

DM8101FinalExam

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Quarto

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Running Code

When you click the **Render** button a document will be generated that includes both content and the output of embedded code. You can embed code like this:

```
1 + 1
```

```
[1] 2
```

You can add options to executable code like this

```
[1] 4
```

The `echo: false` option disables the printing of code (only output is displayed). ## excel相关操作 and tools

```
handle_error <- function(expr) {  
  result <- tryCatch(  
    {  
      expr  
    },  
    error = function(e) {  
      message("发生错误,但继续执行: ", conditionMessage(e))  
      return(NA)  
    }  
  )  
  return(result)  
}  
  
saveExcel <- function(data, sheetname, fileName=NULL){  
  library(openxlsx)  
  #library(xlsx)  
  if (!is.null(fileName)) {  
    excel_file <- fileName  
  }else{  
    excel_file <- "analysisResult.xlsx"  
  }  
  # 检查文件是否存在  
  if (!file.exists(excel_file)) {  
    # 如果文件不存在,则创建一个新的 Excel 工作簿  
    wb <- createWorkbook()  
  }
```

```

}else{
  wb <- loadWorkbook(excel_file)
}

name_str = names(wb)
index <- which(name_str == sheetname)[1]
if (!is.na(index)) {
  Rows = wb$worksheets[[index]]$sheet_data$rows
  Cols = wb$worksheets[[index]]$sheet_data$cols
  #print(length(Cols))
  #print(length(Rows))
  #print(Cols)
  #print(Rows)
  deleteData(wb, sheet=index, cols=1:max(Cols), rows=1:max(Rows), T)
  #for(i in 1:length(Rows)) {
  #  writeData(wb, index, "", startCol = Cols[i], startRow = Rows[i])
  #}
}else{
  print("~~~")
  handle_error( addWorksheet(wb, sheetname) )
  #handle_error( removeWorksheet(wb, sheetname) )
  print("~~~")
}
writeData(wb, sheetname, data)
saveWorkbook(wb, excel_file, overwrite = TRUE)
}

cat_ <-function(...){
  cat(..., sep = "")
}

getNumber <- function(num){
  num = format(round(num, digits = 3), nsmall = 3)
  return(num)
}

getCumulativeList <-function(list_src){
  newlist = list();
  for(i in 1:length(list_src)){
    total=as.numeric(list_src[i])
    if(i>1){
      total=total+newlist[[i-1]]
    }
    newlist <- append(newlist, total)
  }
  return(as.numeric(newlist) )
}

#因为loadings的数据结构太怪异,没有办法直接转matrix或data frames ,所以也保存不了excel,因此手工转换
transLoadings2Matrix<-function(loading){
  #loading[1:dataL,1:dataW ] <- ifelse(loading[1:dataL,1:dataW ] < 0.5, NA, loading[1:
  #summary(loading)
  #saveExcel(loading,"Factor Analysis loading")
  #str(loading)
  row_names <- c("",rownames(loading))
  col_names <- c("",colnames(loading))
  print(col_names)
}

```

```

tr_matrix <- matrix(nrow = length(row_names), ncol = length(col_names))
#rownames(tr_matrix) <- row_names
#colnames(tr_matrix) <- col_names
#tr_matrix[2,3] = 1 第二行,第三列
for (i in 2:length(row_names)) {
  tr_matrix[i,1] <- row_names[i]#loading[i, j]
  #loading[i, j]
  for (j in 2:length(col_names)) {
    #print(paste("Row:", i,row_names[i], ", Column:", j,col_names[j], ", Value:", lo
    if(loading[i-1, j-1]>0.5){
      tr_matrix[i,j] <- getNumber( loading[i-1, j-1] )
    }
  }
}
for (j in 2:length(col_names)) {
  tr_matrix[1,j] <- col_names[j]
}
#print(tr_matrix)
return(tr_matrix)
}

```

加载数据

```
library(memisc)
```

Loading required package: lattice

Loading required package: MASS

Attaching package: 'memisc'

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

```
contr.sum, contr.treatment, contrasts
```

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

```
as.array
```

```

#data0 = as.data.set(spss.system.file("spss/telework_new_office_12_srcdata.sav"))
data_src <- read.csv("spss/sh/data.csv")
dataL_src = nrow(data_src)
filtered_data <- data_src[data_src$totalseconds<30,]
cat("过滤掉:",nrow(filtered_data),"条填写时间少于30s的数据\n")

```

过滤掉: 13 条填写时间少于30s的数据

```

data_src = data_src[data_src$totalseconds>=30,]
#过滤选项全相同的,后面补上
cat("原始数据:",dataL_src,"条","有效数据:",nrow(data_src),"条\n")

```

原始数据: 106 条 有效数据: 93 条

```
rm(dataL_src,filtered_data)

data0 = data_src
#data = as.data.frame(data0)
#data = data[1:100,]
cols_base = c("age","gender","education","natrue_enterprise","type_work","job_title_s_
cols_independent =c("organisationalCulture2","organisationalCulture0","organisationalC

cols_independent_all=c("environment0","environment1","environment2","environment3","jo

cols_dependent = c("job_effectiveness1","job_effectiveness2","job_effectiveness3","job

#data <- lapply(data, as.numeric)
#data <- data0[,13:ncol(data0)]
##---
#data_independent <- data0[cols_independent]
#data_independent_all <- data0[cols_independent_all]
#data_dependent <- data0[cols_dependent]
#data_dependent_all <- data0[cols_dependent]
```

数据整理,分类

```
# Cronbach's α
#按每个变量单独测!!
#自变量
scaleName=c("Teleworker Characteristics","Communication","Management","Organisational
cols_independent_teleworkerCharacteristics=c("teleworkerCharacteristics0","teleworkerC
cols_independent_communication=c("communication0","communication1","communication2","c
cols_independent_management=c("management2","management3","management4")
cols_independent_organisationalCulture=c("organisationalCulture0","organisationalCultu
#因变量
cols_dependent_job_effectiveness = c("job_effectiveness1","job_effectiveness2","job_ef
cols_dependent_work.life_balance = c("work.life_balance1","work.life_balance2","work.l
cols_dependent_well.being = c("well.being1","well.being2","well.being3","well.being4")
#所有变量
all_var = list( cols_independent_teleworkerCharacteristics, cols_independent_communica
#对每个变量求均值
data0$teleworkerCharacteristics = rowMeans(data0[cols_independent_teleworkerCharacteri
data0$communication = rowMeans(data0[cols_independent_communication])
data0$management = rowMeans(data0[cols_independent_management])
data0$organisationalCulture = rowMeans(data0[cols_independent_organisationalCulture])
data0$job_effectiveness = rowMeans(data0[cols_dependent_job_effectiveness])
data0$work.life_balance = rowMeans(data0[cols_dependent_work.life_balance])
data0$well.being = rowMeans(data0[cols_dependent_well.being])

all_var_calculated = c("teleworkerCharacteristics","communication","management","organ
#str(data0)
```

描述性分析

```
library(psych)
library(knitr)
options(cat.sep = "")
data_t <- data0[cols_base]
desc_data <- describe(data_t)
psych::describe(data_t)
```

	vars	n	mean	sd	median	trimmed	mad	min	max	range	skew
age	1	93	1.97	1.10	2	1.97	1.48	0	4	4	0.01
gender	2	93	0.43	0.50	0	0.41	0.00	0	1	1	0.28
education	3	93	1.69	0.88	2	1.61	1.48	0	4	4	0.64
natrue_enterprise	4	93	0.86	0.69	1	0.83	0.00	0	2	2	0.18
type_work	5	93	1.46	0.90	1	1.44	0.00	0	3	3	0.55
job_title_s_data	6	93	0.12	0.44	0	0.00	0.00	0	3	3	4.40
work_experience	7	93	1.59	0.78	1	1.53	1.48	0	3	3	0.44
marital_status	8	93	1.44	1.13	2	1.37	1.48	0	4	4	0.17
partners_work	9	93	0.86	1.06	0	0.76	0.00	0	3	3	0.55
telework_is_active	10	93	0.19	0.40	0	0.12	0.00	0	1	1	1.53
hours_per_day	11	93	2.28	0.73	2	2.37	1.48	0	3	3	-0.81
days_per_week	12	93	2.23	0.78	2	2.32	1.48	0	3	3	-0.81
	kurtosis		se								
age			-0.67	0.11							
gender			-1.94	0.05							
education			-0.16	0.09							
natrue_enterprise			-0.91	0.07							
type_work			-0.73	0.09							
job_title_s_data			21.74	0.05							
work_experience			-0.69	0.08							
marital_status			-0.74	0.12							
partners_work			-1.40	0.11							
telework_is_active			0.33	0.04							
hours_per_day			0.41	0.08							
days_per_week			0.23	0.08							

```
freq_tables <- lapply(data_t, table)
print(freq_tables)
```

