# Overview statistic tests PB

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## Contents

1	Software	2
	1.1 Versies	2
	1.2 Installatie	
2	T-test	3
	2.1 Independent samples T-test	3
	2.2 Paired samples T-test	4
3	Correlation	6
4	Regression	8
5	One-way independent samples ANOVA	10
6	Factorial independent samples ANOVA 6.1 One-way repeated measures ANOVA	12 17
	6.2 Factorial repeated measures ANOVA	
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br	oncode: $\label{eq:https://github.com/jonasvannijnatten/R_Data_Visualization} \label{eq:https://github.com/jonasvannijnatten/R_Data_Visualization}$	

## 1 Software

#### 1.1 Versies

software versions used for this tutorial:

- R version 3.5.1 (2018-07-02)
- car-package version: 3.0.0 (2018-03-23)

### 1.2 Installatie

Benodigde packages downloaden & installeren:

```
install.packages(pkgs="kableExtra", repos="https://www.freestatistics.org/cran/")
install.packages(pkgs="car", repos="https://www.freestatistics.org/cran/")
```

Benodigde packages activeren:

```
library(package=kableExtra)
library(package=car)
```

Table 1: data.long

ID	condition	score
1	A	34.29326
2	A	25.89247
3	A	22.94641
4	A	20.07555
5	A	24.81194
1	В	35.46555
2	В	35.38522
3	В	37.45495
4	В	42.83193
5	В	36.51421

#### $\mathbf{2}$ T-test

##

#### Independent samples T-test

```
Show code for data generation
# generate data
N = 40
data.long = data.frame(
 ID = 1:N,
 condition = rep(x = c("A","B"), each = N),
 score = c(rnorm(n = N, mean = 25, sd = 6.5), rnorm(n = N, mean = 35, sd = 6.5))
Test for assumption of normality
by(data = data.long$score, INDICES = data.long$condition, FUN = shapiro.test)
## data.long$condition: A
##
##
   Shapiro-Wilk normality test
##
## data: dd[x,]
## W = 0.97503, p-value = 0.511
##
## -----
## data.long$condition: B
##
## Shapiro-Wilk normality test
##
## data: dd[x,]
## W = 0.98031, p-value = 0.701
Test for equality of variances
leveneTest(y = data.long$score, group = data.long$condition)
## Levene's Test for Homogeneity of Variance (center = median)
       Df F value Pr(>F)
## group 1 0.1801 0.6725
        78
##
t.test(formula = score ~ condition, data = data.long, paired=FALSE, alternative="two.sided", var.equal=
##
   Two Sample t-test
```

Table 2: data.long

ID	condition	score
1	A	26.576829
2	A	20.551406
3	A	24.620292
4	A	24.503925
5	A	4.718622
1	В	39.235208
2	В	25.407539
3	В	23.709544
4	В	33.823236
5	В	44.859329

```
## data: score by condition
## t = -7.2228, df = 78, p-value = 2.959e-10
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -12.962753 -7.360894
## sample estimates:
## mean in group A mean in group B
         25.69379
##
                          35.85561
```

#### Paired samples T-test 2.2

Show code for data generation

```
# generate data
N = 30
data.long = data.frame(
 ID = rep(1:N,2),
  condition = rep(x = c("A", "B"), each = N),
  score = c(rnorm(n = N, mean = 25, sd = 6.5), rnorm(n = N, mean = 35, sd = 6.5))
)
```

## 95 percent confidence interval:

-10.43889

## -14.358335 -6.519436 ## sample estimates: ## mean of the differences

##

```
Test for assumption of normality
# calculate difference scores
diffScore = data.long$score[data.long$condition=="A"] - data.long$score[data.long$condition=="B"]
shapiro.test(diffScore)
##
##
   Shapiro-Wilk normality test
##
## data: diffScore
## W = 0.96016, p-value = 0.3126
t.test(formula = score ~ condition, data = data.long, paired=TRUE, alternative="two.sided", var.equal=True
## Paired t-test
##
## data: score by condition
## t = -5.4472, df = 29, p-value = 7.325e-06
\#\# alternative hypothesis: true difference in means is not equal to 0
```

