

Hypothesis testing

(1)

Hypothesis: A statement framed in terms of restrictions on the statistical model.

① There is a reason to doubt the hypothesis if the outcomes we observed have low probability when the hypothesis is true.

Ex: Coin tossed \rightarrow

$$\text{claim } P(\text{tails}) = 0.8 \text{ when tossed}$$

Tossed 10 times and lands tails up 4 times.

Binomial ${}^{10}C_4 (0.8)^4 (0.2)^6 = 0.005$

when the claim is true. We can thus take the outcomes of the experiment as evidence against the claim.

on the other hand, 7 tails in the 10 tosses

$${}^{10}C_7 (0.8)^7 (0.2)^3 = 0.201$$

when the claim is true, and we can have little reason to doubt the doubt the claim.

Statistical Hypothesis:

A statistical Hypothesis is some statement or assertion about a population or equivalently about the probability distribution characterising a population which we want to verify on the basis of information available from the sample.

Test of Statistical Hypothesis:

A test of a statistical hypothesis is a two action decision problem after the experimental sample values have been obtained, the two actions being the acceptance or rejection of the hypothesis under consideration.

A statistical test procedure consists of

1. A statistical model
2. A null and alternative hypothesis expressed in terms of the statistical model
3. A function of the data, $T(x_1, x_2, \dots, x_n)$, the test statistics
4. A decision rule :

reject H_0 (Null Hypothesis) in favor of H_1 for $T \in C$ (critical region)
and fail to reject H_0 for $T \notin C$.

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Null and Alternative Hypothesis.

If x_1, x_2, \dots, x_n is a random sample of size n from a normal population with mean μ and variance σ^2 , then the hypothesis

$$H_0: \mu = \mu_0, \sigma^2 = \sigma_0^2$$

is a simple hypothesis
(Null Hypothesis)

Whereas each of the following hypothesis are composite hypothesis are
(Alternative Hypothesis).

i) $\mu = \mu_0$

ii) $\sigma^2 = \sigma_0^2$

iii) $\mu < \mu_0, \sigma^2 = \sigma_0^2$

iv) $\mu > \mu_0, \sigma^2 = \sigma_0^2$

v) $\mu = \mu_0, \sigma^2 < \sigma_0^2$

vi) $\mu = \mu_0, \sigma^2 > \sigma_0^2$

vii) $\mu < \mu_0, \sigma^2 > \sigma_0^2$

Null Hypothesis: This refers to any hypothesis we wish to test and denoted by H_0 . The rejection of H_0 leads to the acceptance of an alternative hypothesis denoted by H_1 .

Reject H_0 : in favour of H_1 because of sufficient evidence in the data

Fail to reject H_0 : because of insufficient evidence in the data.

(Probability of defective)

$$H_0: P = 0.10$$

$$H_1: P > 0.10$$

Now 12 defective items 100 does not proof a $P = 0.10$ so the conclusion "fail to reject". However, 20 defective out of 100 then 'reject H_0 ' in favour of $H_1: P > 0.10$.

Testing a Statistical Hypothesis:

A certain type of cold vaccine is known to be only 25% effective after a period of 2 years.

$$P =$$

20 people chosen

$$\frac{8}{20} = 0.4$$

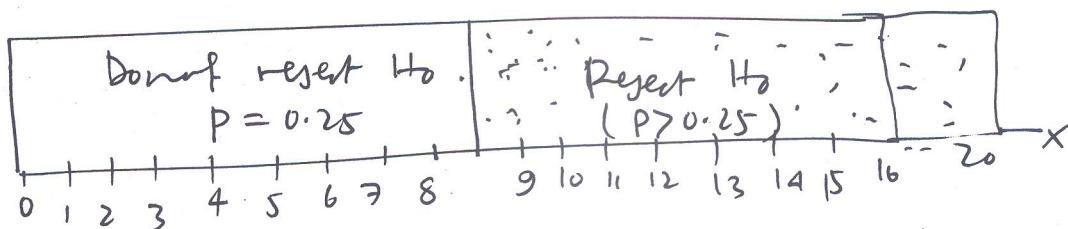
If more than 8 people survive 2 years without contracting the virus, then the new vaccine will be considered superior than previous one.

$$H_0: P = 0.25$$

$$H_1: P > 0.25$$

$$\frac{8}{20} = 0.4$$

Test Statistics:



The critical value is 8

if $x > 8 \rightarrow$ reject H_0 in favor of H_1
if $x \leq 8 \rightarrow$ we fail to reject H_0 .