\$age

0	1	2	3	4
9	22	33	21	8

\$gender

0	1
53	40

\$education

0	1	2	3	4
---	---	---	---	---

3 43 30 14 3

\$natrue_enterprise

0 1 2
29 48 16

\$type_work

0 1 2 3
8 52 15 18

\$job_title_s_data

0 1 2 3
85 6 1 1

\$work_experience

0 1 2 3
3 46 30 14

\$marital_status

0 1 2 3 4
27 14 40 8 4

\$partners_work

0 1 2 3
54 3 31 5

\$telework_is_active

0 1
75 18

\$hours_per_day

0 1 2 3
2 9 43 39

\$days_per_week

0 1 2 3
3 11 41 38

```
describe_item=c("Age","Gender","Education","Nature of corporate ownership","work type")
describe_str=list(
  c("~25","25~30","30~40","40~45","45~"), #age
  c("男","女"), #gender
  c("Junior", "college", "Undergraduate Postgraduate", "Doctor", "Other"), #education
  c("Foreign-funded enterprise", "Private enterprise", "State-owned enterprise"), #nat
  c("management", "R&D", "Salse", "Operation and maintenance"), #type work
```

```

c("Ordinary staff", "Grass-roots management", "Middle management", "Top management")
c("1", "2~5", "5~10", "10~"), #work experience
c("unmarried", "Married without children", "Married with children", "Divorced and childless")
c("unemployed", "In the same company", "Different companies in the same city", "in different cities")
c("active", "passive"), #telework is active
c("1~2", "2~4", "4~8", "8~"), # hours per day
c("1", "2~3", "3~5", "5~") #days per week
)
content = list("Classification percentage");
itemName = list("Item name");
for (i in 1:length(freq_tables)) {
  #cat_(names(data_t)[i], ": ")
  cat_(describe_item[i], ": ")
  itemName <- append(itemName, describe_item[i])
  result_str=""
  for(e in freq_tables[i]){
    #print(names(e))
    #输出均值,标准差,
    #输出频数与非分比
    total = sum(e)
    for(name in names(e)){
      num = e[name]
      freq = num/total*100
      freq = getNumber(freq)#format(round(freq, digits = 3), nsmall = 3)
      namestr = describe_str[[i]][as.integer(name)+1]
      result_str = paste(result_str, namestr, ":", e[name], " ", freq, "% \n")
      cat_(namestr, ":", e[name], " ", freq, "%, ")
    }
    content <- append(content, result_str)
    cat_("\n")
  }
}

```

Age: ~25:9 9.677%, 25~30:22 23.656%, 30~40:33 35.484%, 40~45:21 22.581%, 45~:8 8.602%,
Gender: 男:53 56.989%, 女:40 43.011%,
Education: Junior:3 3.226%, college:43 46.237%, Undergraduate Postgraduate:30 32.258%,
Doctor:14 15.054%, Other:3 3.226%,
Nature of corporate ownership: Foreign-funded enterprise:29 31.183%, Private enterprise:48 51.613%, State-owned enterprise:16 17.204%,
work type: management:8 8.602%, R&D:52 55.914%, Sales:15 16.129%, Operation and maintenance:18 19.355%,
Job title: Ordinary staff:85 91.398%, Grass-roots management:6 6.452%, Middle management:1 1.075%, Top management:1 1.075%,
work experience: 1:3 3.226%, 2~5:46 49.462%, 5~10:30 32.258%, 10~:14 15.054%,
marital status: unmarried:27 29.032%, Married without children:14 15.054%, Married with children:40 43.011%, Divorced and childless:8 8.602%, Divorced with children:4 4.301%,
partners work: unemployed:54 58.065%, In the same company:3 3.226%, Different companies in the same city:31 33.333%, in different cities:5 5.376%,
telework is active: active:75 80.645%, passive:18 19.355%,
working hours pre day: 1~2:2 2.151%, 2~4:9 9.677%, 4~8:43 46.237%, 8~:39 41.935%,
working days pre week: 1:3 3.226%, 2~3:11 11.828%, 3~5:41 44.086%, 5~:38 40.860%,

```
# 使用cbind函数将两个列表按列组合成一个数据集
#dataset0 <- data.frame(itemName, content)
dataset <- cbind(itemName, content)
# 转换结果为数据框
dataset <- as.data.frame(dataset)
#dataset
saveExcel(dataset,"descriptive statistics")
rm(content,data_t,desc_data,describe_str,freq_tables,itemName,dataset,e,describe_item,
```

##三大检验### 独立检验 检验两个分类变量之间是否存在关联

```
data <- table(data0$age, data0$gender)
chisq.test(data)
```

Warning in chisq.test(data): Chi-squared approximation may be incorrect

Pearson's Chi-squared test

data: data
X-squared = 1.684, df = 4, p-value = 0.7936

正态性检验

```
library(nortest)
data <- data0[,13]
result = shapiro.test(data) # 对应group 每组水平下的检验
print(result$p.value)
```

[1] 5.841864e-10

方差齐性

方差齐性特指两个或两个以上总体方差是否具有显著差异的特性

```
bartlett.test(data0$environment0~data0$gender, data = data0)
```

Bartlett test of homogeneity of variances

data: data0\$environment0 by data0\$gender
Bartlett's K-squared = 0.0062887, df = 1, p-value = 0.9368

信度分析 Cronbach's α (克隆巴哈系数)

Internal Reliability • 再测信度 • Cronbach's α (克隆巴哈系数) • 折半信度 • Guttman • 平行模型检验 • 严密平行模型检验 • 库李20信度 Inter-rater Reliability • Kappa系数 • 组内相关系数ICC

```
#print(all_var)
library(psych)
#names(all_var)
```



```

get_var_name <- function(x) {
  deparse(substitute(x))
}
tr_matrix <- matrix(nrow = length(all_var)+sum(lengths(all_var))+1, ncol = 3)
tr_matrix[1,] = c("", "Cronbach's alpha", "N of Items")
#- |Cronbach's alpha|N of Items
#name|0.829          |5
tr_matrixrow=2
for(i in seq_len(length(all_var))){
  selected_column_name = unlist(all_var[i])
  data_t <- data0[ selected_column_name ]
  alpha_result <- psych::alpha(data_t, check.keys=TRUE)
  tr_matrix[tr_matrixrow,1] = scaleName[i]
  tr_matrix[tr_matrixrow,2] = getNumber( alpha_result$total$raw_alpha )
  tr_matrix[tr_matrixrow,3] = length(alpha_result$keys[[1]])
  tr_matrixrow=tr_matrixrow+1
  for(j in 1:length(alpha_result$keys[[1]])) {
    tr_matrix[tr_matrixrow,1] = alpha_result$keys[[1]][j]
    tr_matrix[tr_matrixrow,2] = getNumber( alpha_result$item.stats$raw.r[j] )
    tr_matrixrow=tr_matrixrow+1
  }
}

```

Warning in psych::alpha(data_t, check.keys = TRUE): Some items were negatively correlated with the first principal component and were automatically reversed. This is indicated by a negative sign for the variable name.

```

saveExcel(tr_matrix, "Cronbach's alpha")
print(tr_matrix)

```

	[,1]	[,2]	[,3]
[1,]	"	"Cronbach's alpha"	"N of Items"
[2,]	"Teleworker Characteristics"	"0.889"	"4"
[3,]	"teleworkerCharacteristics0"	"0.903"	NA
[4,]	"teleworkerCharacteristics1"	"0.828"	NA
[5,]	"teleworkerCharacteristics2"	"0.834"	NA
[6,]	"teleworkerCharacteristics3"	"0.903"	NA
[7,]	"Communication"	"0.724"	"4"
[8,]	"-communication0"	"0.364"	NA
[9,]	"communication1"	"0.854"	NA
[10,]	"communication2"	"0.899"	NA
[11,]	"communication3"	"0.868"	NA
[12,]	"Management"	"0.899"	"3"
[13,]	"management2"	"0.902"	NA
[14,]	"management3"	"0.960"	NA
[15,]	"management4"	"0.873"	NA
[16,]	"Organisational Culture"	"0.946"	"4"
[17,]	"organisationalCulture0"	"0.948"	NA
[18,]	"organisationalCulture1"	"0.908"	NA
[19,]	"organisationalCulture2"	"0.938"	NA
[20,]	"organisationalCulture3"	"0.924"	NA
[21,]	"Job Effectiveness"	"0.962"	"4"
[22,]	"job_effectiveness1"	"0.953"	NA

[23,]	"job_effectiveness2"	"0.943"	NA
[24,]	"job_effectiveness3"	"0.961"	NA
[25,]	"job_effectiveness4"	"0.936"	NA
[26,]	"work-life balance"	"0.966"	"4"
[27,]	"work.life_balance1"	"0.950"	NA
[28,]	"work.life_balance2"	"0.967"	NA
[29,]	"work.life_balance3"	"0.959"	NA
[30,]	"work.life_balance4"	"0.941"	NA
[31,]	"well-being"	"0.951"	"4"
[32,]	"well.being1"	"0.953"	NA
[33,]	"well.being2"	"0.950"	NA
[34,]	"well.being3"	"0.959"	NA
[35,]	"well.being4"	"0.904"	NA

KMO检测 and bartlett's test

```
#data_independent <- data0[cols_independent]
#data_independent_all <- data0[cols_independent_all]
#data_dependent <- data0[cols_dependent]
#data_dependent_all <- data0[cols_dependent]
data = data0[c(cols_independent,cols_dependent)]
item = list()
content = list()
kmo <- KMO(data)
bartlett <- bartlett.test(data)

item <- append(item, "Overall MSA" )
content <- append(content, getNumber(kmo$MSA) )

item <- append(item, names(bartlett$statistic) )
content <- append(content, getNumber(bartlett$statistic) )
item <- append(item, "df" )
content <- append(content, bartlett$parameter )
item <- append(item, "Sig." )
content <- append( content, getNumber(bartlett$p.value) )

item <- append(item, names(kmo$MSAi) )
content <- append(content, getNumber(kmo$MSAi) )

#dataset0 <- data.frame(itemName, content)
dataset <- cbind(item, content)
# 转换结果为数据框
dataset <- as.data.frame(dataset)
#dataset
saveExcel(dataset, "KMO and Bartlett's Test")

print(kmo)
```

Kaiser-Meyer-Olkin factor adequacy

Call: KMO(r = data)

