

# TESTS IN R

Tobias Niedermaier

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# PRELUDE

# CONFIDENCE INTERVALS

- it is an interval estimate;
- we construct an interval following a procedure that, if applied in a large number of replications of the experiment, gives intervals which contain the true value  $1 - \alpha$  of the time;
- $1 - \alpha$  is called confidence level;
- usual values are 0.90, 0.95 and 0.99.

# EXAMPLES OF CONFIDENCE INTERVALS

- true mean, gaussian distribution, known variance:

```
1 x <- rnorm(100,mean=1,sd=2)
2 mean.x <- mean(x)
3 low <- mean.x - qnorm(0.975)*2/sqrt(100)
4 up <- mean.x + qnorm(0.975)*2/sqrt(100)
5 c(low,up)
```

```
[1] 0.9227046 1.7066902
```

- true mean, gaussian distribution, unknown variance:

```
1 sd.x <- sd(x)
2 low <- mean.x - qt(0.975, df=99)*sd.x/sqrt(100)
3 up <- mean.x + qt(0.975, df=99)*sd.x/sqrt(100)
4 c(low,up)
```

```
[1] 0.8881171 1.7412777
```

# EXCERCISE 4 TASK 3

# STUDENT'S ONE-SAMPLE TEST

- $H_0 : \mu = \mu_0$
- $H_1 : \mu \neq \mu_0$

`t.test(x, mu= $\mu_0$ , alternative='two-sided')`

`t.test(x, mu= $\mu_0$ , alternative='greater')`

`t.test(x, mu= $\mu_0$ , alternative='less')`

- default alternative is 'two-sided'

```
1 x <- rnorm(100, mean=3, sd=1)
2 t.test(x, mu=1)
```

One Sample t-test

data: x

t = 19.63, df = 99, p-value < 2.2e-16

alternative hypothesis: true mean is not equal to 1

95 percent confidence interval:

2.841634 3.255800

sample estimates:

mean of x

3.048717



# FURTHER EXAMPLES

## Further examples: greater, less

```
1 t.test(x, mu=1, alternative="greater")
```

One Sample t-test

```
data: x
t = 19.63, df = 99, p-value < 2.2e-16
alternative hypothesis: true mean is greater than 1
95 percent confidence interval:
 2.87543      Inf
sample estimates:
mean of x
3.048717
```

```
1 t.test(x, mu=1, alternative="less")
```

One Sample t-test

```
data: x
t = 19.63, df = 99, p-value = 1
alternative hypothesis: true mean is less than 1
95 percent confidence interval:
 -Inf 3.222003
sample estimates:
mean of x
3.048717
```

# TWO-SAMPLE T-TEST (UNEQUAL VARIANCE)

- $Var[X] \neq Var[Y]$
- $H_0 : \mu_X = \mu_Y$

`t.test(x, y, alternative=c('two.sided', 'less', 'greater'))`

Example:

```
1 x <- rnorm(100, mean=4, sd=5)
2 y <- rnorm(100, mean=2, sd=2)
3 t.test(x, y)
```

Welch Two Sample t-test

data: x and y

t = 4.493, df = 129.9, p-value = 1.536e-05

alternative hypothesis: true difference in means is not equal to 0

95 percent confidence interval:

1.237719 3.185308

sample estimates:

mean of x mean of y

4.379350 2.167836

# TWO-SAMPLE T-TEST (EQUAL VARIANCE)

- $Var[X] = Var[Y]$
- $H_0 : \mu_X = \mu_Y$

```
t.test(x, y, var.equal=T, alternative=c('two.sided', 'less',  
'greater'))
```

Example:

```
1 x <- rnorm(100, mean=4, sd=1)
2 y <- rnorm(100, mean=2, sd=1)
3 t.test(x, y, var.equal=T)
```

Two Sample t-test

data: x and y

t = 15.037, df = 198, p-value < 2.2e-16

alternative hypothesis: true difference in means is not equal to 0

95 percent confidence interval:

1.752869 2.282017

sample estimates:

mean of x mean of y

4.082856 2.065413

# F-TEST FOR COMPARING VARIANCES

- $H_0 : \sigma_X^2 = \sigma_Y^2$
- $H_1 : \sigma_X^2 \neq \sigma_Y^2$

```
var.test(x, y, ratio = 1, alternative=c('two.sided', 'less',  
'greater'), conf.level = 0.95)
```

Example:

```
1 x <- rnorm(80, mean=4, sd=5)  
2 y <- rnorm(20, mean=2, sd=2)  
3 var.test(x, y)
```

F test to compare two variances

data: x and y

F = 5.7545, num df = 79, denom df = 19, p-value = 8.826e-05

alternative hypothesis: true ratio of variances is not equal to 1

95 percent confidence interval:

