

22S:105 Statistical Methods and Computing

Two kinds of two-sample t-tests

Lecture 16
Mar. 11, 2019

Paired samples

- We are interested in the unknown population means μ_1 and μ_2 of two different populations.
- In our sample, each observation drawn from the first population is matched up with an observation drawn from the second population.

Two-sample t-tests

So far we have talked about drawing inference about a single population mean μ based on data contained in one sample drawn from that population.

Now we will consider procedures for comparing two *different* population means.

There are different procedures depending on whether the samples are

- paired
- independent

- *self-pairing*: two measurements are taken on each subject

Example:

- systolic blood pressure (sbp) upon entry into a clinical study
- sbp after 1 month on treatment

The population means of interest are

- μ_1 = mean sbp of untreated patients of this type
- μ_2 = mean sbp of patients of this type after 1 month of treatment with the study regimen
- The question of interest is whether the treatment lowers blood pressure, i.e. is $\mu_2 < \mu_1$?

- *matched pairs*: investigator matches each subject in one treatment group with one subject in another treatment group so that members of a pair are as alike as possible

The population means of interest are

- μ_1 = mean response (say sbp at 1 month) of patients receiving treatment 1
- μ_2 = mean response of patients receiving treatment 2
- The question of interest is whether $\mu_1 = \mu_2$

come OC users. This subgroup will be the study sample.

- Measure the sbp of the study sample at the follow-up visit.
- We will compare the baseline and follow-up sbps of the women in the study sample.

Paired t-test

To carry out the hypothesis test of interest, we apply one-sample procedures to the *differences* between values measured on members of each pair.

Example:

- We are interested in whether the use of oral contraceptive (OC) drugs affects the level of systolic blood pressure (sbp) in women.
- We identify a group of nonpregnant, premenopausal women aged 16-49 from a prepaid health plan who are not currently OC users and measure their sbp, which we will refer to as baseline sbp.
- We rescreen these women 1 year later to ascertain a subgroup who have remained non-pregnant throughout the year and have be-

We will do a two-sided test, because we do not know in advance whether to expect μ_1 (mean sbp in OC users) to be higher or lower than μ_2 (mean sbp in non-users).

$$\begin{aligned} H_0 : \mu_1 &= \mu_2 \\ H_a : \mu_1 &\neq \mu_2 \end{aligned}$$

or equivalently:

$$\begin{aligned} H_0 : \mu_1 - \mu_2 &= 0 \\ H_a : \mu_1 - \mu_2 &\neq 0 \end{aligned}$$

or equivalently:

$$\begin{aligned} H_0 : \delta &= 0 \\ H_a : \delta &\neq 0 \end{aligned}$$

where δ denotes $\mu_1 - \mu_2$.

We will use the *observed differences* between the before and after values observed on each woman as our data to carry out the hypothesis test regarding δ at the .05 significance level.

```
data sbpoc ;
infile '/group/ftp/pub/kcowles/datasets/sbpoc.dat' ;
input sbpnooc sbpoc ;
diff = sbpoc - sbpnooc ;
run ;
```

```
proc print ;
run ;
```

| OBS | SBPNOOC | SBPOC | DIFF |
|-----|---------|-------|------|
| 1 | 115 | 128 | 13 |
| 2 | 112 | 115 | 3 |
| 3 | 107 | 106 | -1 |
| 4 | 119 | 128 | 9 |
| 5 | 115 | 122 | 7 |
| 6 | 138 | 145 | 7 |
| 7 | 126 | 132 | 6 |
| 8 | 105 | 109 | 4 |
| 9 | 104 | 102 | -2 |
| 10 | 115 | 117 | 2 |

We will compute the sample mean of the d_i s

$$\bar{d} = \frac{\sum_i^n d_i}{n}$$

and the sample standard deviation of the d_i s

$$s_d = \sqrt{\frac{\sum_i^n (d_i - \bar{d})^2}{n - 1}}$$

```
proc means data = sbpoc ;
var diff ;
run ;
```

Analysis Variable : DIFF

| N | Mean | Std Dev | Minimum | Maximum |
|----|-------|----------|---------|---------|
| 10 | 4.800 | 4.565716 | -2.0000 | 13.0000 |

Then the t statistic is

$$t = \frac{\bar{d} - \delta_0}{s_d / \sqrt{n}}$$

From our data,

$$\begin{aligned} \bar{d} &= 4.80 \\ s_d &= 4.566 \\ t &= \frac{4.80}{4.566 / \sqrt{10}} \\ &= 3.32 \end{aligned}$$

Using Table c, we see that the value that cuts off the upper .025 area under a t distribution with 9 degrees of freedom is 2.262.

Because $3.32 > 2.262$ (our result is more extreme than the required cutoff), we can reject the null hypothesis at the .05 level.

We could use SAS to find the exact p-value, which is 0.0089.

Note that the one-sample t-test in **proc univariate** by default tests the null hypothesis that $\mu = 0$.

```
proc univariate data = sbpoc ;
var diff ;
run ;
```

```
      The UNIVARIATE Procedure
      Variable:  temp
```

```
Tests for Location: Mu0=0
```

| Test | -Statistic- | -----p Value----- |
|-------------|-------------|-------------------|
| Student's t | t 3.324651 | Pr > t 0.0089 |
| Sign | M 3 | Pr >= M 0.1094 |
| Signed Rank | S 24 | Pr >= S 0.0117 |