#### P-values

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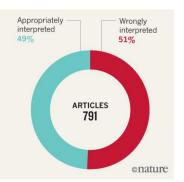
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#### Motivation

#### 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

## 'log Lik.' 0.5772548 (df=6)

```
(Lambda <- 2*(logLik(amodel)-logLik(nmodel)))

## 'log Lik.' 1.098943 (df=6)

k <- attr(logLik(amodel), "df")-attr(logLik(nmodel), "df")
pchisq(Lambda, k,lower.tail=FALSE)</pre>
```

# P-values: what are they?

We want to calculate some test-statistic, say  $\Lambda$ 

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
- ightharpoonup 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

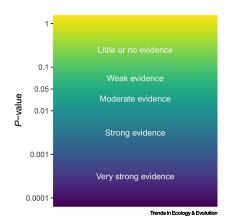
### Trends in Ecology & Evolution



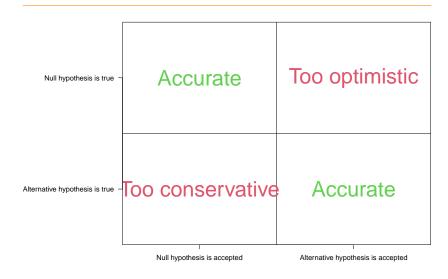
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Rewriting results sections in the language of evidence

Stefanie Muff © , <sup>1,2,4,8</sup> Erlend B. Nilsen, <sup>2,3,4,8</sup> Robert B. O'Hara, <sup>1,2,8</sup> and Chloé R. Nater <sup>2,3,8</sup>



#### **Errors**



# Type 1 error

A false positive: mistakenly rejecting the null hypothesis.

- $\blacktriangleright$  Controlled by  $\alpha$
- $\blacktriangleright$  We choose  $\alpha=0.05$  usually

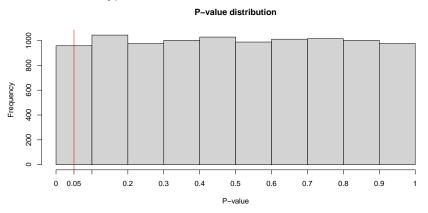
# Type 2 error

A false negative: mistakenly accepting the null hypothesis.

- ) E
- 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.

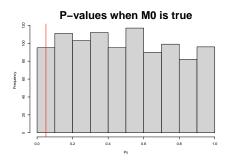
Fit  $M_1$ :

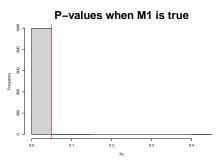
```
amodel <- update(nmodel, formula = .~Time+Site+Diameter)</pre>
```

### 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)</pre>
# Store P-value
(Lambda <- 2*(logLik(amodel2)-logLik(nmodel2)))
k <- attr(logLik(amodel2), "df")-attr(logLik(nmodel2), "df")
Ps <- c(Ps, pchisq(Lambda, k,lower.tail=FALSE))
}
```

## P-values by simulation





## P-values larger than 0.05: 0.95

## P-values larger than 0.05: 0.003

- 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"

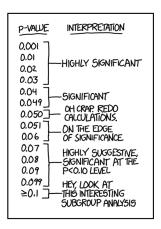


Figure 1: http://xkcd.com/