

Probability and Statistics: To p, or not to p?

Module Leader: Dr James Abdey

5.4 Testing a population mean claim

We consider the **hypothesis test of a population mean** in the context of a claim made by a manufacturer.

As an example, the amount of water in mineral water bottles exhibits slight variations attributable to the bottle-filling machine at the factory not putting in *identical* quantities of water in each bottle. The labels on each bottle may state '500 ml' but this equates to a claim about the average contents of all bottles produced (in the population of bottles).

Let X denote the quantity of water in a bottle. It would seem reasonable to assume a normal distribution for X such that:

$$X \sim N(\mu, \sigma^2)$$

and we wish to test:

$$H_0: \mu = 500 \text{ml}$$
 vs. $H_1: \mu \neq 500 \text{ml}$.

Suppose a random sample of n = 100 bottles is to be taken, and let us assume that $\sigma = 10$ ml. From our work in Section 4.5 we know that:

$$\bar{X} \sim N\left(\mu,\,\frac{\sigma^2}{n}\right) = N\left(\mu,\,\frac{(10)^2}{100}\right) = N(\mu,\,1).$$

Further suppose that the sample mean in our random sample of 100 is $\bar{x} = 503$ ml. Clearly, we see that:

$$\bar{x} = 503 \neq 500 = \mu$$

where 500 is the claimed value of μ being tested in H₀.

The question is whether the difference between $\bar{x} = 503$ and the claim $\mu = 500$ is:

- (a) due to sampling error (and hence H₀ is true)?
- (b) statistically significant (and hence H₁ is true)?

Determination of the p-value will allow us to choose between explanations (a) and (b).

We proceed by standardising X such that:

$$Z = rac{ar{X} - \mu}{\sigma / \sqrt{n}} \sim N(0, 1)$$

acts as our **test statistic**. Note the test statistic includes the effect size, $\bar{X} - \mu$, as well as the sample size, n.

Using our sample data, we now obtain the test statistic value (noting the influence of both the effect size and the sample size, and hence ultimately the influence on the *p*-value):

$$\frac{503 - 500}{10/\sqrt{100}} = 3.$$

The p-value is the probability of our test statistic value or a more extreme value conditional on H_0 . Noting that $H_1: \mu \neq 500$, 'more extreme' here means a z-score > 3 and < -3. Due to the symmetry of the standard normal distribution about zero, this can be expressed as:

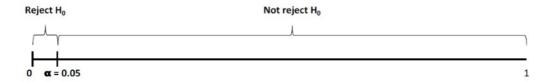
$$p$$
-value = $P(Z > |3|) = 0.0027$.

Note this value can easily be obtained using Microsoft Excel, say, as:

=NORM.S.DIST(
$$-3$$
)*2 or =(1-NORM.S.DIST(3))*2

where the function NORM.S.DIST(z) returns $P(Z \le z)$ for $Z \sim N(0, 1)$.

Recall the **p-value decision rule**, shown below for $\alpha = 0.05$:



Therefore, since 0.0027 < 0.05 we reject H_0 and conclude that the result is 'statistically significant' at the 5% significance level (and also, of course, at the 1% significance level). Hence there is (strong) evidence that $\mu \neq 500$. Since $\bar{x} > \mu$ we might go further and suppose that $\mu > 500$.

Finally, recall the possible decision space:

		Decision made	
		H_0 not rejected	H_0 rejected
True state	H_0 true	Correct decision	Type I error
of nature	H_1 true	Type II error	Correct decision

As we have rejected H_0 this means one of two things:

- ullet we have correctly rejected H_0
- we have committed a Type I error.

Although the p-value is very small, indicating it is $highly\ unlikely$ that this is a Type I error, unfortunately we cannot be certain which outcome has actually occurred!