## **Two-sample Hypothesis Testing**

## Задача 1

Проведено е допитване с въпрос: приемате ли увеличаване на цената на цигарите с  $25\,\%$ , като начин за намаляване на тютюнопушенето. 351 от 605 непушачи са отговорили с да, докато при пушачите 71 от 195 отговарят с да. Може ли да се приеме, че мнението на пушачите и непушачите съвпада.

#### Решение:

	Non-Smokers	Smokers
Yes:	351	71
No:	254	124
	605	195

 $p_1$  - the proportion of the non-smokers who reply "yes"

 $p_2$  - the proportion of the smokers who reply "yes"

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

One way to solve the task is with the rejection region approach

```
> n1 <- 605

> p1_hat <- 351 / n1

> n2 <- 195

> p2_hat <- 71 / n2

> p_hat <- (71 + 351) / (n1 + n2)

> z <- ((p1_hat - p2_hat) - 0) / sqrt(p_hat * (1 - p_hat) * (1/n1 + 1/n2))

> z

[1] 5.255544
```

## Let's compute the critical area

```
> alpha <- 0.05
> qnorm(alpha / 2, 0, 1)
[1] -1.959964
> qnorm(1 - alpha/2, 0, 1)
[1] 1.959964
```

So the critical area is  $W_{\alpha}=\{Z\leq -1.96\cup Z\geq 1.96\}$ . In our case z=5.26 and is much higher than 1.96, so it is in the critical area for  $H_0$  and we reject  $H_0$ .

Another way to solve the task is with the **p-value approach** 

```
> 2 * pnorm(z, 0, 1, lower.tail = FALSE)
```

The  $p-value=0.000000148<0.05=\alpha$ , so we reject  $H_0$ . Or we can use the prop.test function

```
> prop.test(c(351, 71), c(605, 195))
```

2-sample test for equality of proportions with continuity correction

data: c(351, 71) out of c(605, 195)

X-squared = 26.761, df = 1, p-value = 2.303e-07

alternative hypothesis: two.sided

95 percent confidence interval:
0.1345204 0.2976051

sample estimates:
prop 1 prop 2
0.5801653 0.3641026

The  $p-value = 0.0000002303 < 0.05 = \alpha$ , so we reject  $H_0$ .

### Задача 2:

Измерено е времето (в дни) за излекуване от дадена болест след прием на ново лекарство 15 10 13 7 9 8 21 9 14 8. Извършено е измерване и на контролна група приемаща пласебо: 15 14 12 8 14 10 7 16 10 15 12. Може ли да се приеме, че новото лекарство подобрява състоянието на пациентитите?

#### Решение:

```
> treatment = c(15, 10, 13, 7, 9, 8, 21, 9, 14, 8)
> placebo = c(15, 14, 12, 8, 14, 10, 7, 16, 10, 15, 12)
```

First we check if the data is normally distributed.

## > library(StatDA)

Warning: package 'StatDA' was built under R version 4.0.3

Loading required package: sgeostat

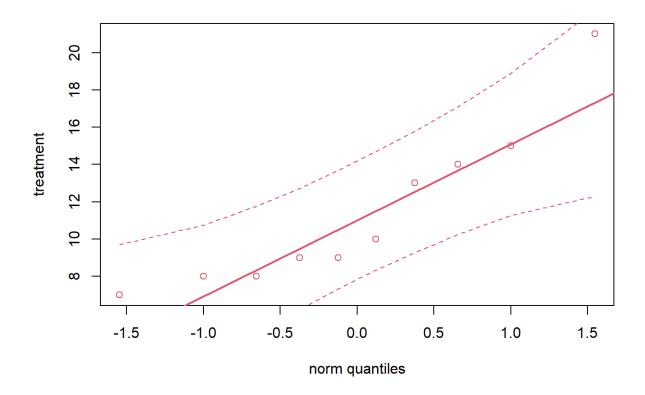
Warning: package 'sgeostat' was built under R version 4.0.3

