- true value differs from the null hypothesis by a small amount
- null was not rejected because of chance
- power of the test was limited by sample size

We interpret our results as the data are *compatible* or *consistent* with the null hypothesis.



# reporting the results

include the following information in the summary of the results of a statistical test

- value of the test statistic
- the sample size
- the  $P$ -value

It's also useful to provide confidence intervals, or at least the standard errors, for the parameters of interest

# errors in hypothesis testing

	Reality	
Decision	$H_0$ True	$H_0$ False
Reject $H_0$	Type I error	Correct
Do Not Reject $H_0$	Correct	Type II Error

**Type I error** is rejecting a true null hypothesis. The significance level  $\alpha$  sets the probability of committing a Type I error.

**Type II error** is failing to reject a false null hypothesis.

The **power** of a test is the probability that a random sample will lead to rejection of a false null hypothesis

# one-sided tests

alternative hypothesis includes parameter values on only one side of the value specified by the null hypothesis

$H_0$  is rejected only if the data depart from it in the direction stated by the  $H_A$