Overall MSA = 0.91

MSA for each item =

organisationalCulture2	organisationalCulture0
0.80	0.82
organisationalCulture3	organisationalCulture1
0.88	0.91
management2	management3
0.94	0.89
management4	teleworkerCharacteristics1
0.86	0.79
teleworkerCharacteristics2	teleworkerCharacteristics3
0.88	0.92
teleworkerCharacteristics0	communication3
0.85	0.85
communication2	communication1
0.88	0.87
job_effectiveness1	job_effectiveness2
0.92	0.89
job_effectiveness3	job_effectiveness4
0.93	0.97
work.life_balance1	work.life_balance2
0.91	0.97
work.life_balance3	work.life_balance4
0.94	0.91
well.being1	well.being2
0.91	0.98
well.being3	well.being4
0.97	0.92

```
print(bartlett)
```

Bartlett test of homogeneity of variances

data: data

Bartlett's K-squared = 256.41, df = 25, p-value < 2.2e-16

```
rm(item,content,kmo,bartlett,dataset,data)
```

因子分析 EFA:

Factor Analysis Total Variance Explained & communalities

```
library(psych)
library(paran)
#library(rgl) #需要opengl
library(ggfortify)
```

Loading required package: ggplot2

Attaching package: 'ggplot2'

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

%, alpha

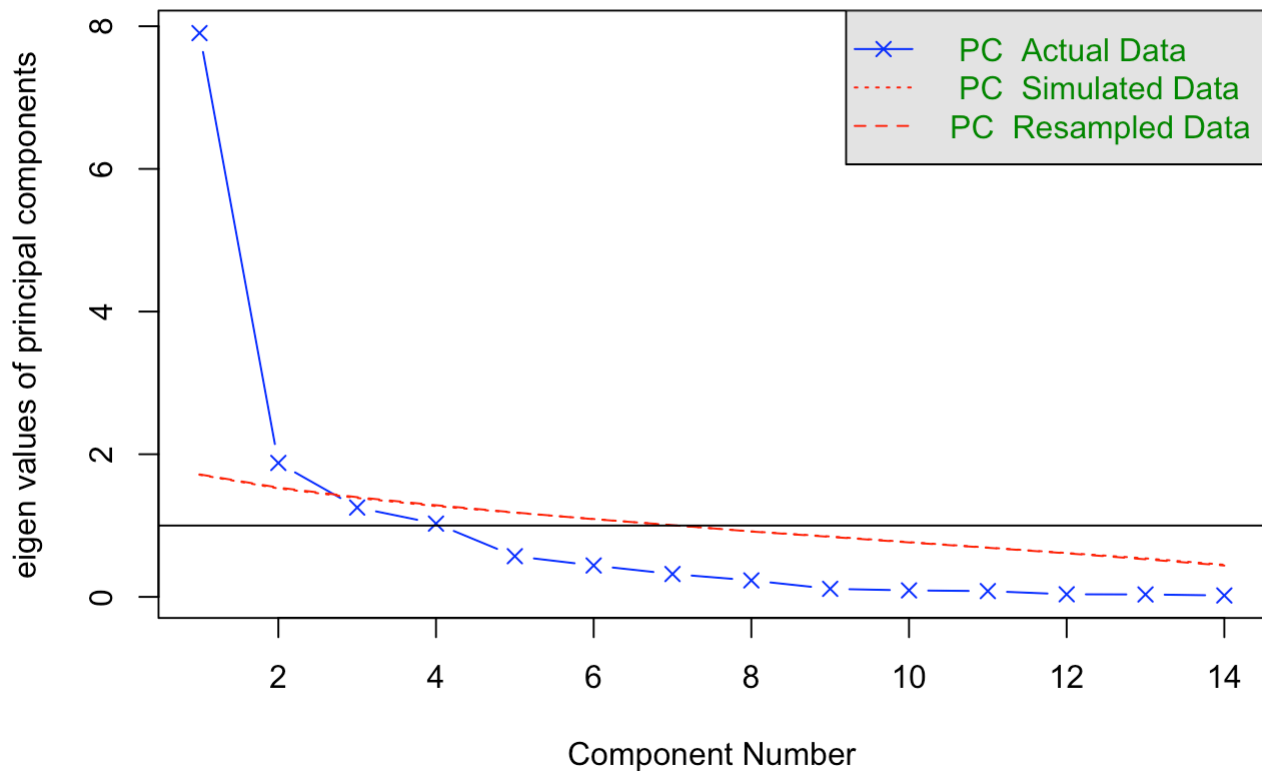
The following object is masked from 'package:memisc':

syms

```
data <- data0[cols_independent]
#data <- data0[cols_dependent]

# AIC BIC 比较模型 parallel平行分析 可得因子数,否则只能自己手动尝试观察
parallel_result <- fa.parallel(data, fa = "pc", n.iter = 100, main = "Parallel Analysis")
```

Parallel Analysis



Parallel analysis suggests that the number of factors = NA and the number of components = 2

```
factors = parallel_result$nfact
if( is.na(parallel_result$nfact) ){
  factors = parallel_result$ncomp
  if( is.na(parallel_result$ncomp) ){
    stop("没有得到因子数,请检查数据") # 抛出异常
    quit(save = "no", status = 1) # 退出当前会话
  }
}
print(factors)
```

[1] 2

```
rm(parallel_result)
## 因子分析
fa_results <- fa(data, nfactors = factors, rotate = "varimax")
#得到Communalities表 Initial都为1 , Extraction就是下面的值
communalities = fa_results$communalities
component = names(communalities)
Initial = rep(1,length(component))
Extraction = getNumber( as.numeric(abs(communalities)) )
dataset <- cbind(component, Initial, Extraction)
dataset <- as.data.frame(dataset)
saveExcel(dataset,"Factor Analysis communalities")
#得到Total Variance Explained
dataL = length(fa_results$values)
print(dataL)
```

[1] 14

```
component <- 1:dataL
#得到Initial Eigenvalues

variance_explained = abs(fa_results$values) / sum(abs(fa_results$values)) *100
Total = getNumber( abs(fa_results$values)) # Total
Variance = getNumber( variance_explained) # % of Variance
Cumulative = getNumber( getCumulativeList(variance_explained)) # Cumulative %
print(Total)
```

```
[1] "7.580" "1.508" "0.873" "0.604" "0.224" "0.077" "0.000" "0.088" "0.147"
[10] "0.234" "0.268" "0.316" "0.331" "0.395"
```

```
print(Variance)
```

```
[1] "59.938" "11.928" " 6.903" " 4.780" " 1.775" " 0.606" " 0.003" " 0.697"
[9] " 1.161" " 1.849" " 2.122" " 2.497" " 2.617" " 3.125"
```

```
print(Cumulative)
```

```
[1] " 59.938" " 71.866" " 78.769" " 83.549" " 85.324" " 85.930" " 85.933"
[8] " 86.630" " 87.791" " 89.640" " 91.762" " 94.258" " 96.875" "100.000"
```

```
#得到 Extraction Sums of Squared Loadings 与Initial Eigenvalues一样,只是只显示因子项
print(class(Total))
```

[1] "character"

```
Total1 = Total
Variance1 = Variance
Cumulative1 = Cumulative
startIndex = factors+1
endIndex = dataL
```

```
Total1[startIndex:endIndex] <- NA # Total
Variance1[startIndex:endIndex] <- NA # % of Variance
Cumulative1[startIndex:endIndex] <- NA # Cumulative %
#得到 Rotation Sums of Squared Loadings
variance_explained = abs(fa_results$e.values) / sum(abs(fa_results$e.values)) *100
Total2 = getNumber( abs(fa_results$e.values) ) # Total
Variance2 = getNumber( variance_explained) # % of Variance
Cumulative2 = getNumber( getCumulativeList(variance_explained)) # Cumulative %
print(Total2)
```

```
[1] "7.904" "1.878" "1.253" "1.027" "0.570" "0.440" "0.320" "0.231" "0.113"
[10] "0.091" "0.081" "0.037" "0.034" "0.020"
```

```
print(Variance2)
```

```
[1] "56.457" "13.412" " 8.948" " 7.337" " 4.074" " 3.145" " 2.289" " 1.651"
[9] " 0.806" " 0.650" " 0.581" " 0.261" " 0.244" " 0.142"
```

```
print(Cumulative2)
```

```
[1] " 56.457" " 69.869" " 78.817" " 86.155" " 90.229" " 93.374" " 95.663"
[8] " 97.314" " 98.121" " 98.771" " 99.352" " 99.613" " 99.858" "100.000"
```

```
Total2[startIndex:endIndex] <- NA # Total
Variance2[startIndex:endIndex] <- NA # % of Variance
Cumulative2[startIndex:endIndex] <- NA # Cumulative %
dataset <- cbind(component, "Total"=Total, "% of Variance"=Variance,"Cumulative %"=Cum
dataset <- as.data.frame(dataset)
saveExcel(dataset,"Factor Analysis Total")

##获得旋转矩阵 loadings
#summary(fa_results)
print(fa_results$loadings)
```

Loadings:

	MR1	MR2
organisationalCulture2	0.770	0.186
organisationalCulture0	0.786	0.165
organisationalCulture3	0.765	0.392
organisationalCulture1	0.813	0.439
management2	0.457	0.642
management3	0.516	0.688
management4	0.632	0.349
teleworkerCharacteristics1		0.715
teleworkerCharacteristics2	0.203	0.696
teleworkerCharacteristics3	0.302	0.854
teleworkerCharacteristics0	0.365	0.823
communication3	0.654	0.170
communication2	0.743	0.261
communication1	0.746	0.194

