

MS-C1620 Statistical inference

4 Inference for binary data

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Binary observations

In many applications the observations are binary.

- Something is true/false.
- Something happened/did not happen.
- Someone belongs/does not belong to a group.

Such observations are conveniently coded as 0/1-valued *indicator variables*.

Recall that if we have a iid sample of binary observations, their distribution is necessarily the *Bernoulli distribution*.

Bernoulli distribution

A random variable x has the *Bernoulli distribution* with the *success probability* θ if,

$$\mathbb{P}(x = 1) = \theta \quad \text{and} \quad \mathbb{P}(x = 0) = 1 - \theta.$$

The expected value and variance of x are

$$\begin{aligned}\mathbb{E}(x) &= \theta \\ \text{Var}(x) &= \theta(1 - \theta).\end{aligned}$$

Note that the Bernoulli distribution has only a single parameter to estimate (no separate “variance parameter”).

The sum of n i.i.d. Bernoulli random variables with the success probability θ has the binomial distribution with the parameters n and θ .

(Often the success probability is called p , but here we use θ to emphasize that it is a parameter to estimate, and to avoid confusion with p-values in hypothesis testing.)

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Approximate confidence interval

Central limit theorem can be used to obtain a confidence interval for the success probability θ of a Bernoulli distribution.

Let x_1, x_2, \dots, x_n be an i.i.d. sample from the Bernoulli distribution with the success probability/expected value θ .

For large n , a level $100(1 - \alpha)\%$ confidence interval for the success probability θ is obtained as

$$\left(\hat{\theta} - z_{\alpha/2} \frac{\sqrt{\hat{\theta}(1 - \hat{\theta})}}{\sqrt{n}}, \hat{\theta} + z_{\alpha/2} \frac{\sqrt{\hat{\theta}(1 - \hat{\theta})}}{\sqrt{n}} \right),$$

where $\hat{\theta}$ is the observed proportion of successes and $z_{\alpha/2}$ is the $(1 - \alpha/2)$ -quantile of the standard normal distribution.

One-sample proportion test

To test whether the success probability of a Bernoulli distribution equals some pre-specified value, we employ **one-sample proportion test**.

One-sample proportion test, assumptions

Let x_1, x_2, \dots, x_n be an i.i.d. sample from a Bernoulli distribution with the success probability θ .

One-sample proportion test, hypotheses

$$H_0 : \theta = \theta_0 \quad H_1 : \theta \neq \theta_0.$$

One-sample proportion test

One-sample proportion test, test statistic

- The test statistic,

$$C = \sum_{i=1}^n x_i,$$

follows the binomial distribution with parameters n and θ_0 under H_0 .

- Under H_0 , the test statistic has $E[C] = n\theta_0$ and $\text{Var}(C) = n\theta_0(1 - \theta_0)$ and both large and both **large** and **small** values of the test statistic suggest that the null hypothesis H_0 is false.

The distribution of the test statistic C is tabulated and statistical software calculate exact p -values of the test.

Asymptotic one-sample proportion test

If the sample size is large, then under the null hypothesis H_0 the standardized test statistic,

$$Z = \frac{\hat{\theta} - \theta_0}{\sqrt{\theta_0(1 - \theta_0)/n}},$$

where $\hat{\theta}$ is the unbiased estimator $\hat{\theta} = \frac{1}{n} \sum_{i=1}^n x_i$ of the parameter θ , follows approximately the standard normal distribution.

The approximation is usually accurate enough if $n\hat{\theta} > 10$ and $n(1 - \hat{\theta}) > 10$. For smaller samples one should use the exact distribution of the test statistic C .

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Two-sample proportion test

The one-sample proportion test can be seen as the equivalent of t -test when the normal distribution is replaced by the Bernoulli distribution.

As with t -test, a two-sample version readily follows and in **two-sample proportion test** parameters of two independent Bernoulli-distributed samples are compared.

Two-sample proportion test, assumptions

Let x_1, x_2, \dots, x_n be an i.i.d. sample from a Bernoulli distribution with the success probability θ_x and let y_1, y_2, \dots, y_m be an i.i.d. sample from a Bernoulli distribution with the success probability θ_y . Furthermore, let the two samples be independent.

Two-sample proportion test, hypotheses

$$H_0 : \theta_x = \theta_y \quad H_1 : \theta_x \neq \theta_y.$$

Two-sample proportion test

Two-sample proportion test, test statistic

- Calculate the sample proportions

$$\hat{\theta}_x = \frac{1}{n} \sum_{i=1}^n x_i, \quad \hat{\theta}_y = \frac{1}{m} \sum_{i=1}^m y_i, \quad \hat{\theta} = \frac{n\hat{\theta}_x + m\hat{\theta}_y}{n + m}.$$

- The test statistic

$$Z = \frac{\hat{\theta}_x - \hat{\theta}_y}{\sqrt{\hat{\theta}(1 - \hat{\theta})\left(\frac{1}{n} + \frac{1}{m}\right)}},$$

follows *for large n* under H_0 the standard normal distribution.

- Values **far from zero** (positive or negative) suggest that the null hypothesis H_0 is false.

The normal approximation is usually good enough if $n\hat{\theta}_x > 5$, $n(1 - \hat{\theta}_x) > 5$, $m\hat{\theta}_y > 5$ and $m(1 - \hat{\theta}_y) > 5$.

Frequency tables

Assuming a “paired binary sample”, the previous test is no longer valid.

id	X	Y
1	0	1
2	0	0
3	0	1
4	1	1
\vdots	\vdots	\vdots

This kind of data is conveniently represented in a **contingency table** (aka **cross tabulation**).

	Y = 0	Y = 1
X = 0	173	40
X = 1	65	53

Inference from contingency tables is discussed next time.

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Lecture quiz

A lecture quiz to determine what you have learned thus far!

Answer the following questions on your own or in small groups.

Lecture quiz

Question 1

Consider the following random sample: 5, -4, -2, 2. Calculate the following sample quantities:

- 1 Sample mean
- 2 Sample standard deviation
- 3 Sample median
- 4 Sample median absolute deviation
- 5 Sample range
- 6 Signs of the sample points
- 7 Ranks of the sample points
- 8 Signed ranks of the sample points with respect to distance to 0.

Lecture quiz

Question 2

Give concrete examples when you would/would not use the following measures of location:

- 1 Sample mean
- 2 Sample median
- 3 Mode

Question 3

Give concrete examples when you would/would not use the following measures of scatter:

- 1 Standard deviation
- 2 Median absolute deviation
- 3 Sample range

Lecture quiz

Question 4

What does it mean in practice if:

- The confidence interval of a parameter is narrow
- The significance level of a test is set to small value
- The p -value of a test is high
- Type I error occurs in a statistical test
- Type II error occurs in a statistical test

Lecture quiz

Question 5

How would you visualize the following samples:

- The heights of the male and female students attending a course.
- The exam points (0-24) on a large course.
- The proportions of faulty products produced by 5 different production lines.
- Stock prices of 3 companies over some time interval
- The monthly salaries and postal codes of adults living in Helsinki area.

Lecture quiz

Question 6

The following plots show the distributions of the test statistics of **A.** *t*-test, **B.** sign test, **C.** signed rank test for the null hypothesis of zero location, with a sample of $n = 10$ points from the standard normal distribution. Which plot is from which test?

