

Biometry. Lecture 14

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1 One-dimensional data

- Tests for proportions

2 Two-dimensional statistics

- Hypotheses and tests
- Tests for the independence of two variables



1 One-dimensional data

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```
> setwd("<working folder>")  
or  
"Change dir"  
in menu!
```

On Mac, be sure that startup option is working: `getwd()`
(`getwd()` checks if R is in working folder, `dir()` checks the folder content)



One-dimensional data

Tests for proportions



Why we need to test proportions

- Proportions are secondary data
- The main question is: how well the proportion calculated from sample represents the population proportion?
- Null is that proportion of sample does not differ significantly from population proportion



Smokers and non-smokers example

- In hospital, among lung cancer patients, 356 from 476 are smokers ($\approx 75\%$)
- However, among all patients this proportion is lightly lower.
- How well our sample (lung cancer group) represents the whole hospital? In other words, is the deviation we see accidental?



Exact binomial test

```
> binom.test(x=356, n=476, p=0.7, alternative="two.sided")
```

"two.sided" means that the deviation may be to the both possible sides. It was possible to write "greater" instead; in this case we would test if the proportion in our sample is bigger. One-sided tests are normally more powerful but you should **never** use two and one-sided tests together (this is not far from falsification of results)!



Proportion test

Proportion tests are more universal than binomial, but return very similar results:

```
> prop.test(x=356, n=476, p=0.7, alternative="two.sided")
```



Voters example

In the exit poll, 262 persons were questioned. 136 ($\approx 53\%$) said they voted for the candidate A. Check if candidate A won.

```
> prop.test(x=136, n=262, p=.5, alt="greater")
```

1-sample proportions test with continuity correction

data: 136 out of 262, null probability 0.5

X-squared = 0.3092, df = 1, p-value = 0.2891

alternative hypothesis: true p is greater than 0.5

95 percent confidence interval:

0.4664802 1.0000000

sample estimates:

p

0.519084



Two-dimensional statistics

Hypotheses and tests



Hypotheses are cornerstones of science

- The inferential science is based on hypotheses construction and calculation of their probability.
- The simplest approach is to establish null hypothesis and reject it if needed.
- More complicated approach is to consider null and alternative hypotheses together.



Statistical errors

- Type I error is a false alarm: we accept alternative when null is true
- Type II error is a carelessness: we accept null when alternative is true



Level of significance

- The probability to have greater or equal effect when null hypothesis is true is a p-value
- We may ignore this probability if it is too low, in other words, below the level of significance
- The level of significance is a matter of experience and agreement, it could be 0.05, but sometimes also 0.1 and 0.01
- p-value is related with Type I error



Power

- Probability NOT to make a Type II error is a *power*
- The significance level for the power is normally around 0.8, tests with lesser power should be considered as weak



Two-dimensional statistics

Tests for the independence of two variables



What is tested?

- Null: difference equal to 0 \approx similar \approx related \approx samples came from same population
- Alternative: difference not equal to 0 \approx different \approx non-related \approx samples came from different populations



Tests are based on central values

```
> a <- 51:59  
> b <- 1:9  
> x <- rep(5, 9)  
> t.test(a, b)  
> t.test(b, x)
```

Homoscedasticity, similarity of variance (like in a and b but not like in b an x) is an important assumption of all two variable tests. In R, the Welch correction for **non-homogeneity of variance** is by default applied inside `t.test()`



Paired and non-paired

- Paired: came from one set of objects (e.g., measurements done at different time)
- Non-paired: do not belong to one set of objects



Artificial example of the paired test

```
> set.seed(1); t.test(a, (a+rnorm(9)), paired=T)
```

We introduced here a random noise (`rnorm()` function)



Parametric and non-parametric

- Parametric: Student's, or t-test (in R, with Welch correction for **non-homogeneity of variance**)
- Non-parametric: Wilcoxon tests



Leaves example

```
> leaves <- read.table(  
+ "http://ashipunov.info/data/leaves.txt", h=T)  
> Normality3 <- function(df, p=.05)  
+ {  
+   sapply(df, function(.x)  
+     ifelse(shapiro.test(.x)$p.value > p,  
+     "NORMAL", "NOT NORMAL"))  
+ }  
> Normality3(leaves) # all normal!  
> t.test(leaves[,1], leaves[,2], paired=T)  
> wilcox.test(leaves[,1], leaves[,2], paired=T)  
> t.test(leaves[,1], leaves[,3])  
> wilcox.test(leaves[,1], leaves[,3])
```



Two main questions

- Normal?
- Paired?



Final question (3 points)



Final question (3 points)

Please explain the difference between null and alternative hypotheses.



Finishing...

Save your commands!

`(savehistory(<todaysdate>.r) or File -> Save as... on Mac)`



Summary: most important commands

- `binom.test()` and `prop.test()`—tests for the equality of proportions
- `t.test()`—paired and non-paired two-sample parametric test
- `wilcox.test()`—paired and non-paired two-sample non-parametric test



For Further Reading



A. Shipunov.

Biometry [Electronic resource].

2012—onwards.

Mode of access:

http://ashipunov.info/shipunov/school/biol_240



A. Shipunov, and others.

Visual statistics. Use R!

Ongoing translation from Russian.