Registered S3 method overwritten by 'geoR':

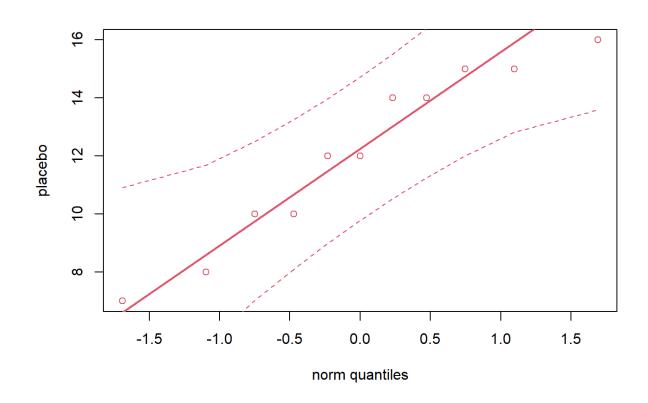
method from

plot.variogram sgeostat

# > qqplot.das(treatment)



# > qqplot.das(placebo)



## > shapiro.test(treatment)

Shapiro-Wilk normality test

data: treatment

W = 0.86663, p-value = 0.09131

The  $p-value = 0.09131 > 0.05 = \alpha$ , so we have no evidence to reject  $H_0$ .

# > shapiro.test(placebo)

Shapiro-Wilk normality test

data: placebo

W = 0.93174, p-value = 0.4287

The  $p-value = 0.4287 > 0.05 = \alpha$ , so we have no evidence to reject  $H_0$ .

We can assume that both are normally distributed, so we can use t-test

 $H_0: \mu_{treatment} = \mu_{placebo}$  $H_A: \mu_{treatment} \neq \mu_{placebo}$ 

# > t.test(treatment, placebo, alternative = "greater")

Welch Two Sample t-test

data: treatment and placebo

t = -0.41894, df = 15.853, p-value = 0.6596

alternative hypothesis: true difference in means is greater than 0

95 percent confidence interval:

-3.571812 Inf sample estimates: mean of x mean of y 11.40000 12.09091

The  $p-value=0.6596>0.05=\alpha$ , so we have no evidence to reject  $H_0$ . Or we can check if the variances are equal

 $H_0: \sigma_{treatment} = \sigma_{placebo}$   $H_0: \sigma_{treatment} \neq \sigma_{treatment}$ 

 $H_A: \sigma_{treatment} \neq \sigma_{placebo}$ 

## > var.test(treatment, placebo)

F test to compare two variances

data: treatment and placebo

F = 2.0827, num df = 9, denom df = 10, p-value = 0.2685 alternative hypothesis: true ratio of variances is not equal to 1

95 percent confidence interval: 0.5511213 8.2554098 sample estimates: ratio of variances 2.082667

The  $p-value=0.2685>0.05=\alpha$ , so we have no evidence to reject  $H_0$ . So we can assume that the variances are equal.

> t.test(treatment, placebo, alternative = "greater", var.equal = TRUE)

Two Sample t-test

data: treatment and placebo
t = -0.42639, df = 19, p-value = 0.6627
alternative hypothesis: true difference in means is greater than 0
95 percent confidence interval:
-3.492741 Inf
sample estimates:
mean of x mean of y
11.40000 12.09091

The  $p-value = 0.6627 > 0.05 = \alpha$ , so we have no evidence to reject  $H_0$ .