"The new vaccine may be no better than the one now in use"

$x > 8 \rightarrow$ ~~reject H_0~~

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⑥ Two types of Errors:

The decision to accept or reject the null hypothesis H_0 is made on the basis of the information supplied by the observed sample observations. The conclusion drawn on the basis of a particular sample may not always be true in respect of the population. The four possible situations that arise in any test procedure are given in the following table

- ⑩ Double dichotomy relating to decision and hypothesis^{(to oppose parts) / knowability}:

		Decision From Sample	
		Reject H_0	Accept H_0
True Statement	H_0 : True	Wrong (Type I error)	Correct
	H_0 : False (H_1 True)	Correct	Wrong Type II error

Errors of Type I and Type II:

The error of rejecting H_0 (accepting H_1) when H_0 is true is called Type I error, and the error of accepting H_0 when H_0 is false (H_1 is true) is called Type II error. The probabilities of Type I and Type II errors are denoted by α and β respectively.

- α = Probability of rejecting H_0 when H_0 is true
- β = Probability of accepting H_0 when H_0 is false.
- In quality control: $\alpha \rightarrow$ producer's risk
- $\beta \rightarrow$ consumer's risk.

$$\alpha = P(\text{type I error})$$

$$= P(X > 8 \text{ when } P = \frac{1}{4})$$

$$= \sum_{x=9}^{20} b(x; 20, \frac{1}{4})$$

$$= 1 - \sum_{n=0}^8 b(n; 20, \frac{1}{4})$$

$$= 1 - 0.9591$$

$$= \underline{0.0409}$$

$$\begin{array}{c} H_0: P = \frac{1}{4} \\ H_1: P = \frac{1}{2} \end{array}$$

$\beta = P(\text{type II error})$

$$= P(X \leq 8 \text{ when } P = \frac{1}{2})$$

$$= \sum_{n=0}^8 b(n; 20, \frac{1}{2})$$

$$= 0.2571$$

$$1 - \beta = 1 - 0.2571$$

Level of Significance: α ,

The probability of type I error, is known as the level of significance of the test. It is also known as the size of the critical region.

Power of the test: $1 - \beta$ is called the power function of the test hypothesis H_0 against the alternative hypothesis H_1 . The value of the power function at a particular point is called the power of the test at that point.

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Let p be the probability that a coin will fall head in a single toss in order to test $H_0: p = \frac{1}{2}$ against $H_1: p = \frac{3}{4}$. The coin is tossed 5 times and H_0 is rejected if more than 3 heads are obtained. Find the probability of type I error and power of the test.

Soln:
Use $H_0: p = \frac{1}{2}$
and $H_1: p = \frac{3}{4}$

If the random variable x denotes the number of heads in 5 tosses of a coin then
 $x \sim B(n, p)$ so that

$$\begin{aligned} P(x=x) &= \binom{n}{x} p^x (1-p)^{n-x} \\ &= \binom{5}{x} \left(\frac{1}{2}\right)^x \left(\frac{3}{4}\right)^{5-x} \end{aligned}$$

$$\begin{aligned} \alpha &= P(\text{type I error}) \\ &= P(x \geq 4 \text{ when } p = \frac{1}{2}) \\ &= P(x=4 | p = \frac{1}{2}) + P(x=5 | p = \frac{1}{2}) \\ &= \binom{5}{4} \left(\frac{1}{2}\right)^4 \left(\frac{1}{2}\right)^{5-4} + \binom{5}{5} \left(\frac{1}{2}\right)^5 \\ &= \frac{3}{16} \end{aligned}$$

$$\begin{aligned} \beta &= \text{Probability of Type II error} \\ &= 1 - [P(x=4 | p = \frac{3}{4}) + P(x=5 | p = \frac{3}{4})] \\ &= 1 - \left[\binom{5}{4} \left(\frac{3}{4}\right)^4 \left(\frac{1}{4}\right) + \binom{5}{5} \left(\frac{3}{4}\right)^5 \right] \\ &= 1 - \frac{81}{256} \\ &= \frac{175}{256} \end{aligned}$$