Table 3: data.long

experience	salary
12.47743	9862.848
19.15308	10209.297
11.23352	10123.747
15.21043	10096.900
20.13432	10094.887

#### 3 Correlation

```
# generate data
set.seed(05)
nrobs = 100
experience = rnorm(n = nrobs, mean = 15, sd = 3)
salary = 10000 + (5 * experience) + rnorm(n = nrobs, mean = 0, sd = 100)
data.long = data.frame(experience, salary)
# calculate correlation coefficient r
corr_coef = cor(x = data.long$experience, y = data.long$salary)
rm(list = c("nrobs", "experience", "salary"))
Test for assumption of normality
apply(X = data.long, MARGIN = 2, FUN = shapiro.test)
## $experience
##
##
   Shapiro-Wilk normality test
##
## data: newX[, i]
## W = 0.98711, p-value = 0.445
##
##
## $salary
##
   Shapiro-Wilk normality test
##
##
## data: newX[, i]
## W = 0.99257, p-value = 0.8607
Pearson correlation
cor.test(x=data.long$experience, y=data.long$salary, alternative = "two.sided", method = "pearson")
##
## Pearson's product-moment correlation
##
## data: data.long$experience and data.long$salary
## t = 2.5509, df = 98, p-value = 0.01229
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.05584335 0.42510763
## sample estimates:
##
         cor
## 0.2495245
Spearman correlation
cor.test(x=data.long$experience, y=data.long$salary, alternative = "two.sided", method = "spearman")
##
```

```
## Spearman's rank correlation rho
##
## data: data.long$experience and data.long$salary
## S = 121190, p-value = 0.006181
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
## rho
## 0.2727873
```

Table 4: data.long

experience	salary
12.47743	9862.848
19.15308	10209.297
11.23352	10123.747
15.21043	10096.900
20.13432	10094.887

## 4 Regression

Show code for data generation

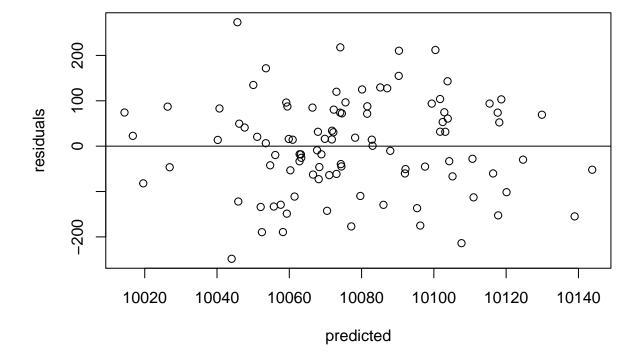
```
# generate data
set.seed(05)
nrobs = 100
experience = rnorm(n = nrobs, mean = 15, sd = 3)
salary = 10000 + (5 * experience) + rnorm(n = nrobs, mean = 0, sd = 100)
data.long = data.frame(experience, salary)
# calculate correlation coefficient r
corr_coef = cor(x = data.long$experience, y = data.long$salary)
rm(list = c("nrobs", "experience", "salary"))
```

Fit linear model

```
linearModel = lm(formula = salary ~ experience, data = data.long)
```

