P-values

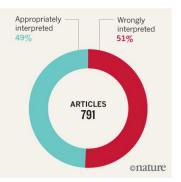
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WRONG INTERPRETATIONS

An analysis of 791 articles across 5 journals* found that around half mistakenly assume non-significance means no effect.

*Data taken from: P. Schatz et al. Arch. Clin. Neuropsychol. 20, 1053–1059 (2005); F. Fidler et al. Conserv. Biol. 20, 1539–1544 (2006); R. Hoekstra et al. Psychon. Bull. Rev. 13, 1033–1037 (2006); F. Bernardi et al. Eur. Sociol. Rev. 33, 1–15 (2017).



Source: V. Amrhein et al.

Recap

Null hypothesis: There is no interaction of Time and Site Fit ${\cal M}_0$:

Alternative hypothesis: There is an interaction of Time and Site Fit M_1 :

Lizards: LRT

```
(Lambda <- 2*(c(logLik(amodel))-c(logLik(nmodel))))
## [1] 1.098943
```

k <- attr(logLik(amodel), "df")-attr(logLik(nmodel), "df")</pre>

[1] 0.5772548

pchisq(Lambda, k,lower.tail=FALSE)

P-values: what are they?

We want to calculate some test-statistic, say Λ

The p-value is correctly interpreted as:

The probability to get this test-statistic or more extreme, assuming the null hypothesis is true

Unpacking:

- It is a probability
- Related to a test statistic
- (Usually) requires knowledge of its distribution

P-values: how to interpret them?

P < 0.05: This result is probably not due to chance P > 0.05: This result is likely to due chance

- P-values from a single study will change!
- P-values are computed assuming the null hypothesis is true
- P-values are a function of **effect size** and **sample size**

Given enough data, an effect near zero will be statistically significant

P-values: how not to interpret them

- Probability of the null hypothesis
- Probability of wrongly rejecting the null hypothesis
- P > 0.05 as lack of effect.
- Importance of the effect
- Anything else?

P-values are not a reflection of scientific importance

P-values as measure of evidence

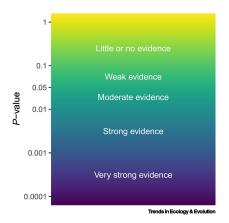
Trends in **Ecology & Evolution**



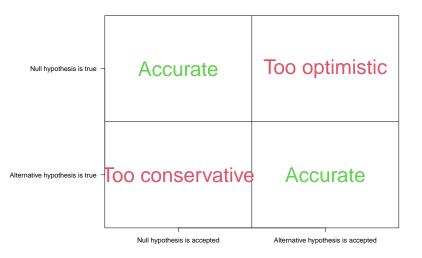
Opinion

Rewriting results sections in the language of evidence

Stefanie Muff @, 1,2,*,@ Erlend B. Nilsen, 2,3,4,@ Robert B. O'Hara, 1,2,@ and Chloé R. Nater^{2,3,®}



Errors



Type 1 error

A false positive: mistakenly rejecting the null hypothesis.

- ightharpoonup Controlled by α
- We choose $\alpha = 0.05$ usually

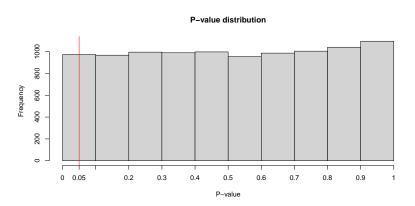
Type 2 error

A false negative: mistakenly accepting the null hypothesis.

- D A
- Not controlled explicitly
- Implicitly controlled by the sample size

Distribution of p-values

Under the null hypothesis:



When the alternative hypothesis is true, the distribution is clustered near zero.

Example: Lizards

Fit M_0 :

Alternative hypothesis: Diameter matters.

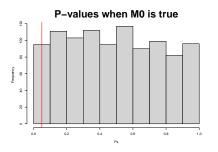
Fit M_1 :

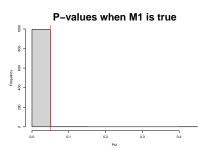
```
amodel <- update(nmodel, formula = .~Time+Site+Diameter)</pre>
```

P-values by simulation

```
Ps <- NULL
for(i in 1:1000){
ynew <- as.matrix(stats::simulate(nmodel))</pre>
nmodel2 <- glm(ynew~Time+Site,
             data = lizards, family="binomial")
amodel2 <- update(nmodel2, formula = .~Time+Site+Diameter)
# Store P-value
(Lambda <- 2*(logLik(amodel2)-logLik(nmodel2)))
k <- attr(logLik(amodel2), "df")-attr(logLik(nmodel2), "df")</pre>
Ps <- c(Ps, pchisq(Lambda, k,lower.tail=FALSE))
}
```

P-values by simulation





P-values larger than 0.05:

0.954 0.003

Conclusion

- 5% of the time we will wrongly accept the alternative hypothesis
- Thus your statistically significant result may be due to chance
- ?% of the time we will wrongly accept the null hypothesis
- "Highly significant" does not exist. Everything p < 0.05 is statistically significant
- Rejection of the alternative hypothesis does not mean "no effect"

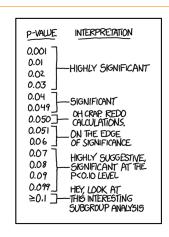


Figure 1: http://xkcd.com/