\therefore Power of the test is
 $1 - \beta = \frac{18}{256} //$

Electronic parts

⑪ An electronic manufacturer believes that the proportion of orders for raw materials arriving late is $P = 0.6$. If a random sample of 10 orders shows that 3 or fewer arrived late, the hypothesis that $P = 0.6$ should be rejected in favor of the alternative $P < 0.6$. Use the binomial distribution.

- a) Find the probability of committing a type-I error if $P = 0.6$
- b) Find the probability of committing a type-II error for $P = 0.3$, $P = 0.4$ and $P = 0.5$

Try

10.8

(10.8)

① One and two tailed test

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A test of any statistical hypothesis, where the alternative is one sided, such as

$$H_0: \theta = \theta_0$$

$$H_1: \theta > \theta_0$$

& perhaps,

$$H_0: \theta = \theta_0,$$

$$H_1: \theta < \theta_0$$

is called a one tailed test.

on the other hand where the alternative is two sided, such as :

$$H_0: \theta = \theta_0$$

$$H_1: \theta \neq \theta_0, (\theta < \theta_0 \text{ or } \theta > \theta_0)$$

is called a two sided.

Example | 10.1 → one tail test.
| 10.2 → ~~two~~ Two tail test.

Significance level:

- If the probability of a value of the test statistic 'as bad as worse' as that obtained is P then we reject the null hypothesis if P is less than or equal to the significance level α .
- The significance level is the level of probability that we call unlikely. If your test gives a probability as unlikely as the significance level then you reject the null hypothesis. The usual significance level is 5% or 0.05 but other levels such as 1% (0.01) or 10% (0.1) are often used.

Example:

4 - people playing cards
So winning $P(\text{winning each person}) = \frac{1}{4}$

claim Sarah cheating.

$$H_0: P = \frac{1}{4} \quad (\text{innocent})$$

$$H_p: P > \frac{1}{4} \quad (\text{guilty / not innocent})$$

If Sarah wins 5 out of 10 then proportion is $\frac{1}{2}$. (as bad as worse).
 $X \sim B(10, \frac{1}{4})$

$$\begin{aligned} \therefore P(X \geq 5 | X \sim B(10, \frac{1}{4})) &= 1 - P(X \leq 4 | X \sim B(10, \frac{1}{4})) \\ &= 1 - 0.9219 \\ &= 0.0781 \end{aligned}$$

ie, 7.8% is reasonably large (more than 5%)
so there is no reason to suspect ^{the validity of H_0 and} Sarah remains innocent. There is insufficient evidence to reject the null hypothesis.

(b)

What value of x would provide sufficient evidence to reject the claim that

$P = \frac{1}{4}$? i.e., Sarah is guilty.

If we use a 5% significance level then we need the value of c such that

$$P(X \geq c) \leq 0.05 \quad \text{where } X \sim B(10, 0.25)$$

~~$i.e., 1 - P(X \leq c) \leq 0.05$~~

$$1 - P(X \leq c-1) \leq 0.05$$

$$\underline{P(X \leq c-1) > 0.95}$$

$$P(X \leq 4) = 0.9219$$

$$P(X \leq 5) = 0.9803$$

$$i.e., c-1 = 5 \quad \text{or} \quad c = 6$$

$$\text{we have } P(X \geq 6) = 0.0197 = 1.97\% (< 5\%)$$

So if Sarah had won 6 or more games out of 10 we would had a significant result to reject H_0 .

↓ i.e. we reject H_0 at the 5% level of significance.

↑ $X \geq 6 \rightarrow$ critical region
 $6 \rightarrow$ critical value.