Test for assumption of normality & equal variances

```
# plot the residuals and look at the distribution around the O-line and if the spread is equal over all
plot(x=linearModel$fitted.values, y=linearModel$residuals, xlab = "predicted", ylab = "residuals"); abl:
```



shapiro.test(linearModel\$residuals)

```
##
## Shapiro-Wilk normality test
##
## data: linearModel$residuals
## W = 0.99208, p-value = 0.8268
Regression test
summary(linearModel)
##
## Call:
## lm(formula = salary ~ experience, data = data.long)
## Residuals:
##
    Min
             1Q Median
                            3Q
                                     Max
## -248.24 -61.55 10.13 74.34 273.17
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 9934.674
                        56.812 174.870 <2e-16 ***
## experience
             9.437
                          3.700 2.551
                                         0.0123 *
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 104.4 on 98 degrees of freedom
## Multiple R-squared: 0.06226, Adjusted R-squared:
## F-statistic: 6.507 on 1 and 98 DF, p-value: 0.01229
```

Table 5: data.long

subj	condition	score
1	A	17.537437
2	A	6.978033
21	В	21.403886
22	В	19.120283
41	C	15.895668
42	С	13.543431

## 5 One-way independent samples ANOVA

Toon code om voorbeeld data te genereren

```
set.seed(05)
             # set seed
nrofconds = 3 # set number of conditions
nrofsubs = 20 # set number of subjects
subj = as.factor(1:(nrofsubs*nrofconds)) # create array with subject IDs
condition = as.factor(rep(LETTERS[1:nrofconds],each=nrofsubs)) # create array with condition values
score = as.vector( replicate(
         nrofconds, rnorm(n = nrofsubs, mean = sample(8,1)+10, sd = sample(5,1))
                                           # create array with measurement values
data.long = data.frame(subj, condition, score); # combine arrays into a data.frame
rm(list=c("subj", "condition", "score", "nrofconds", "nrofsubs")) # delete unnecessary variables
Test for assumption of normality
by(data = data.long$score, INDICES = data.long$condition, FUN = shapiro.test)
## data.long$condition: A
##
## Shapiro-Wilk normality test
##
## data: dd[x,]
## W = 0.9637, p-value = 0.6201
##
## -----
## data.long$condition: B
##
## Shapiro-Wilk normality test
##
## data: dd[x,]
## W = 0.92868, p-value = 0.1456
##
## -----
## data.long$condition: C
##
## Shapiro-Wilk normality test
## data: dd[x,]
## W = 0.94927, p-value = 0.3561
Test for equality of variances
leveneTest(y = data.long$score, group = data.long$condition)
## Levene's Test for Homogeneity of Variance (center = median)
   Df F value
                     Pr(>F)
## group 2 8.9065 0.0004306 ***
##
        57
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

#### ANOVA\* with the aov method

```
myModel = aov(formula = score ~ condition, data = data.long)
summary(myModel)
              Df Sum Sq Mean Sq F value
                                        Pr(>F)
## condition
              2 475.4 237.69
                                 24.88 1.71e-08 ***
## Residuals
            57 544.5
                          9.55
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
\mathbf{ANOVA}^* with the linear model method
myModel = lm(formula = score ~ condition, data = data.long)
summary(myModel)
##
## Call:
## lm(formula = score ~ condition, data = data.long)
##
## Residuals:
##
      Min
              1Q Median
                               ЗQ
## -7.9619 -1.5380 -0.1243 1.4997 7.6197
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 11.2260
                         0.6911 16.243 < 2e-16 ***
              6.8584
                           0.9774 7.017 2.98e-09 ***
## conditionB
## conditionC
              2.8167
                           0.9774 2.882 0.00557 **
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.091 on 57 degrees of freedom
## Multiple R-squared: 0.4661, Adjusted R-squared: 0.4474
## F-statistic: 24.88 on 2 and 57 DF, p-value: 1.708e-08
```

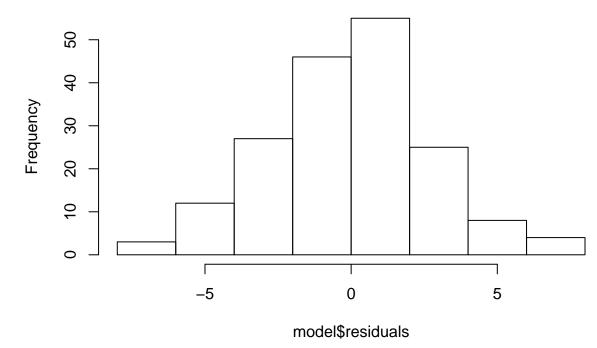
Table 6: data.long

subj	score	treatment	control
1	12.347533	A	control
2	15.659598	A	experimental
61	11.536699	В	control
62	8.740125	В	experimental
121	18.740275	С	control
122	18.493456	С	experimental

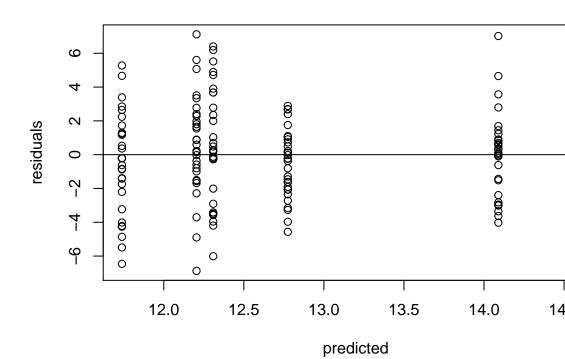
## 6 Factorial independent samples ANOVA