	MR1	MR2
SS loadings	5.136	3.953
Proportion Var	0.367	0.282
Cumulative Var	0.367	0.649

```
tr_matrix = transLoadings2Matrix(fa_results$loadings)
```

```
[1] "" "MR1" "MR2"
```

```
#saveExcel(tr_matrix,"Factor Analysis loading")
print(fa_results$Vaccounted)
```

	MR1	MR2
SS loadings	5.1357240	3.9525900
Proportion Var	0.3668374	0.2823279
Cumulative Var	0.3668374	0.6491653
Proportion Explained	0.5650910	0.4349090
Cumulative Proportion	0.5650910	1.0000000

```
tr_matrix1 = transLoadings2Matrix(fa_results$Vaccounted)
```

```
[1] "" "MR1" "MR2"
```

```
tr_matrix = rbind(tr_matrix,tr_matrix1)
#print(tr_matrix)
saveExcel(tr_matrix,"Factor Analysis loading")
## EFA 分析,探索性分析
#fa = factanal(data,factors=factors,rotation = "varimax")
#summary(fa)
#autoplot3d(fa, color = "Factor")
#scatter3d(fa$scores[, 1], fa$scores[, 2], fa$scores[, 3], color = fa$loadings[, 1])
#print(fa)

rm(data,dataset,factors,fa_results,fa,loadings,filtered_loadings,tr_matrix,tr_matrix1,
```

Warning in rm(data, dataset, factors, fa_results, fa, loadings, filtered_loadings, : object 'fa' not found

Warning in rm(data, dataset, factors, fa_results, fa, loadings, filtered_loadings, : object 'loadings' not found

Warning in rm(data, dataset, factors, fa_results, fa, loadings, filtered_loadings, : object 'filtered_loadings' not found

Warning in rm(data, dataset, factors, fa_results, fa, loadings, filtered_loadings, : object 'Cumulative3' not found

T-Test

单样本T检验

- 默认前提条件是数据需要符合正态分布性
- 结果是否显著等于某一值
- 男性的工资显著等于3000元

```
# 创建一个数值向量
data <- c(1, 2, 3, 4, 5, 6, 7, 8, 9)
# 执行单样本t检验, 检验均值是否显著不同于0
t_test_result <- t.test(data, mu = 0)
# 打印结果
print(t_test_result)
```

One Sample t-test

```
data: data
t = 5.4772, df = 8, p-value = 0.0005894
alternative hypothesis: true mean is not equal to 0
95 percent confidence interval:
 2.894916 7.105084
sample estimates:
mean of x
      5
```

独立样本T检验

- 要求因变量(y)需要符合正态分布性
- X与Y的差异是否显著
- 例:研究男性工资与女性工资之间的差异

```
# 创建两个数值向量
#group1 <- c(1, 2, 3, 4, 5)
#group2 <- c(6, 7, 8, 9, 10)
# 执行独立样本t检验
#t_test_result <- t.test(group1, group2)
# 打印结果
#print(t_test_result)
t_test_result <- t.test(data0[data0$gender==0,]$job_effectiveness, data0[data0$gender=
print(t_test_result)
```

Welch Two Sample t-test

```
data: data0[data0$gender == 0,]$job_effectiveness and data0[data0$gender == 1,
]$job_effectiveness
t = 1.4103, df = 71.08, p-value = 0.1628
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
 -0.1059034  0.6176958
sample estimates:
mean of x mean of y
 4.193396  3.937500
```

```
t_test_result <- t.test(data0[data0$gender==0,]$work.life_balance, data0[data0$gender=
print(t_test_result)
```


Welch Two Sample t-test

```
data: data0[data0$gender == 0, ]$work.life_balance and data0[data0$gender == 1,
]$work.life_balance
t = 0.46999, df = 76.724, p-value = 0.6397
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
 -0.2565232  0.4150138
sample estimates:
mean of x mean of y
 3.929245  3.850000
```

```
t_test_result <- t.test(data0[data0$gender==0,]$well.being, data0[data0$gender==1,]$we
print(t_test_result)
```

Welch Two Sample t-test

```
data: data0[data0$gender == 0, ]$well.being and data0[data0$gender == 1, ]$well.being
t = 0.72014, df = 73.148, p-value = 0.4737
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
 -0.2601080  0.5544476
sample estimates:
mean of x mean of y
 4.04717  3.90000
```

配对样本T检验

• 默认前提条件是差值数据需要符合正态分布性 • 利用来自两个总体的配对样本，推断两个总体的均值是否存在显著差异。 • 办公室提供免费咖啡和没有提供免费咖啡的两组员工，生产力是否一样？

```
# 创建配对的数值向量
before <- c(1, 2, 3, 4, 5)
after <- c(2, 3, 4, 5, 6)
# 执行配对样本t检验
#t_test_result <- t.test(before, after, paired = TRUE)
# 打印结果
#print(t_test_result)
```

方差分析

ONE-WAY ANOVA 单因素方差分析 方差齐检验homogeneity of variance test

ANOVA, LevenTest, Tukey's HSD, Duncan's C

temp

```
library(car)
```

Loading required package: carData

Attaching package: 'car'

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

logit

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

recode

```
library(multcomp)
```

Loading required package: mvtnorm

Loading required package: survival

Loading required package: TH.data

Attaching package: 'TH.data'

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

geyser

```
library(multcompView)
## 按性别分析因变量
anova<-function(x_data,y_data,tr_matrix_old=NULL){

  #加一行title
  tr_matrix <- matrix(nrow = 2, ncol = 7)
  xname = paste(as.character(substitute(y_data)),collapse="")
  yname = paste(as.character(substitute(x_data)),collapse="")
  #str(xname)
  tempstr = paste("diff " ,xname , " by " ,yname )
  print(tempstr)
  tr_matrix[2,1] = tempstr
  if(!is.null(tr_matrix_old)){
    tr_matrix = rbind(tr_matrix_old, tr_matrix)
  }

  x <- factor( as.character(x_data) )
  y <- y_data
  data = data.frame(x,y)
  model <- aov(y ~ x, data=data)
  leveneTest_result = leveneTest(model)
  # 进行图基事后比较
  tukey_comparison <- glht(model, linfct = mcp(x = "Tukey"))
  # 进行邓尼特事后比较, 其中"control"是对照组
  dunnett_comparison <- glht(model, linfct = mcp(x = "Dunnett"))
```

```

print("model")
sum = summary(model) #spss中的ANOVA全在这个结果里面
#Df (自由度)：这列显示了每个方差来源的自由度。
#Sum Sq (平方和)：这列显示了每个方差来源的平方和。平方和(组间(值小),组内(值大),总计(前两相加))
#Mean Sq (均方)：这列显示了每个方差来源的均方(即平方和除以自由度)。平方和/自由度
#F value (F统计量)：这是组间均方与组内均方的比值。
#Pr(>F) (p值)：这是与F统计量相关联的p值。
tr_matrix_old = tr_matrix
tr_matrix <- matrix(nrow = 4, ncol = 7)
tr_matrix[1,]=c("", "平方和", "自由度", "均方", "F", "显著性", NA)
tr_matrix[,1]=c("", "组间", "组内", "总计")
tr_matrix[2,2] = getNumber(sum[[1]]["Sum Sq"][1,1])
tr_matrix[3,2] = getNumber(sum[[1]]["Sum Sq"][2,1])
tr_matrix[4,2] = getNumber(sum(sum[[1]]["Sum Sq"]))
tr_matrix[2,3] = sum[[1]]["Df"][1,1]
tr_matrix[3,3] = sum[[1]]["Df"][2,1]
tr_matrix[4,3] = sum(sum[[1]]["Df"])
tr_matrix[2,4] = getNumber(sum[[1]]["Mean Sq"][1,1])
tr_matrix[3,4] = getNumber(sum[[1]]["Mean Sq"][2,1])
tr_matrix[2,5] = getNumber(sum[[1]]["F value"][1,1])
tr_matrix[2,6] = getNumber(sum[[1]]["Pr(>F)"][1,1])
print(tr_matrix)
tr_matrix = rbind(tr_matrix_old, tr_matrix)

print("leveneTest_result")
#print(leveneTest_result) #spss中的方差齐性检验
#F value 莱文统计
#df 自由度1,2
#Pr(>F) (p值)：这是与F统计量相关联的p值
tr_matrix_old = tr_matrix
tr_matrix <- matrix(nrow = 2, ncol = 7)
tr_matrix[1,]=c("莱文统计", "自由度1", "自由度2", "显著性", NA, NA, NA)
tr_matrix[2,1] = getNumber(leveneTest_result["F value"][1,1])
tr_matrix[2,2] = leveneTest_result["Df"][1,1]
tr_matrix[2,3] = leveneTest_result["Df"][2,1]
tr_matrix[2,4] = getNumber(leveneTest_result["Pr(>F)"][1,1])
print(tr_matrix)
tr_matrix = rbind(tr_matrix_old, tr_matrix)

print("tukey_comparison")
sum = summary(tukey_comparison) #图基HSD
str(sum$test)
print(sum)
diffname = names(sum$test$coefficients)
t_critical <- qt(0.975, 348)
tr_matrix_old = tr_matrix
tr_matrix <- matrix(nrow = length(diffname)+2, ncol = 7)
tr_matrix[1,]=c("", "", "", "", "", "95%置信区间", "")
tr_matrix[2,]=c("group", "group", "平均值差值", "标准差", "显著性", "下限", "上限")
lastname = ""
for(i in 1:length(diffname)){
  names = split_str <- strsplit(diffname[i], " - ")
  if(lastname!=names[[1]][2]){

