

Hypothesis testing

25.10.2005

GE02: day 4 part 1

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Statistic

- An experiment leads to data (observed)
- There is some expectation (null hypothesis) about what the data could be
- “Statistic” is a quantity which measures how (un)likely the deviation of observed from the expected is, under the null hypothesis. The more is deviation, the higher is statistics
- Examples of a statistic:
Z, Wald, Score, χ^2 , Likelihood Ratio Test...

P-value

- P-value is probability that the (observed or even more extreme) deviation of observed from expected could have occurred purely by chance
- Given a test statistic used, it is computed as the probability that the test statistic has a value “**at least as extreme**” as the observed value

Example: association study

- Association study between a disease and an $A \rightarrow T$ polymorphism
- Null hypothesis is that T has equal frequency in cases and controls
- We observe that the frequency of T in cases is 0.47 and the frequency of T in controls is 0.40

Two-sided P-value

We do not know *a priori*, is T or A involved in the disease

- Alternative hypothesis: frequency of T is not equal in cases and controls (T may be more frequent in cases or more frequent in controls)
- Though we observe that T is 7% more frequent in cases, the P-value is the probability that frequency of T would differ by $\geq 7\%$ between cases and controls
- That is P-value is probability under null that T would be 7% more frequent in cases OR that T is 7% less frequent in cases
- This is called 2-sided P-values

One-sided P-value

We know that T disrupts the product of the gene or have other very strong evidence that T is a “bad” variant

- Alternative hypothesis: T is more frequent in cases
- P-value is probability that T would be $\geq 7\%$ more frequent by chance
- This is called one-sided P

Hypothesis testing

If P-value is less than or equal to some pre-specified threshold, then you say that

- “At this threshold, I do not believe that this deviation from expected is purely by chance”
 - “Therefore I reject the null hypothesis”
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- Usually the threshold is taken to be 0.05 (5%)

P-value: an example

- Experiment: four heads in 20 trials
- Null hypothesis is that the coin is “fair”
 - $P(\text{head})=P(\text{tail})=0.5$
- Alternative hypothesis is that it is **not** fair
 - *A priori* we do not know in what way (heads or tails) it is not fair
- We see that deviation from expected (10) is 6
- P-value is the chance that a fair coin in 20 trials would deviate from expected (10) by 6 or more

Solution: 4 out of 20, Binomial

$$\begin{aligned} \text{P-value} = & P(k=0) + P(k=1) + P(k=2) + P(k=3) + \\ & P(k=4) + P(k=16) + P(k=17) + P(k=18) + \\ & P(k=19) + P(k=20) = \end{aligned}$$

$$\begin{aligned} & \frac{1}{2}^{20} + 20 \frac{1}{2}^{20} + 190 \frac{1}{2}^{20} + 1140 \frac{1}{2}^{20} + \\ & 4845 \frac{1}{2}^{20} + 4845 \frac{1}{2}^{20} + 1140 \frac{1}{2}^{20} + 190 \frac{1}{2}^{20} + \\ & 20 \frac{1}{2}^{20} + \frac{1}{2}^{20} = 6196 \frac{1}{2}^{20} = \end{aligned}$$

0.012

...finally!..

Because it is symmetric:

$$\text{P-value} = 2 \times [P(k=0) + P(k=1) + P(k=2) + P(k=3) + P(k=4)]$$

P-value using Normal app-n (score test)

$$P(k \leq x) = \Phi\left(\frac{(x + 0.5) - \mu}{\sigma}\right) = \Phi\left(\frac{(x + 0.5) - np_0}{\sqrt{np_0(1 - p_0)}}\right)$$

$$\begin{aligned} 2 \cdot P(k \leq 4) &= 2 \cdot \Phi\left(\frac{4 - 10 + 0.5}{\sqrt{20 \cdot 0.5 \cdot 0.5}}\right) = \\ &= 2 \cdot \Phi(-2.46) = 2 \cdot (1 - \Phi(2.46)) = \\ &= 2 \cdot (1 - 0.993) = 0.014 \end{aligned}$$

Score (Z) test

With (Yate's) continuity correction

$$Z = \begin{cases} \frac{|O - E - 0.5|}{\sqrt{Var_0}}, & \text{if } O > E \\ \frac{|O - E + 0.5|}{\sqrt{Var_0}}, & \text{if } O < E \end{cases}$$