2.571244 10.968061

sample estimates:

ratio of variances

5.754537

# PROPORTION TEST 1

- $H_0 : p = p_0$
- $H_1 : p \neq p_0$

`prop.test(x, n, p)`

- x: number of successes
- n: total number of trials
- p: proportion to be tested.
- Example:

```
1 # Load preprocessed version of the data saved as .RData file
2 load(file="myNhanes.RData")
3 prop.test(sum(mydata$male), length(mydata$male), p=0.5)
```

1-sample proportions test with continuity correction

data: sum(mydata\$male) out of length(mydata\$male), null probability 0.5

X-squared = 0.2738, df = 1, p-value = 0.6008

alternative hypothesis: true p is not equal to 0.5

95 percent confidence interval:

0.4898438 0.5177503

sample estimates:

p

0.5038

# PROPORTION TEST 2

- $H_0 : p_1 = p_2$
- $H_1 : p_1 \neq p_2$

`prop.test(table(x, y))`

- x: first (categorical) variable
- y second (binary) variable
- Example:

```
1 # prop.test(table(educ, male))
2 tbl <- table(mydata$educ, mydata$male)
3 prop.test(tbl)
```

5-sample test for equality of proportions without continuity correction

```
data:  tbl
X-squared = 13.481, df = 4, p-value = 0.00915
alternative hypothesis: two.sided
sample estimates:
```



prop 1	prop 2	prop 3	prop 4	prop 5
0.5218295	0.5082707	0.5292929	0.4605078	0.4881356

# QUICK SUMMARY

```
1 t.test(height[male==T],height[male==F])
2
3 var.test(height[male==T],height[male==F])
```

- and a Z-test, testing the equality of proportions:

```
1 table(male,heartdis_ever,dnn=c("Male","Heart disease"))
2 prop.table(table(male,!heartdis_ever,dnn=c("Male","Heart disease")),1)
3 prop.test(table(male,!heartdis_ever))
```

# CHI-SQUARE TEST

- $H_0 : \pi_1 = \pi_2 = \pi$

```
1 chisq.test(x, y = NULL, correct = TRUE, p = rep(1/length(x), length(x)), rescale.p = FALSE,  
2           simulate.p.value = FALSE, B = 2000)
```

- you can pass to the function:
  - a contingency table;
  - directly two categorical vectors.

## Examples:

```
1 x <- c("M", "F", "M", "F", "M", "M", "F", "F", "M")  
2 y <- c("Y", "N", "N", "N", "Y", "Y", "N", "Y", "N")  
3 chisq.test(x,y)
```

Pearson's Chi-squared test with Yates' continuity correction

data: x and y

X-squared = 0.14062, df = 1, p-value = 0.7077

or, with a contingency table,

```
1 tab <- table(x,y)
2 chisq.test(tab)
```

Pearson's Chi-squared test with Yates' continuity correction

data: tab

X-squared = 0.14062, df = 1, p-value = 0.7077

# FISHER TEST

- exact test

```
1 fisher.test(x, y = NULL, workspace = 200000,  
2     hybrid = FALSE, control = list(), or = 1,  
3     alternative = "two.sided", conf.int = TRUE,  
4     conf.level = 0.95, simulate.p.value = FALSE, B =  
5     2000)
```

- you can pass to the function two factor vectors

## Examples:

```
1 fisher.test(x,y)
```

Fisher's Exact Test for Count Data

data: x and y  
p-value = 0.5238  
alternative hypothesis: true odds ratio is not equal to 1  
95 percent confidence interval:  
0.1497998 312.5621395  
sample estimates:  
odds ratio

3.764195

## ... or a contingency table

```
1 tab <- table(x,y)
2 fisher.test(tab)
```

Fisher's Exact Test for Count Data

data: tab

p-value = 0.5238

alternative hypothesis: true odds ratio is not equal to 1

95 percent confidence interval:

0.1497998 312.5621395

sample estimates:

odds ratio

3.764195

# MCNEMAR TEST

- symmetry of rows and columns;

```
1 mcnemar.test(x, y = NULL, correct = TRUE)
```

- you can pass to the function two factor vectors

## Examples:

```
1 mcnemar.test(x,y)
```

McNemar's Chi-squared test with continuity correction

data: x and y

McNemar's chi-squared = 0, df = 1, p-value = 1

## ...or a contingency table

```
1 tab <- table(x,y)
2 mcnemar.test(tab)
```

McNemar's Chi-squared test with continuity correction

data: tab

# **EXCERCISE 4 TASK 4**



# NON-PARAMETRIC TESTS

# SIGN TEST

- $H_0 : \text{median} = md_0$
- we need the package BSDA

```
1 SIGN.test(x, y = NULL, md = 0, alternative = "two.sided", conf.level = 0.95)
```