```

```

    lastname = tr_matrix[2+i,1] = names[[1]][2]
  }
  tr_matrix[2+i,2] = names[[1]][1]
  c=""
  if( sum$test$pvalues[i]<0.001){
    c="***"
  }else if( sum$test$pvalues[i]<0.01){
    c="**"
  }else if( sum$test$pvalues[i]<0.05){
    c="*"
  }
  tr_matrix[2+i,3] = paste( getNumber( sum$test$coefficients[i] ), c,collapse="")
  tr_matrix[2+i,4] = getNumber( sum$test$sigma[i] )
  tr_matrix[2+i,5] = getNumber( sum$test$pvalues[i] )
  tr_matrix[2+i,6] = getNumber( sum$test$coefficients[i] - t_critical * sum$test$sig
  tr_matrix[2+i,7] = getNumber( sum$test$coefficients[i] + t_critical * sum$test$sig
}
print(tr_matrix)
tr_matrix = rbind(tr_matrix_old, tr_matrix)
#print(tr_matrix)
##multcomp_plot(tukey_comparison)
#print("dunnett_comparison")
#print(summary(dunnett_comparison))
return(tr_matrix)
}
#anova(data0$gender,data0$job_effectiveness)
dataset = anova(data0$age,data0$job_effectiveness,tr_matrix_old=NULL)

```

```
[1] "diff $data0job_effectiveness by $data0age"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"6.720"	"4"	"1.680"	"2.535"	"0.046"	NA
[3,]	"组内"	"58.322"	"88"	"0.663"	NA	NA	NA
[4,]	"总计"	"65.042"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.407"	"4"	"88"	"0.803"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```

$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:10] 0.527 0.902 0.786 0.521 0.375 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ sigma : Named num [1:10] 0.322 0.306 0.324 0.396 0.224 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ tstat : Named num [1:10] 1.63 2.94 2.42 1.32 1.67 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ pvalues : num [1:10] 0.4694 0.0312 0.114 0.6728 0.445 ...
..- attr(*, "error")= num 0.000376
$ type : chr "single-step"
- attr(*, "class")= chr "mtest"

```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.526515	0.322123	1.635	0.4694
2 - 0 == 0	0.901515	0.306140	2.945	0.0312 *
3 - 0 == 0	0.785714	0.324342	2.422	0.1140
4 - 0 == 0	0.520833	0.395578	1.317	0.6728
2 - 1 == 0	0.375000	0.224072	1.674	0.4450
3 - 1 == 0	0.259199	0.248363	1.044	0.8285
4 - 1 == 0	-0.005682	0.336107	-0.017	1.0000
3 - 2 == 0	-0.115801	0.227250	-0.510	0.9856
4 - 2 == 0	-0.380682	0.320822	-1.187	0.7514
4 - 3 == 0	-0.264881	0.338235	-0.783	0.9322

```
----
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.527 "	"0.322"	"0.469"	"-0.107"	"1.160"
[4,]	NA	"2"	"0.902 *"	"0.306"	"0.031"	"0.299"	"1.504"
[5,]	NA	"3"	"0.786 "	"0.324"	"0.114"	"0.148"	"1.424"
[6,]	NA	"4"	"0.521 "	"0.396"	"0.673"	"-0.257"	"1.299"
[7,]	"1"	"2"	"0.375 "	"0.224"	"0.445"	"-0.066"	"0.816"
[8,]	NA	"3"	"0.259 "	"0.248"	"0.829"	"-0.229"	"0.748"
[9,]	NA	"4"	"-0.006 "	"0.336"	"1.000"	"-0.667"	"0.655"
[10,]	"2"	"3"	"-0.116 "	"0.227"	"0.986"	"-0.563"	"0.331"
[11,]	NA	"4"	"-0.381 "	"0.321"	"0.751"	"-1.012"	"0.250"
[12,]	"3"	"4"	"-0.265 "	"0.338"	"0.932"	"-0.930"	"0.400"

```
dataset = anova(data0$age,data0$work.life_balance,tr_matrix_old=dataset)
```

```
[1] "diff $data0work.life_balance by $data0age"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"4.199"	"4"	"1.050"	"1.761"	"0.144"	NA
[3,]	"组内"	"52.466"	"88"	"0.596"	NA	NA	NA
[4,]	"总计"	"56.665"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.174"	"4"	"88"	"0.951"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```

$ qfunction      :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:10] 0.367 0.697 0.524 0.302 0.33 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ sigma         : Named num [1:10] 0.306 0.29 0.308 0.375 0.213 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ tstat         : Named num [1:10] 1.203 2.4 1.703 0.805 1.551 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ pvalues       : num [1:10] 0.742 0.12 0.427 0.925 0.523 ...
..- attr(*, "error")= num 0.000338
$ type          : chr "single-step"
- attr(*, "class")= chr "mtest"

```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

Fit: aov(formula = y ~ x, data = data)

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.36742	0.30552	1.203	0.742
2 - 0 == 0	0.69697	0.29037	2.400	0.120
3 - 0 == 0	0.52381	0.30763	1.703	0.427
4 - 0 == 0	0.30208	0.37519	0.805	0.925
2 - 1 == 0	0.32955	0.21253	1.551	0.523
3 - 1 == 0	0.15639	0.23557	0.664	0.962
4 - 1 == 0	-0.06534	0.31879	-0.205	1.000
3 - 2 == 0	-0.17316	0.21554	-0.803	0.926
4 - 2 == 0	-0.39489	0.30429	-1.298	0.685
4 - 3 == 0	-0.22173	0.32081	-0.691	0.956

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间" ""	
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.367 "	"0.306"	"0.742"	"-0.233"	"0.968"
[4,]	NA	"2"	"0.697 "	"0.290"	"0.120"	"0.126"	"1.268"
[5,]	NA	"3"	"0.524 "	"0.308"	"0.427"	"-0.081"	"1.129"
[6,]	NA	"4"	"0.302 "	"0.375"	"0.925"	"-0.436"	"1.040"
[7,]	"1"	"2"	"0.330 "	"0.213"	"0.523"	"-0.088"	"0.748"
[8,]	NA	"3"	"0.156 "	"0.236"	"0.962"	"-0.307"	"0.620"
[9,]	NA	"4"	"-0.065 "	"0.319"	"1.000"	"-0.692"	"0.562"
[10,]	"2"	"3"	"-0.173 "	"0.216"	"0.926"	"-0.597"	"0.251"
[11,]	NA	"4"	"-0.395 "	"0.304"	"0.685"	"-0.993"	"0.204"
[12,]	"3"	"4"	"-0.222 "	"0.321"	"0.956"	"-0.853"	"0.409"

```
dataset = anova(data0$age,data0$well.being,tr_matrix_old=dataset)
```

```

[1] "diff $data0well.being by $data0age"
[1] "model"
      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
[1,] ""    "平方和" "自由度" "均方" "F"    "显著性" NA

```

```
[2,] "组间" "5.877" "4" "1.469" "1.699" "0.157" NA
[3,] "组内" "76.099" "88" "0.865" NA NA NA
[4,] "总计" "81.976" "92" NA NA NA NA
[1] "leveneTest_result"
      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
[1,] "莱文统计" "自由度1" "自由度2" "显著性" NA NA NA
[2,] "0.500" "4" "88" "0.736" NA NA NA
[1] "tukey_comparison"
```

List of 7

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:10] 0.549 0.841 0.476 0.365 0.292 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ sigma : Named num [1:10] 0.368 0.35 0.37 0.452 0.256 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ tstat : Named num [1:10] 1.493 2.405 1.285 0.807 1.14 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ pvalues : num [1:10] 0.56 0.119 0.692 0.925 0.778 ...
..- attr(*, "error")= num 0.00019
$ type : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

Fit: aov(formula = y ~ x, data = data)

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.54924	0.36796	1.493	0.560
2 - 0 == 0	0.84091	0.34970	2.405	0.119
3 - 0 == 0	0.47619	0.37049	1.285	0.692
4 - 0 == 0	0.36458	0.45186	0.807	0.925
2 - 1 == 0	0.29167	0.25595	1.140	0.778
3 - 1 == 0	-0.07305	0.28370	-0.257	0.999
4 - 1 == 0	-0.18466	0.38393	-0.481	0.988
3 - 2 == 0	-0.36472	0.25958	-1.405	0.617
4 - 2 == 0	-0.47633	0.36647	-1.300	0.683
4 - 3 == 0	-0.11161	0.38636	-0.289	0.998

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.549 "	"0.368"	"0.560"	"-0.174"	"1.273"
[4,]	NA	"2"	"0.841 "	"0.350"	"0.119"	"0.153"	"1.529"
[5,]	NA	"3"	"0.476 "	"0.370"	"0.692"	"-0.252"	"1.205"
[6,]	NA	"4"	"0.365 "	"0.452"	"0.925"	"-0.524"	"1.253"
[7,]	"1"	"2"	"0.292 "	"0.256"	"0.778"	"-0.212"	"0.795"
[8,]	NA	"3"	"-0.073 "	"0.284"	"0.999"	"-0.631"	"0.485"
[9,]	NA	"4"	"-0.185 "	"0.384"	"0.988"	"-0.940"	"0.570"
[10,]	"2"	"3"	"-0.365 "	"0.260"	"0.617"	"-0.875"	"0.146"

```
[11,] NA      "4"      "-0.476 "    "0.366"    "0.683"    "-1.197"    "0.244"
[12,] "3"      "4"      "-0.112 "    "0.386"    "0.998"    "-0.872"    "0.648"
```

```
dataset = anova(data0$gender,data0$job_effectiveness,tr_matrix_old=dataset)
```