```
set.seed(01) # set seed
nrofcondsf1 = 3  # set number of conditions for factor 1
nrofcondsf2 = 2 # set number of conditions for factor 2
nrofsubs = nrofcondsf1*nrofcondsf2*30 # set number of subjects per condition
subj = as.factor(1:(nrofsubs))
                                    # create array with subject IDs
# create array witht treatment conditions
treatment = as.factor(rep(LETTERS[1:nrofcondsf1],each=nrofsubs/nrofcondsf1))
# create array with control / experimental
control = as.factor(rep(c("control","experimental"),times=nrofsubs/nrofcondsf2))
# create array with measurement values
score = as.vector( replicate(nrofcondsf1, replicate (
         nrofcondsf2 , rnorm(
           n = (nrofsubs/(nrofcondsf1*nrofcondsf2)),
           mean = 0 , sd = sample(5,1) ) + sample(8,1)+10
# combine arrays into a data.frame
data.long = data.frame(subj, score, treatment, control);
# delete unnecessary arrays
rm(list=c("control","nrofcondsf1","nrofcondsf2","nrofsubs","score","subj","treatment"))
Fit model with multiple prediction factors
model = lm(formula = score ~ treatment+control, data = data.long)
Test for assumption of normality
hist(model$residuals)
plot(x = fitted(model), y = residuals(model), xlab="predicted", ylab="residuals"); abline(h=0)
by(data = data.long$score, INDICES = paste(data.long$treatment, data.long$control), FUN = shapiro.test)
```

## Histogram of model\$residuals



 $Show\ residuals\ histogram$ 



```
Show residuals vs predicted plot
Show output of Shapiro-Wilk test
```

```
## paste(data.long$treatment, data.long$control): A control
##
## Shapiro-Wilk normality test
##
## data: dd[x,]
```

```
## W = 0.98372, p-value = 0.9135
##
## -----
## paste(data.long$treatment, data.long$control): A experimental
##
   Shapiro-Wilk normality test
##
##
## data: dd[x,]
## W = 0.97615, p-value = 0.7166
## -----
## paste(data.long$treatment, data.long$control): B control
##
##
   Shapiro-Wilk normality test
##
## data: dd[x,]
## W = 0.98224, p-value = 0.8815
## -----
## paste(data.long$treatment, data.long$control): B experimental
##
  Shapiro-Wilk normality test
##
## data: dd[x,]
## W = 0.95899, p-value = 0.2919
##
## -----
## paste(data.long$treatment, data.long$control): C control
##
##
  Shapiro-Wilk normality test
##
## data: dd[x,]
## W = 0.94015, p-value = 0.09179
##
## -----
## paste(data.long$treatment, data.long$control): C experimental
##
##
   Shapiro-Wilk normality test
##
## data: dd[x,]
## W = 0.93785, p-value = 0.07961
Test for equality of variances
leveneTest(y = data.long$score, group = as.factor(paste(data.long$treatment, data.long$control)))
## Levene's Test for Homogeneity of Variance (center = median)
      Df F value Pr(>F)
##
## group 5 1.7187 0.1327
##
       174
ANOVA with the linear model method
model = lm(formula = score ~ treatment+control, data = data.long)
library(car)
Anova(mod = model, type = 'II')
## Anova Table (Type II tests)
##
## Response: score
           Sum Sq Df F value
                              Pr(>F)
## treatment 185.93 2 12.2267 1.066e-05 ***
           14.63 1 1.9244
## control
                              0.1671
## Residuals 1338.23 176
```