## Example:

```
1 library(BSDA)
2 x <- rpois(100,5)
3 SIGN.test(x, md=5)
```

### One-sample Sign-Test

```
data:  x
s = 39, p-value = 0.3374
alternative hypothesis: true median is not equal to 5
95 percent confidence interval:
 4 5
sample estimates:
median of x
      5
```

Achieved and Interpolated Confidence Intervals:

		Conf.Level	L.E.pt	U.E.pt
Lower	Achieved CI	0.9431	4	5

# WILCOXON TEST

- $H_0 : median = md_0$
- test based on ranks

```
1 wilcox.test(x, y = NULL, alternative = c("two.sided", "less", "greater"), mu = 0,  
2           paired = FALSE, exact = NULL, correct = TRUE, conf.int = FALSE,  
3           conf.level = 0.95, ...)
```

## Example:

```
1 x <- rpois(100,5)  
2 wilcox.test(x,mu=5)
```

Wilcoxon signed rank test with continuity correction

data: x

V = 1551.5, p-value = 0.7999

alternative hypothesis: true location is not equal to 5

# RECALL: SIGN TEST VS. WILCOXON TEST VS. T-TEST PART 1

Sign test:

- Applicable for ordinal or higher scale level
- No distributional assumptions
- Low power

Wilcoxon test:

- Applicable for discrete or continuous quantitative data
- Symmetrical distribution required
- Intermediate power

T-test:

- Normal distribution required
- Highest power

# COMPARE ALL TESTS ON THE SAME DATA (WE ASSUME NORMAL DISTRIBUTED DATA!)

```
1 set.seed(123)
2 x <- rnorm(20,0.5)
3 c(
4   signTest = BSDA::SIGN.test(x, md = 0)$p.value,
5   wilcoxonTest =wilcox.test(x,mu=0)$p.value,
6   tTest = t.test(x, mu = 0)$p.value
7 )
```

signTest	wilcoxonTest	tTest
0.04138947	0.01068878	0.00822065

# MANN-WHITNEY U TEST

- $H_0 : median_1 = median_2$
- non-parametric test for independent samples
- package coin (Why extra package and not using `wilcox.test`? -> <https://stats.stackexchange.com/questions/31417/what-is-the-difference-between-wilcox-test-and-coinwilcox-test-in-r>)

```
1 wilcox_test(formula, data, subset = NULL, weights= NULL, ...)
```

- needs a formula, e.g.  $y \sim x$ .

Example hdl in NHANES data (not normal):

```
1 hist(mydata$hdl[mydata$male == T], breaks = 40, col = "steelblue", freq = F, ylim=c(0,1.5))
2 hist(mydata$hdl[mydata$male == F], breaks = 40, col = "coral", add = T, freq = F)
```

```
1 coin::wilcox_test(hdl ~ as.factor(male), data = mydata)
```

Asymptotic Wilcoxon-Mann-Whitney Test



```
data: hdl by as.factor(male) (FALSE, TRUE)
Z = -21.838, p-value < 2.2e-16
alternative hypothesis: true mu is not equal to 0
```

```
1 wilcox.test(hdl ~ as.factor(male), data = mydata)
```

Wilcoxon rank sum test with continuity correction

```
data: hdl by as.factor(male)
W = 1520720, p-value < 2.2e-16
alternative hypothesis: true location shift is not equal to 0
```

# CORRELATION

```
1 cor(X,Y)
2
3 cor(x, y = NULL, use = "everything", method = c("pearson", "kendall", "spearman"))
```

- correlation between x and y
- with incomplete observations, we may want to set use = "complete.obs"
- based on the argument method, the function cor() computes
  - Pearson's r (default)
  - Kendall's  $\tau$  (rank based)
  - Spearman's  $\rho$  (rank based)

## Example:

```
1 x<-rnorm(10)
2 y<-rnorm(10)
3 cor(x,y)
```

# CORRELATION TEST

- test for correlation between paired samples

```
1 cor.test(x, y, alternative = c("two.sided", "less", "greater"), method = c("pearson", "kendall", "spe
```

- $H_1$  is defined through the argument `alternative`
- the argument `method` sets the correlation coefficient

## Example:

```
1 cor.test(x,y)
```

Pearson's product-moment correlation

```
data:  x and y
t = -1.3752, df = 8, p-value = 0.2063
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 -0.8365717  0.2654408
sample estimates:
      cor
-0.4372649
```

# **EXCERCISE 4 TASK 5**