## Задача 3

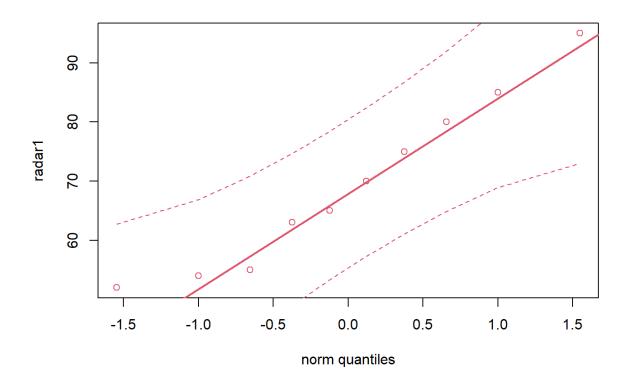
Сравняват се два радара за определяне скоростта на автомобил. Направени са десет наблюдения, измерванията на първия са: 70 85 63 54 65 80 75 95 52 55, а на втория: 72 86 62 55 63 80 78 90 53 57. Да се провери дали двата радара са еднакви.

## Решение:

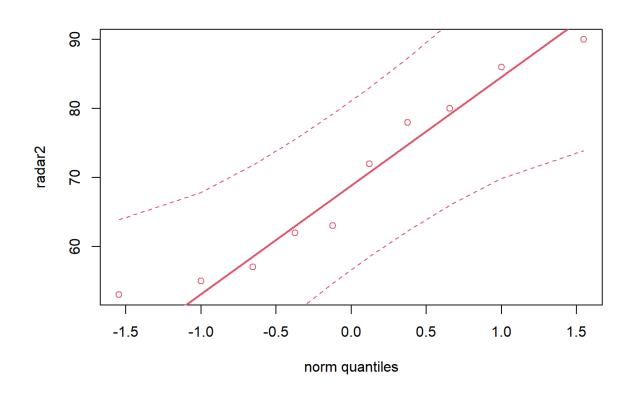
```
> radar1 <- c(70, 85, 63, 54, 65, 80, 75, 95, 52, 55)
> radar2 <- c(72, 86, 62, 55, 63, 80, 78, 90, 53, 57)
```

First we check if the data is normally distributed.

# > qqplot.das(radar1)



# > qqplot.das(radar2)



# > shapiro.test(radar1)

Shapiro-Wilk normality test

data: radar1

W = 0.94879, p-value = 0.6543

The  $p-value = 0.6543 > 0.05 = \alpha$ , so we have no evidence to reject  $H_0$ .

## > shapiro.test(radar2)

Shapiro-Wilk normality test

data: radar2

W = 0.92451, p-value = 0.3961

The  $p-value = 0.3961 > 0.05 = \alpha$ , so we have no evidence to reject  $H_0$ .

We can assume that both are normally distributed, the data is paired, so we can use t-test

 $H_0: \mu_{radar_1} = \mu_{radar_2}$  $H_A: \mu_{radar_1} \neq \mu_{radar_2}$ 

# > t.test(radar1, radar2, paired = TRUE)

Paired t-test

data: radar1 and radar2

t = -0.26941, df = 9, p-value = 0.7937

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

95 percent confidence interval:

-1.879354 1.479354 sample estimates:

mean of the differences

-0.2

The  $p-value=0.7937>0.05=\alpha$ , so we have no evidence to reject  $H_0$ .

#### Задача 4

Разгледайте данните ewr от пакета UsingR. Сравнете времето за напускане на летището от такси обслужващо компаниите American airlines и Northwest airlines.

## Решение:

#### > library(UsingR)

Warning: package 'UsingR' was built under R version 4.0.3

Loading required package: MASS Loading required package: HistData Loading required package: Hmisc Loading required package: lattice Loading required package: survival Loading required package: Formula Loading required package: ggplot2

Attaching package: 'Hmisc'

The following objects are masked from 'package:base':

format.pval, units

Attaching package: 'UsingR'

The following object is masked from 'package:survival':