```
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
summary(model)
##
## Call:
## lm(formula = score ~ treatment + control, data = data.long)
## Residuals:
            1Q Median
     Min
                              30
## -6.8762 -1.6438 0.1135 1.5856 7.1196
##
## Coefficients:
                     Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                     12.2045 0.4111 29.691 < 2e-16 ***
                      -0.4658
                                0.5034 -0.925 0.356064
## treatmentB
## treatmentC
                      1.8850
                                0.5034 3.744 0.000245 ***
## controlexperimental 0.5702
                                0.4111 1.387 0.167127
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
\#\# Residual standard error: 2.757 on 176 degrees of freedom
## Multiple R-squared: 0.1303, Adjusted R-squared: 0.1155
## F-statistic: 8.793 on 3 and 176 DF, p-value: 1.829e-05
```

```
set.seed(01) # set seed
nrofcondsf1 = 3  # set number of conditions for factor 1
nrofcondsf2 = 2 # set number of conditions for factor 2
nrofsubs = nrofcondsf1*nrofcondsf2*20 # set number of subjects per condition
subj = as.factor(1:(nrofsubs))
                               # create array with subject IDs
# create array witht treatment conditions
treatment = as.factor(rep(LETTERS[1:nrofcondsf1],each=nrofsubs/nrofcondsf1))
# create array with control / experimental
control = as.factor(rep(c("control","experimental"),times=nrofsubs/nrofcondsf2))
# create array with measurement values
score = as.vector( replicate(nrofcondsf1, replicate (
         {\tt nrofcondsf2} \ , \ {\tt rchisq}(
           n = (nrofsubs/(nrofcondsf1*nrofcondsf2)),
           df = 3)
       )+ sample(14,1)+10 ))
# combine arrays into a data.frame
data.long = data.frame(subj, score, treatment, control);
# delete unnecessary arrays
rm(list=c("control","nrofcondsf1","nrofcondsf2","nrofsubs","score","subj","treatment"))
```

Table 7: data.long

$\operatorname{subj}$	cond	score
1	A	13.36729
2	A	11.32874
1	В	17.22369
2	В	11.03194
1	С	18.11333
2	С	15.62249

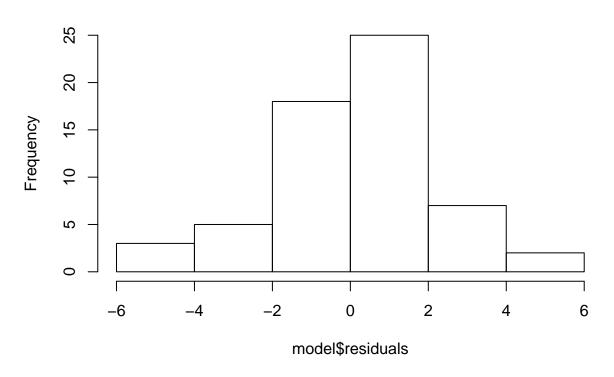
### 6.1 One-way repeated measures ANOVA

Show code for data generation