```
[1] "diff $data0job_effectiveness by $data0gender"
```

```
[1] "model"
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
```

```
[1,] ""          "平方和"  "自由度"  "均方"    "F"          "显著性"  NA
```

```
[2,] "组间"      "1.493"    "1"        "1.493"    "2.138"    "0.147"    NA
```

```
[3,] "组内"      "63.549"   "91"        "0.698"    NA          NA          NA
```

```
[4,] "总计"      "65.042"   "92"        NA          NA          NA          NA
```

```
[1] "leveneTest_result"
```

```
      [,1]      [,2]      [,3]      [,4]      [,5] [,6] [,7]
```

```
[1,] "莱文统计"  "自由度1"  "自由度2"  "显著性"  NA     NA   NA
```

```
[2,] "1.956"     "1"        "91"        "0.165"   NA     NA   NA
```

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```
$ qfunction :function (conf.level, adjusted = TRUE, ...)
```

```
$ coefficients: Named num -0.256
```

```
..- attr(*, "names")= chr "1 - 0"
```

```
$ sigma : Named num 0.175
```

```
..- attr(*, "names")= chr "1 - 0"
```

```
$ tstat : Named num -1.46
```

```
..- attr(*, "names")= chr "1 - 0"
```

```
$ pvalues : num 0.147
```

```
..- attr(*, "error")= num 0
```

```
$ type : chr "single-step"
```

```
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

```
Linear Hypotheses:
```

```
Estimate Std. Error t value Pr(>|t|)
```

```
1 - 0 == 0 -0.2559 0.1750 -1.462 0.147
```

```
(Adjusted p values reported -- single-step method)
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
```

```
[1,] ""          ""          ""          ""          ""          "95%置信区间" ""
```

```
[2,] "group"      "group"  "平均值差值"  "标准差"  "显著性"  "下限"          "上限"
```

```
[3,] "0"          "1"        "-0.256 "    "0.175"    "0.147"    "-0.600"    "0.088"
```

```
dataset = anova(data0$gender,data0$work.life_balance,tr_matrix_old=dataset)
```

```
[1] "diff $data0work.life_balance by $data0gender"
```

```
[1] "model"
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
```



```
[1,] ""          "平方和" "自由度" "均方" "F"          "显著性" NA
[2,] "组间" "0.143" "1"          "0.143" "0.230" "0.632" NA
[3,] "组内" "56.522" "91"         "0.621" NA      NA      NA
[4,] "总计" "56.665" "92"         NA      NA      NA      NA
[1] "leveneTest_result"
      [,1]      [,2]      [,3]      [,4]      [,5] [,6] [,7]
[1,] "莱文统计" "自由度1" "自由度2" "显著性" NA    NA    NA
[2,] "0.441"     "1"        "91"        "0.509" NA    NA    NA
[1] "tukey_comparison"
```

List of 7

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num -0.0792
..- attr(*, "names")= chr "1 - 0"
$ sigma      : Named num 0.165
..- attr(*, "names")= chr "1 - 0"
$ tstat       : Named num -0.48
..- attr(*, "names")= chr "1 - 0"
$ pvalues      : num 0.632
..- attr(*, "error")= num 0
$ type         : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

```
          Estimate Std. Error t value Pr(>|t|)
1 - 0 == 0 -0.07925    0.16507   -0.48    0.632
(Adjusted p values reported -- single-step method)
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
[1,] ""          ""          ""          ""          ""          "95%置信区间" ""
[2,] "group" "group" "平均值差值" "标准差" "显著性" "下限"      "上限"
[3,] "0"      "1"      "-0.079" "0.165" "0.632" "-0.404" "0.245"
```

```
dataset = anova(data0$gender,data0$well.being,tr_matrix_old=dataset)
```

```
[1] "diff $data0well.being by $data0gender"
[1] "model"
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
[1,] ""          "平方和" "自由度" "均方" "F"          "显著性" NA
[2,] "组间" "0.494" "1"          "0.494" "0.551" "0.460" NA
[3,] "组内" "81.482" "91"         "0.895" NA      NA      NA
[4,] "总计" "81.976" "92"         NA      NA      NA      NA
[1] "leveneTest_result"
      [,1]      [,2]      [,3]      [,4]      [,5] [,6] [,7]
[1,] "莱文统计" "自由度1" "自由度2" "显著性" NA    NA    NA
[2,] "1.092"     "1"        "91"        "0.299" NA    NA    NA
```

```
[1] "tukey_comparison"
List of 7
 $ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
 $ qfunction :function (conf.level, adjusted = TRUE, ...)
 $ coefficients: Named num -0.147
   ..- attr(*, "names")= chr "1 - 0"
 $ sigma : Named num 0.198
   ..- attr(*, "names")= chr "1 - 0"
 $ tstat : Named num -0.743
   ..- attr(*, "names")= chr "1 - 0"
 $ pvalues : num 0.46
   ..- attr(*, "error")= num 0
 $ type : chr "single-step"
 - attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	-0.1472	0.1982	-0.743	0.46

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"-0.147"	"0.198"	"0.460"	"-0.537"	"0.243"

```
dataset = anova(data0$education,data0$job_effectiveness,tr_matrix_old=dataset)
```

```
[1] "diff $data0job_effectiveness by $data0education"
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"27.240"	"4"	"6.810"	"15.853"	"0.000"	NA
[3,]	"组内"	"37.802"	"88"	"0.430"	NA	NA	NA
[4,]	"总计"	"65.042"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.516"	"4"	"88"	"0.724"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
 $ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
 $ qfunction :function (conf.level, adjusted = TRUE, ...)
 $ coefficients: Named num [1:10] 0.891 1.742 1.405 -0.583 0.85 ...
   ..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
 $ sigma : Named num [1:10] 0.391 0.397 0.417 0.535 0.156 ...
   ..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ tstat      : Named num [1:10] 2.28 4.39 3.37 -1.09 5.45 ...
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
$ pvalues    : num [1:10] 1.44e-01 2.21e-04 8.11e-03 7.91e-01 3.16e-06 ...
..- attr(*, "error")= num 0.00037
$ type       : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.8915	0.3914	2.278	0.14355
2 - 0 == 0	1.7417	0.3969	4.388	< 0.001 ***
3 - 0 == 0	1.4048	0.4170	3.369	0.00811 **
4 - 0 == 0	-0.5833	0.5351	-1.090	0.79141
2 - 1 == 0	0.8502	0.1559	5.453	< 0.001 ***
3 - 1 == 0	0.5133	0.2017	2.545	0.07848 .
4 - 1 == 0	-1.4748	0.3914	-3.768	0.00229 **
3 - 2 == 0	-0.3369	0.2121	-1.588	0.47757
4 - 2 == 0	-2.3250	0.3969	-5.858	< 0.001 ***
4 - 3 == 0	-1.9881	0.4170	-4.768	< 0.001 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.891 "	"0.391"	"0.144"	"0.122"	"1.661"
[4,]	NA	"2"	"1.742 ***"	"0.397"	"0.000"	"0.961"	"2.522"
[5,]	NA	"3"	"1.405 **"	"0.417"	"0.008"	"0.585"	"2.225"
[6,]	NA	"4"	"-0.583 "	"0.535"	"0.791"	"-1.636"	"0.469"
[7,]	"1"	"2"	"0.850 ***"	"0.156"	"0.000"	"0.544"	"1.157"
[8,]	NA	"3"	"0.513 "	"0.202"	"0.078"	"0.117"	"0.910"
[9,]	NA	"4"	"-1.475 **"	"0.391"	"0.002"	"-2.245"	"-0.705"
[10,]	"2"	"3"	"-0.337 "	"0.212"	"0.478"	"-0.754"	"0.080"
[11,]	NA	"4"	"-2.325 ***"	"0.397"	"0.000"	"-3.106"	"-1.544"
[12,]	"3"	"4"	"-1.988 ***"	"0.417"	"0.000"	"-2.808"	"-1.168"

```
dataset = anova(data0$education,data0$work.life_balance,tr_matrix_old=dataset)
```

```
[1] "diff $data0work.life_balance by $data0education"
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"9.109"	"4"	"2.277"	"4.214"	"0.004"	NA
[3,]	"组内"	"47.557"	"88"	"0.540"	NA	NA	NA
[4,]	"总计"	"56.665"	"92"	NA	NA	NA	NA

```

[1] "leveneTest_result"
      [,1]      [,2]      [,3]      [,4]      [,5] [,6] [,7]
[1,] "莱文统计" "自由度1" "自由度2" "显著性" NA   NA   NA
[2,] "0.149"    "4"      "88"      "0.963" NA   NA   NA
[1] "tukey_comparison"
List of 7
 $ pfunction      :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
 $ qfunction      :function (conf.level, adjusted = TRUE, ...)
 $ coefficients: Named num [1:10] 8.62e-01 1.26 1.15 1.76e-15 3.96e-01 ...
 ..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
 $ sigma         : Named num [1:10] 0.439 0.445 0.468 0.6 0.175 ...
 ..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
 $ tstat         : Named num [1:10] 1.96 2.83 2.47 2.93e-15 2.26 ...
 ..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
 $ pvalues       : num [1:10] 0.2651 0.0385 0.0939 1 0.1477 ...
 ..- attr(*, "error")= num 0.000702
 $ type          : chr "single-step"
 - attr(*, "class")= chr "mtest"

```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

Fit: aov(formula = y ~ x, data = data)

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	8.624e-01	4.390e-01	1.965	0.2651
2 - 0 == 0	1.258e+00	4.451e-01	2.827	0.0385 *
3 - 0 == 0	1.155e+00	4.677e-01	2.469	0.0939 .
4 - 0 == 0	1.756e-15	6.002e-01	0.000	1.0000
2 - 1 == 0	3.959e-01	1.749e-01	2.264	0.1477
3 - 1 == 0	2.924e-01	2.262e-01	1.292	0.6700
4 - 1 == 0	-8.624e-01	4.390e-01	-1.965	0.2652
3 - 2 == 0	-1.036e-01	2.379e-01	-0.435	0.9913
4 - 2 == 0	-1.258e+00	4.451e-01	-2.827	0.0386 *
4 - 3 == 0	-1.155e+00	4.677e-01	-2.469	0.0941 .