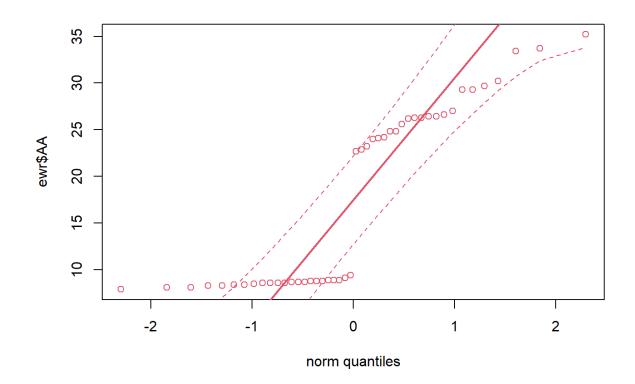
#### cancer

## > head(ewr)

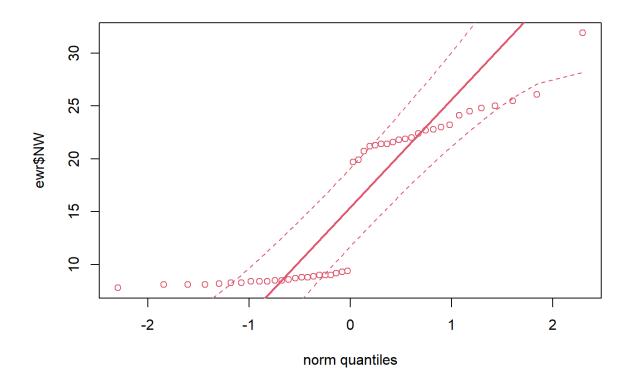
```
Year Month AA CO DL HP NW TW UA US inorout
1 2000 Nov 8.6 8.3 8.6 10.4 8.1 9.1 8.4 7.6
                                             in
2 2000 Oct 8.5 8.0 8.4 11.2 8.2 8.5 8.5 7.8
                                             in
3 2000 Sep 8.1 8.5 8.4 10.2 8.3 8.6 8.2 7.6
                                             in
4 2000 Aug 8.9 9.1 9.2 14.5 9.0 10.3 9.2 8.7
                                              in
5 2000 Jul 8.3 8.9 8.2 11.5 8.8 9.1 9.2 8.2
                                            in
6 2000 Jun 8.8 9.0 8.8 14.9 8.4 10.8 8.9 8.3
                                              in
> ewr_out <- subset(ewr, inorout == "out", select = c("AA","NW"))
> ewr_AA <- ewr_out$AA
> ewr_NW <- ewr_out$NW
```

First we check if the data is normally distributed.

> qqplot.das(ewr\$AA)



# > qqplot.das(ewr\$NW)



# > shapiro.test(ewr\$AA)

Shapiro-Wilk normality test

data: ewr\$AA

W = 0.78856, p-value = 1.092e-06

The  $p-value = 0.000001092 < 0.05 = \alpha$ , so we reject  $H_0$ .

## > shapiro.test(ewr\$NW)

Shapiro-Wilk normality test

data: ewr\$NW

W = 0.79173, p-value = 1.278e-06

The  $p-value = 0.000001278 < 0.05 = \alpha$ , so we reject  $H_0$ .

The data is not normally distributed, so we can use hypothesis for the median

 $H_0: Me(AA) = Me(NW)$  $H_A: Me(AA) \neq Me(NW)$ 

# > wilcox.test(ewr\$AA, ewr\$NW)

Warning in wilcox.test.default(ewr\$AA, ewr\$NW): cannot compute exact p-value

## with ties

Wilcoxon rank sum test with continuity correction

data: ewr\$AA and ewr\$NW W = 1263, p-value = 0.1101

alternative hypothesis: true location shift is not equal to 0

The  $p-value = 0.1101 > 0.05 = \alpha$ , so we have no evidence to reject  $H_0$ .