```
# Generate dataset
set.seed(01) # set seed
nrofsubs = 20 # set number of subjects
nrofconds = 3  # set number of conditions
subj = as.factor(rep(1:nrofsubs,nrofconds))
                                              # create array with subject IDs
cond = as.factor(rep(LETTERS[1:nrofconds], each=nrofsubs))  # create array with condition values
score = as.vector( replicate(
         nrofconds, rnorm(n = nrofsubs, mean = sample(8,1)+10, sd = sample(5,1))
                                               # create array with measurement values
data.long = data.frame(subj, cond, score); # combine arrays into a data.frame
rm(list=c("cond","nrofconds","nrofsubs","score","subj")) # delete arrays
# Generate dataset
set.seed(01) # set seed
nrofsubs = 20 # set number of subjects
data.wide = data.frame(
 subj = as.factor(1:nrofsubs)
 A = rnorm(n = nrofsubs, mean = sample(8,1)+10, sd = sample(5,1)),
B = rnorm(n = nrofsubs, mean = sample(8,1)+10, sd = sample(5,1)),
 C = rnorm(n = nrofsubs, mean = sample(8,1)+10, sd = sample(5,1))
rm(list=c("nrofsubs")) # delete arrays
fit linear model
model = lm(formula = cbind(data.wide$A, data.wide$B, data.wide$C)~1)
Test for assumption of normality
```

hist(model\$residuals)

## Histogram of model\$residuals



```
shapiro.test(model$residuals)
##
   Shapiro-Wilk normality test
##
## data: model$residuals
## W = 0.97837, p-value = 0.3637
Test for assumption of sphericity
mauchly.test(model, X=~1)
##
##
   Mauchly's test of sphericity
   Contrasts orthogonal to
   ~1
##
##
## data: SSD matrix from lm(formula = cbind(data.wide$A, data.wide$B, data.wide$C) ~ 1)
## W = 0.88731, p-value = 0.341
ANOVA
anova(model, X = ~1, test="Spherical")
## Analysis of Variance Table
##
##
## Contrasts orthogonal to
##
##
## Greenhouse-Geisser epsilon: 0.8987
## Huynh-Feldt epsilon:
                               0.9867
##
                       F num Df den Df
##
                                            Pr(>F)
                                                       G-G Pr
                                                                  H-F Pr
## (Intercept) 1 21.953
                              2
                                    38 4.6015e-07 1.4959e-06 5.3683e-07
```

## Residuals 19

Table 8: data.long

subj	score	treatment	control
_1	13.73940	A	control
2	19.35138	A	control
1	21.60730	A	experimental
2	20.96033	A	experimental
1	31.81861	В	control
2	24.01975	В	control
1	13.66997	В	experimental
2	14.21302	В	experimental
1	24.98184	С	control
2	20.38903	С	control
2	14.74345	С	experimental

### 6.2 Factorial repeated measures ANOVA

```
set.seed(02) # set seed
nrofcondsf1 = 3  # set number of conditions for factor 1
nrofcondsf2 = 2  # set number of conditions for factor 2
nrofsubs = 10 # set number of subjects
# create array with subject IDs
subj = as.factor(rep(1:(nrofsubs),times=nrofcondsf1*nrofcondsf2))
# create array witht treatment conditions
treatment = as.factor(rep(LETTERS[1:nrofcondsf1],each=nrofsubs*nrofcondsf2))
# create array with control / experimental
control = as.factor(
 rep(rep(c("control","experimental"),each=nrofsubs),times=nrofcondsf1))
# create array with measurement values
score = as.vector( replicate(nrofcondsf1,
                  replicate(nrofcondsf2,
                  rnorm(n = (nrofsubs), mean = sample(14,1)+10, sd = sample(5,1)
                                   )))))
# combine arrays into a data.frame
data.long = data.frame(subj, score, treatment, control);
# delete arrays
rm(list=c("control","nrofcondsf1","nrofcondsf2","nrofsubs", "score", "subj", "treatment"))
```