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.862 "	"0.439"	"0.265"	"-0.001"	"1.726"
[4,]	NA	"2"	"1.258 *"	"0.445"	"0.039"	"0.383"	"2.134"
[5,]	NA	"3"	"1.155 "	"0.468"	"0.094"	"0.235"	"2.075"
[6,]	NA	"4"	"0.000 "	"0.600"	"1.000"	"-1.181"	"1.181"
[7,]	"1"	"2"	"0.396 "	"0.175"	"0.148"	"0.052"	"0.740"
[8,]	NA	"3"	"0.292 "	"0.226"	"0.670"	"-0.153"	"0.737"
[9,]	NA	"4"	"-0.862 "	"0.439"	"0.265"	"-1.726"	"0.001"
[10,]	"2"	"3"	"-0.104 "	"0.238"	"0.991"	"-0.572"	"0.364"

```
[11,] NA      "4"      "-1.258 *"    "0.445"    "0.039"    "-2.134"    "-0.383"
[12,] "3"      "4"      "-1.155 "    "0.468"    "0.094"    "-2.075"    "-0.235"
```

```
dataset = anova(data0$education,data0$well.being,tr_matrix_old=dataset)
```

```
[1] "diff $data0well.being by $data0education"
```

```
[1] "model"
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
```

```
[1,] ""          "平方和"  "自由度"  "均方"    "F"        "显著性" NA
```

```
[2,] "组间"      "13.246"  "4"       "3.312"   "4.240"    "0.003" NA
```

```
[3,] "组内"      "68.730"  "88"      "0.781"   NA         NA      NA
```

```
[4,] "总计"      "81.976"  "92"      NA        NA         NA      NA
```

```
[1] "leveneTest_result"
```

```
      [,1]      [,2]      [,3]      [,4]      [,5] [,6] [,7]
```

```
[1,] "莱文统计"  "自由度1"  "自由度2"  "显著性" NA   NA   NA
```

```
[2,] "0.668"     "4"        "88"       "0.616" NA   NA   NA
```

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```
$ qfunction :function (conf.level, adjusted = TRUE, ...)
```

```
$ coefficients: Named num [1:10] 1.0097 1.5583 1.1845 0.0833 0.5486 ...
```

```
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ sigma : Named num [1:10] 0.528 0.535 0.562 0.722 0.21 ...
```

```
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ tstat : Named num [1:10] 1.913 2.912 2.107 0.115 2.61 ...
```

```
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ pvalues : num [1:10] 0.2901 0.0307 0.2034 1 0.0672 ...
```

```
..- attr(*, "error")= num 0.000345
```

```
$ type : chr "single-step"
```

```
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

```
Linear Hypotheses:
```

```
      Estimate Std. Error t value Pr(>|t|)
1 - 0 == 0  1.00969    0.52773   1.913   0.2901
2 - 0 == 0  1.55833    0.53514   2.912   0.0307 *
3 - 0 == 0  1.18452    0.56225   2.107   0.2034
4 - 0 == 0  0.08333    0.72158   0.115   1.0000
2 - 1 == 0  0.54864    0.21023   2.610   0.0672 .
3 - 1 == 0  0.17483    0.27194   0.643   0.9630
4 - 1 == 0 -0.92636    0.52773  -1.755   0.3757
3 - 2 == 0 -0.37381    0.28604  -1.307   0.6606
4 - 2 == 0 -1.47500    0.53514  -2.756   0.0463 *
4 - 3 == 0 -1.10119    0.56225  -1.959   0.2680
```

```
----
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
(Adjusted p values reported -- single-step method)
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"1.010 "	"0.528"	"0.290"	"-0.028"	"2.048"
[4,]	NA	"2"	"1.558 *"	"0.535"	"0.031"	"0.506"	"2.611"
[5,]	NA	"3"	"1.185 "	"0.562"	"0.203"	"0.079"	"2.290"
[6,]	NA	"4"	"0.083 "	"0.722"	"1.000"	"-1.336"	"1.503"
[7,]	"1"	"2"	"0.549 "	"0.210"	"0.067"	"0.135"	"0.962"
[8,]	NA	"3"	"0.175 "	"0.272"	"0.963"	"-0.360"	"0.710"
[9,]	NA	"4"	"-0.926 "	"0.528"	"0.376"	"-1.964"	"0.112"
[10,]	"2"	"3"	"-0.374 "	"0.286"	"0.661"	"-0.936"	"0.189"
[11,]	NA	"4"	"-1.475 *"	"0.535"	"0.046"	"-2.528"	"-0.422"
[12,]	"3"	"4"	"-1.101 "	"0.562"	"0.268"	"-2.207"	"0.005"

```
dataset = anova(data0$natrue_enterprise,data0$job_effectiveness,tr_matrix_old=dataset)
```

```
[1] "diff $data0job_effectiveness by $data0natrue_enterprise"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"28.248"	"2"	"14.124"	"34.548"	"0.000"	NA
[3,]	"组内"	"36.794"	"90"	"0.409"	NA	NA	NA
[4,]	"总计"	"65.042"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.437"	"2"	"90"	"0.647"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:3] -0.989 -1.51 -0.521
..- attr(*, "names")= chr [1:3] "1 - 0" "2 - 0" "2 - 1"
$ sigma : Named num [1:3] 0.15 0.199 0.185
..- attr(*, "names")= chr [1:3] "1 - 0" "2 - 0" "2 - 1"
$ tstat : Named num [1:3] -6.58 -7.58 -2.82
..- attr(*, "names")= chr [1:3] "1 - 0" "2 - 0" "2 - 1"
$ pvalues : num [1:3] 4.08e-09 4.17e-11 1.59e-02
..- attr(*, "error")= num 0.000406
$ type : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	-0.9889	0.1504	-6.576	<0.001 ***

```
2 - 0 == 0 -1.5097      0.1991 -7.582 <0.001 ***
2 - 1 == 0 -0.5208      0.1846 -2.822 0.0159 *
```

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"-0.989 ***"	"0.150"	"0.000"	"-1.285"	"-0.693"
[4,]	NA	"2"	"-1.510 ***"	"0.199"	"0.000"	"-1.901"	"-1.118"
[5,]	"1"	"2"	"-0.521 *"	"0.185"	"0.016"	"-0.884"	"-0.158"

```
dataset = anova(data0$natrue_enterprise,data0$work.life_balance,tr_matrix_old=dataset)
```

```
[1] "diff $data0work.life_balance by $data0natrue_enterprise"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"36.428"	"2"	"18.214"	"81.005"	"0.000"	NA
[3,]	"组内"	"20.237"	"90"	"0.225"	NA	NA	NA
[4,]	"总计"	"56.665"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"1.958"	"2"	"90"	"0.147"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:3] -1.253 -1.56 -0.307
..- attr(*, "names")= chr [1:3] "1 - 0" "2 - 0" "2 - 1"
$ sigma : Named num [1:3] 0.112 0.148 0.137
..- attr(*, "names")= chr [1:3] "1 - 0" "2 - 0" "2 - 1"
$ tstat : Named num [1:3] -11.24 -10.57 -2.24
..- attr(*, "names")= chr [1:3] "1 - 0" "2 - 0" "2 - 1"
$ pvalues : num [1:3] 0 0 0.068
..- attr(*, "error")= num 0.000197
$ type : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	-1.2531	0.1115	-11.236	<0.001 ***
2 - 0 == 0	-1.5603	0.1477	-10.566	<0.001 ***
2 - 1 == 0	-0.3073	0.1369	-2.245	0.068 .

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"-1.253 ***"	"0.112"	"0.000"	"-1.472"	"-1.034"
[4,]	NA	"2"	"-1.560 ***"	"0.148"	"0.000"	"-1.851"	"-1.270"
[5,]	"1"	"2"	"-0.307 "	"0.137"	"0.068"	"-0.577"	"-0.038"

```
dataset = anova(data0$natrue_enterprise,data0$well.being,tr_matrix_old=dataset)
```

```
[1] "diff $data0well.being by $data0natrue_enterprise"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
--	------	------	------	------	------	------	------

[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
------	----	-------	-------	------	-----	-------	----

[2,]	"组间"	"49.969"	"2"	"24.984"	"70.253"	"0.000"	NA
------	------	----------	-----	----------	----------	---------	----

[3,]	"组内"	"32.007"	"90"	"0.356"	NA	NA	NA
------	------	----------	------	---------	----	----	----

[4,]	"总计"	"81.976"	"92"	NA	NA	NA	NA
------	------	----------	------	----	----	----	----

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
--	------	------	------	------	------	------	------

[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
------	--------	--------	--------	-------	----	----	----

[2,]	"2.746"	"2"	"90"	"0.070"	NA	NA	NA
------	---------	-----	------	---------	----	----	----

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```
$ qfunction :function (conf.level, adjusted = TRUE, ...)
```

