Statistical hypotheses

Bayesian and non-Bayesian

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Statistical hypothesis

A statistical hypothesis is a model for data.

For example,

$$Y \sim Ber(\theta)$$

or

$$Y \sim Bin(10, 0.25)$$

or

$$Y_i \stackrel{ind}{\sim} N(0, \sigma^2)$$

or

$$Y_i \stackrel{ind}{\sim} N(\mu, \sigma^2).$$

Translating a scientific hypothesis into a statistical hypothesis

Scientific hypothesis: the coin is fair

Statistical hypothesis:
Let Y be an indicator that the coin is flipped heads.

$$Y \sim Ber(0.5)$$

Scientific hypothesis: the coin is biased, but we don't know the probability

Statistical hypothesis:

$$Y \sim Ber(\theta).$$

Null hypothesis

Wikipedia definition:

> *the null hypothesis, H_0, is the [model] that there is no relationship between two measured phenomena or no association among groups*

My definition:

> *the null hypothesis is the straw man model that nobody believes is true*

For example, the coin is fair

$$H_0 : Y \sim Ber(0.5).$$

Alternative hypothesis

Wikipedia definition:

> *the alternative hypothesis, H_A, is [the model] that states something is happening, a new theory is preferred instead of an old one (null hypothesis).*

My definition:

> *the alternative hypothesis is the model that the researcher believes*

For example, the coin is biased, but we don't know the probability

$$H_A : Y \sim Ber(\theta)$$

Null vs alternative hypothesis

We typically simplify notation and write null and alternative hypotheses like this:

Model:

$$Y \sim Ber(\theta)$$

Hypotheses:

$$H_0 : \theta = 0.5 \qquad \text{versus} \qquad H_A : \theta \neq 0.5$$

I prefer

$$H_0 : Y \sim Ber(0.5) \qquad \text{versus} \qquad H_A : Y \sim Ber(\theta)$$

so that we remind ourselves that these hypotheses are models.

Bayesian hypotheses

Bayesian hypotheses are full probability models for the data.

For example,

$$Y \sim Ber(0.5)$$

or

$$Y|\theta \sim Ber(\theta) \qquad \text{and} \qquad \theta \sim Be(a,b)$$

for known values of a and b.

Prior predictive distribution

The prior predictive distribution is the distribution for the data with all the parameters integrated out, i.e.

$$p(y) = \int p(y|\theta)p(\theta)d\theta.$$

For example, if

$$Y|\theta \sim Ber(\theta) \qquad \text{and} \qquad \theta \sim Be(a,b)$$

then

$$\begin{aligned} p(y) &= \int p(y|\theta)p(\theta)d\theta \\ &= \int_0^1 y^\theta(1-y)^{1-\theta}\frac{1}{Beta(a,b)}\theta^{a-1}(1-\theta)^{b-1}d\theta \\ &= \frac{1}{Beta(a,b)}\int_0^1 \theta^{a+y-1}(1-\theta)^{b+n-y-1}d\theta \\ &= \frac{Beta(a+y,b+n-y)}{Beta(a,b)} \end{aligned}$$

which is the probability mass function for the beta-binomial distribution

Comments

Three points about Bayesian hypotheses:

- Must use proper priors.
- No special hypotheses.
- Not restricted to 2 hypotheses.

Summary

- Model:

$$Y \sim Ber(\theta)$$

- Null hypothesis:

$$H_0 : \theta = 0.5$$

- Alternative hypothesis:

$$H_A : \theta \neq 0.5$$