# Задача 6

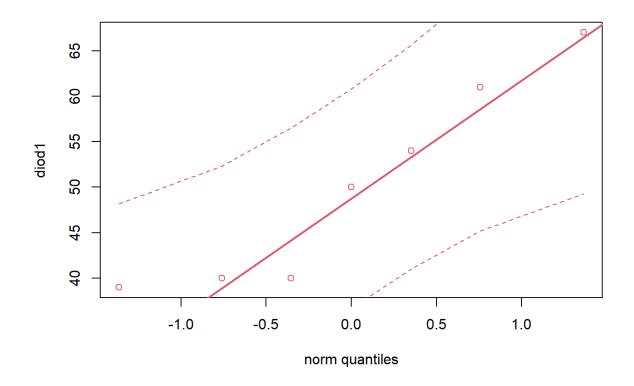
Измервани са напреженията на пробив на диоди от две партиди получени са следните наблюдения: 39, 50, 61, 67, 40, 40, 54 за първата партида и 60, 53, 42, 41, 40, 54, 63, 69 за втората. Може ли да се приеме че диодите имат еднакво напрежение на пробив.

## Решение:

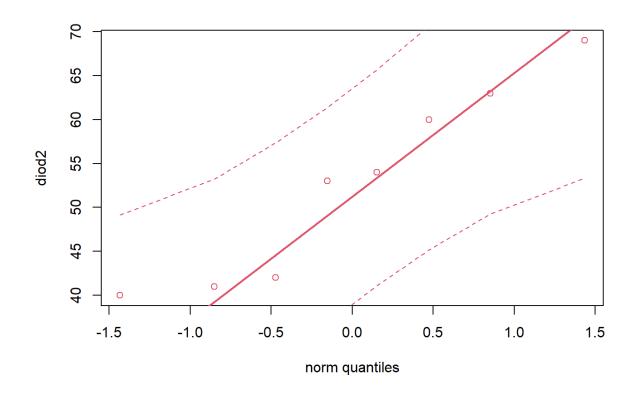
```
> diod1 <- c(39, 50, 61, 67, 40, 40, 54)
> diod2 <- c(60, 53, 42, 41, 40, 54, 63, 69)
```

First we check if the data is normally distributed.

> qqplot.das(diod1)



# > qqplot.das(diod2)



## > shapiro.test(diod1)

Shapiro-Wilk normality test

data: diod1 W = 0.8903, p-value = 0.2762

The  $p-value = 0.2762 > 0.05 = \alpha$ , so we have no evidence to reject  $H_0$ .

# > shapiro.test(diod2)

Shapiro-Wilk normality test

data: diod2 W = 0.9149, p-value = 0.3899

The  $p-value=0.3899>0.05=\alpha$ , so we have no evidence to reject  $H_0$ .

We can assume that both are normally distributed, so we can use t-test

$$\begin{split} H_0: \mu_{diod_1} &= \mu_{diod_2} \\ H_A: \mu_{diod_1} &\neq \mu_{diod_2} \end{split}$$

## > t.test(diod1, diod2)

Welch Two Sample t-test

data: diod1 and diod2
t = -0.45541, df = 12.665, p-value = 0.6565
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
-15.008141 9.793855
sample estimates:
mean of x mean of y
50.14286 52.75000

The  $p-value = 0.6565 > 0.05 = \alpha$ , so we have no evidence to reject  $H_0$ .

Or we can check if the variances are equal

 $H_0: \sigma_{diod_1} = \sigma_{diod_2}$  $H_A: \sigma_{diod_1} \neq \sigma_{diod_2}$ 

## > var.test(diod1, diod2)

F test to compare two variances

data: diod1 and diod2

F = 1.0379, num df = 6, denom df = 7, p-value = 0.9473

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

95 percent confidence interval:

0.2027742 5.9114393

sample estimates:

ratio of variances

1.037919

The  $p-value=0.9473>0.05=\alpha$ , so we have no evidence to reject  $H_0$ . So we can assume that the variances are equal.

## > t.test(diod1, diod2, var.equal = TRUE)

Two Sample t-test

data: diod1 and diod2 t = -0.45602, df = 13, p-value = 0.6559 alternative hypothesis: true difference in means is not equal to 0 95 percent confidence interval: -14.958318 9.744033 sample estimates: mean of x mean of y 50.14286 52.75000

The  $p-value = 0.6559 > 0.05 = \alpha$ , so we have no evidence to reject  $H_0$ .