

STAT 417 Lectures Notes 18.

§ 7.2.3. Bayes Factors

A motivating example eg1: Suppose we have $n=99$ i.i.d measurements.

$x_1, x_2, \dots, x_{99}$ from $N(\mu, \frac{1}{100})$, where $\mu \in \mathbb{R}$ is unknown.

Want to test $H_0: \mu \leq 0$ vs $H_1: \mu > 0$.

In STAT 350. We talked about $H_0: \mu = \mu_0$. Here, the above H_0 is more challenging. Below is how we solve the problem.

Assume prior distribution on μ : $\mu \sim N(0, 1)$.

then the posterior of μ is

$$\mu | x_1, \dots, x_{99} \sim N(\hat{\mu}, \sigma^2), \quad \text{with } \hat{\mu} = \frac{99\bar{x} + (0)(1)}{(99)(1) + 1} = \frac{99\bar{x}}{100}$$

$$\sigma^2 = \frac{(1)(1)}{(99)(1) + 1} = \frac{1}{100}.$$

~~Suppose~~ Suppose in practice we observe $\bar{x} = 0$.

then $\hat{\mu} = 0$, so $\mu | x_1, \dots, x_{99} \sim N(0, 0.01)$

To test H_0 vs H_1 , we just evaluate the posterior probability of H_0 and H_1 : the bigger one is preferred.

Exactly, $p(H_0 | x_1, \dots, x_{99}) = p(\mu \leq 1 | x_1, \dots, x_{99})$

$$= p\left(\frac{\mu - 0}{0.1} \leq \frac{0.1 - 0}{0.1} \mid x_1, \dots, x_{99}\right)$$

$$= \Phi(1)$$

$$= 1 - \Phi(-1) = 1 - 0.1587 = 0.8413.$$

so, $p(H_1 | x_1, \dots, x_{99}) = 1 - 0.8413 = 0.1587.$

so $p(H_0 | x_1, \dots, x_{99}) > p(H_1 | x_1, \dots, x_{99}).$

so we do NOT reject H_0 .

The above procedure makes much sense since our decision is made based on the probability of H_0 and H_1 . The bigger one is of course preferred.

The ratio $\frac{p(\mu \leq 0.1 | x_1, \dots, x_{99})}{p(\mu > 0.1 | x_1, \dots, x_{99})}$ is called ~~Bayes Factor~~ the posterior odds ratio in favor of H_0 .

we always compare ~~Bayes Factor~~ posterior odds ratio with one to make decision.

More generally, if A is a set of $\mathbb{R}$. we want to test

$$H_0: \theta \in A$$

$$H_1: \theta \notin A.$$

Suppose we have got the posterior distribution of θ , which is $f(\theta|s)$,
~~where~~ where s is the sample. then the posterior odds ratio
in favor of H_0 is

$$\frac{p(H_0|s)}{p(H_1|s)} = \frac{p(\theta \in A|s)}{p(\theta \notin A|s)}$$

If the ratio is greater than one, then we prefer H_0 , otherwise, we prefer H_1 . Note that the above ratio is equal to

$$\frac{p(\theta \in A|s)}{1 - p(\theta \in A|s)}.$$

Ex 2. Suppose the posterior distribution of θ is $f(\theta|s)$ defined as

$$f(\theta|s) = \begin{cases} e^{-(\theta-s)}, & \text{if } \theta \geq s, \\ 0, & \text{if } \theta < s \end{cases}$$

Test $H_0: \theta > 2$. vs $H_1: \theta \leq 2$.

① if $s = 0.9$

② if $s = 1$.

Sol: ① $p(\theta > 2 | s) = \int_2^{\infty} e^{-(\theta-1.9)} d\theta = \int_{0.1}^{\infty} e^{-\theta} d\theta = e^{-0.1} = 0.9048.$

$$p(\theta \leq 2 | s) = 1 - 0.9048 = 0.0952$$

so the posterior odds ratio in favor of H_0 is

$$\frac{0.9048}{0.0952} = 9.51 > 1.$$

so we prefer H_0 , i.e., do not reject H_0 .

②. $p(\theta > 2 | s) = \int_2^{\infty} e^{-(\theta-1)} d\theta = \int_1^{\infty} e^{-\theta} d\theta = e^{-1} = 0.3679.$

$$so \quad p(\theta \leq 2 | s) = 1 - 0.3679 = 0.6321.$$

so the posterior odds ratio of H_0 in favor of H_0 is

$$\frac{0.3679}{0.6321} = 0.58 < 1.$$

so we reject H_0 .

EX1. For what value of the sample s do you reject H_0 ?
In Eg 2.

Sol: If $s > 2$. then $p(H_0 | s) = \int_2^{\infty} f(\theta | s) d\theta = \int_s^{\infty} e^{-(\theta-s)} d\theta = 1.$

so the posterior odds ratio in favor of H_0 is $\frac{1}{1-1} = \infty > 1.$

so we always do not reject H_0 .

If $0 \leq s \leq 2$, $p(H_0 | s) = \int_2^{\infty} e^{-(\theta-s)} d\theta = \int_{2-s}^{\infty} e^{-\theta} d\theta = e^{-(2-s)}.$

the posterior odds ratio is $\frac{e^{-(2-s)}}{1 - e^{-(2-s)}} < 1$ to reject H_0 .

so $s < \log\left(\frac{e^2}{e^2 - 1}\right) \approx 1.31$