Without correction

$$Z = \frac{|O - E|}{\sqrt{Var_0}}$$

Where Var_0 is the variance expected under Null

With 2-tailed P, $Z > 1.96$ give $P < 0.05$

P-value using Normal (Wald test)

$$P(k \leq x) = \Phi\left(\frac{(x + 0.5) - \mu}{\sigma}\right) = \Phi\left(\frac{(x + 0.5) - np_0}{\sqrt{n\hat{p}(1 - \hat{p})}}\right)$$

$$\begin{aligned} 2 \cdot P(k \leq 4) &= 2 \cdot \Phi\left(\frac{4 - 10 + 0.5}{\sqrt{20 \cdot 0.2 \cdot 0.8}}\right) = \\ &= 2 \cdot \Phi(-3.07) = 2 \cdot (1 - \Phi(3.07)) = \\ &= 2 \cdot (1 - 0.999) = 0.002 \end{aligned}$$

Wald test

With (Yate's) continuity correction

$$Z = \begin{cases} \frac{|O - E - 0.5|}{\sqrt{Var_1}}, & \text{if } O > E \\ \frac{|O - E + 0.5|}{\sqrt{Var_1}}, & \text{if } O < E \end{cases}$$

Without correction

$$Z = \frac{|O - E|}{\sqrt{Var_1}}$$

Where Var_0 is the variance expected under Null

Chi-squared distribution

- Chi-squared with m d.f. is sum of m squared standard Normal
- Consider several classes $i = 1, 2, \dots, m$
- in each class expected (E_i) and observed (O_i) is known
- Chi-squared statistic (with and w/o cont-ty corr.)

$$\sum_{i=1}^m \frac{(O_i - E_i \pm 0.5)^2}{E_i} \infty \chi_{m-1}^2$$

$$\sum_{i=1}^m \frac{(O_i - E_i)^2}{E_i} \infty \chi_{m-1}^2$$

Coin tossing with $n = 20$, $k = 4$

- class one: observed 4 heads, expected 10
- class two: observed 16 tails, expected 10

$$\sum_{i=1}^2 \frac{(O_i - E_i \pm 0.5)^2}{E_i} = \frac{(4 - 10 + 0.5)^2}{10} + \frac{(16 - 10 - 0.5)^2}{10}$$

$$= 3.025 \times 2 = 6.05 \text{ (P-value} = 0.014)$$

Exactly the same as Score test without correction!

Critical values of χ_1^2

- For 1 d.f. can be computed by squaring Normal values:
 - $P=0.05$: 3.84 ($=1.96^2$)
 - $P=0.01$: 6.63 ($=2.57^2$)
- More degrees of freedom:
 - Use Tables

Likelihood Ratio Test

- Two hypothesis considered
 - Null, H_0 : the parameter \mathbf{p} has some fixed value
 - $\mathbf{p} = \mathbf{p}_0$
 - Alternative, H_1 : the parameter \mathbf{p} has a value which maximises the likelihood of getting the observed data
 - $\mathbf{p} = \mathbf{p}'$
- Let the probability of the data, computed under
 - Null $L(K | p_0)$
 - Alternative $L(K | \hat{p})$

Likelihood Ratio Test

■ Then
$$LRT = 2 \log_e \left[\frac{L(K | \hat{p})}{L(K | p_0)} \right] \infty \chi_1^2$$

More general:

- Hierarchical hypotheses are considered
- The number of degrees of freedom is the difference in number of parameters under estimation

Likelihood Ratio Test

- Coin tossing with $n = 20$, $k = 4$
 - $H_0: p_0 = 1/2$
 - $L_0 = C(20,4) (1/2)^{20}$
 - $H_1: p' = 4/20$
 - $L_1 = C(20,4) (4/20)^4 (16/20)^{16}$

$$LRT = 2 \log_e \left[\frac{L(K | \hat{p})}{L(K | p_0)} \right] =$$
$$2 \log_e \left[\frac{0.2^4 \cdot 0.8^{16}}{(1/2)^{20}} \right] = 2 \log_e [47.22] = 2 \cdot 3.85 = 7.7$$

P-value = 0.005

Many P's to choose from...

In coin tossing with $n = 20$, $k = 4$

P-value is equal to

- 0.014 (Z = χ^2 -test with continuity correction)
- **0.012 (exact binomial)**
- 0.007 (Z = χ^2 -test without continuity correction)
- 0.005 (Likelihood Ratio Test)
- 0.002 (Wald test with continuity correction)
- 0.001 (Wald test without continuity correction)

Do DO continuity correction!

When numbers are small, large numbers approximation may work poor

Importance of large numbers

If $n = 200$, $k = 82$

P-value is equal to

- **0.013 (exact binomial)**
- 0.013 (Z-test with continuity correction)
- 0.012 (Wald test with continuity correction)
- 0.011 (Z-test without continuity correction)
- 0.011 (LRT)
- **0.010 (Wald test without continuity correction)**