```
$ coefficients: Named num [1:3] -1.248 -2.061 -0.813
```

```
..- attr(*, "names")= chr [1:3] "1 - 0" "2 - 0" "2 - 1"
```

```
$ sigma : Named num [1:3] 0.14 0.186 0.172
```

```
..- attr(*, "names")= chr [1:3] "1 - 0" "2 - 0" "2 - 1"
```

```
$ tstat : Named num [1:3] -8.9 -11.1 -4.72
```

```
..- attr(*, "names")= chr [1:3] "1 - 0" "2 - 0" "2 - 1"
```

```
$ pvalues : num [1:3] 4.75e-14 0.00 2.11e-05
```

```
..- attr(*, "error")= num 2.36e-06
```

```
$ type : chr "single-step"
```

```
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	-1.2484	0.1403	-8.901	< 1e-05 ***
2 - 0 == 0	-2.0609	0.1857	-11.097	< 1e-05 ***
2 - 1 == 0	-0.8125	0.1722	-4.720	2.11e-05 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"-1.248 ***"	"0.140"	"0.000"	"-1.524"	"-0.973"
[4,]	NA	"2"	"-2.061 ***"	"0.186"	"0.000"	"-2.426"	"-1.696"
[5,]	"1"	"2"	"-0.813 ***"	"0.172"	"0.000"	"-1.151"	"-0.474"

```
dataset = anova(data0$type_work,data0$job_effectiveness,tr_matrix_old=dataset)
```

```
[1] "diff $data0job_effectiveness by $data0type_work"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
--	------	------	------	------	------	------	------

[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
------	----	-------	-------	------	-----	-------	----

[2,]	"组间"	"7.168"	"3"	"2.389"	"3.674"	"0.015"	NA
------	------	---------	-----	---------	---------	---------	----

[3,]	"组内"	"57.874"	"89"	"0.650"	NA	NA	NA
------	------	----------	------	---------	----	----	----

[4,]	"总计"	"65.042"	"92"	NA	NA	NA	NA
------	------	----------	------	----	----	----	----

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
--	------	------	------	------	------	------	------

[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
------	--------	--------	--------	-------	----	----	----

[2,]	"0.369"	"3"	"89"	"0.776"	NA	NA	NA
------	---------	-----	------	---------	----	----	----

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```
$ qfunction :function (conf.level, adjusted = TRUE, ...)
```

```
$ coefficients: Named num [1:6] 0.452 -0.075 -0.167 -0.527 -0.619 ...
```

```
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```

```
$ sigma : Named num [1:6] 0.306 0.353 0.343 0.236 0.221 ...
```

```
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```

```
$ tstat : Named num [1:6] 1.476 -0.212 -0.486 -2.23 -2.805 ...
```

```
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```

```
$ pvalues : num [1:6] 0.4458 0.9964 0.9603 0.1179 0.0295 ...
```

```
..- attr(*, "error")= num 0.000342
```

```
$ type : chr "single-step"
```

```
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.45192	0.30625	1.476	0.4458
2 - 0 == 0	-0.07500	0.35304	-0.212	0.9964
3 - 0 == 0	-0.16667	0.34265	-0.486	0.9603
2 - 1 == 0	-0.52692	0.23634	-2.230	0.1179
3 - 1 == 0	-0.61859	0.22052	-2.805	0.0295 *
3 - 2 == 0	-0.09167	0.28192	-0.325	0.9875

```
----
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.452 "	"0.306"	"0.446"	"-0.150"	"1.054"
[4,]	NA	"2"	"-0.075 "	"0.353"	"0.996"	"-0.769"	"0.619"
[5,]	NA	"3"	"-0.167 "	"0.343"	"0.960"	"-0.841"	"0.507"
[6,]	"1"	"2"	"-0.527 "	"0.236"	"0.118"	"-0.992"	"-0.062"
[7,]	NA	"3"	"-0.619 *"	"0.221"	"0.030"	"-1.052"	"-0.185"
[8,]	"2"	"3"	"-0.092 "	"0.282"	"0.988"	"-0.646"	"0.463"

```
dataset = anova(data0$type_work,data0$work.life_balance,tr_matrix_old=dataset)
```

```
[1] "diff $data0work.life_balance by $data0type_work"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"8.859"	"3"	"2.953"	"5.498"	"0.002"	NA
[3,]	"组内"	"47.806"	"89"	"0.537"	NA	NA	NA
[4,]	"总计"	"56.665"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.620"	"3"	"89"	"0.604"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:6] 0.748 0.36 0.066 -0.387 -0.682 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ sigma : Named num [1:6] 0.278 0.321 0.311 0.215 0.2 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ tstat : Named num [1:6] 2.686 1.123 0.212 -1.803 -3.401 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ pvalues : num [1:6] 0.03987 0.66707 0.99646 0.27003 0.00505 ...
..- attr(*, "error")= num 0.000453
$ type : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.74760	0.27834	2.686	0.03987 *
2 - 0 == 0	0.36042	0.32086	1.123	0.66707
3 - 0 == 0	0.06597	0.31142	0.212	0.99646

```

2 - 1 == 0 -0.38718    0.21480 -1.803  0.27003
3 - 1 == 0 -0.68162    0.20043 -3.401  0.00505 **
3 - 2 == 0 -0.29444    0.25622 -1.149  0.65083

```

```
---
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.748 *"	"0.278"	"0.040"	"0.200"	"1.295"
[4,]	NA	"2"	"0.360 "	"0.321"	"0.667"	"-0.271"	"0.991"
[5,]	NA	"3"	"0.066 "	"0.311"	"0.996"	"-0.547"	"0.678"
[6,]	"1"	"2"	"-0.387 "	"0.215"	"0.270"	"-0.810"	"0.035"
[7,]	NA	"3"	"-0.682 **"	"0.200"	"0.005"	"-1.076"	"-0.287"
[8,]	"2"	"3"	"-0.294 "	"0.256"	"0.651"	"-0.798"	"0.209"

```
dataset = anova(data0$type_work,data0$well.being,tr_matrix_old=dataset)
```

```

[1] "diff $data0well.being by $data0type_work"
[1] "model"
      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
[1,] ""      "平方和" "自由度" "均方" "F"      "显著性" NA
[2,] "组间" "13.471" "3"      "4.490" "5.834" "0.001" NA
[3,] "组内" "68.505" "89"     "0.770" NA      NA      NA
[4,] "总计" "81.976" "92"     NA      NA      NA      NA
[1] "leveneTest_result"
      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
[1,] "莱文统计" "自由度1" "自由度2" "显著性" NA      NA      NA
[2,] "0.211"     "3"      "89"     "0.888" NA      NA      NA
[1] "tukey_comparison"
List of 7
 $ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
 $ qfunction :function (conf.level, adjusted = TRUE, ...)
 $ coefficients: Named num [1:6] 0.558 -0.05 -0.361 -0.608 -0.919 ...
 ..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
 $ sigma      : Named num [1:6] 0.333 0.384 0.373 0.257 0.24 ...
 ..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
 $ tstat       : Named num [1:6] 1.674 -0.13 -0.969 -2.363 -3.83 ...
 ..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
 $ pvalues     : num [1:6] 0.33372 0.99917 0.76009 0.08796 0.00112 ...
 ..- attr(*, "error")= num 0.000596
 $ type        : chr "single-step"
 - attr(*, "class")= chr "mtest"

```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.5577	0.3332	1.674	0.33372
2 - 0 == 0	-0.0500	0.3841	-0.130	0.99917
3 - 0 == 0	-0.3611	0.3728	-0.969	0.76009
2 - 1 == 0	-0.6077	0.2571	-2.363	0.08796 .
3 - 1 == 0	-0.9188	0.2399	-3.830	0.00112 **
3 - 2 == 0	-0.3111	0.3067	-1.014	0.73348

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.558 "	"0.333"	"0.334"	"-0.098"	"1.213"
[4,]	NA	"2"	"-0.050 "	"0.384"	"0.999"	"-0.805"	"0.705"
[5,]	NA	"3"	"-0.361 "	"0.373"	"0.760"	"-1.094"	"0.372"
[6,]	"1"	"2"	"-0.608 "	"0.257"	"0.088"	"-1.113"	"-0.102"
[7,]	NA	"3"	"-0.919 **"	"0.240"	"0.001"	"-1.391"	"-0.447"
[8,]	"2"	"3"	"-0.311 "	"0.307"	"0.733"	"-0.914"	"0.292"

```
dataset = anova(data0$job_title_s_data,data0$job_effectiveness,tr_matrix_old=dataset)
```

[1] "diff \$data0job_effectiveness by \$data0job_title_s_data"

[1] "model"

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"0.547"	"3"	"0.182"	"0.251"	"0.860"	NA
[3,]	"组内"	"64.495"	"89"	"0.725"	NA	NA	NA
[4,]	"总计"	"65.042"	"92"	NA	NA	NA	NA

[1] "leveneTest_result"

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"1.154"	"3"	"89"	"0.332"	NA	NA	NA

[1] "tukey_comparison"

List of 7

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:6] -0.1863 -0.6029 -0.1029 -0.4167 0.0833 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ sigma : Named num [1:6] 0.36 0.856 0.856 0.919 0.919 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ tstat : Named num [1:6] -0.518 -0.7042 -0.1202 -0.4532 0.0906 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ pvalues : num [1:6] 0.949 0.883 0.999 0.965 1 ...
..- attr(*, "error")= num 1.7e-05
$ type : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	-0.18627	0.35959	-0.518	0.949
2 - 0 == 0	-0.60294	0.85626	-0.704	0.883
3 - 0 == 0	-0.10294	0.85626	-0.120	0.999
2 - 1 == 0	-0.41667	0.91948	-0.453	0.965
3 - 1 == 0	0.08333	0.91948	0.091	1.000
3 - 2 == 0	0.50000	1.20388	0.415	0.972

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"-0.186 "	"0.360"	"0.949"	"-0.894"	"0.521"
[4,]	NA	"2"	"-0.603 "	"0.856"	"0.883"	"-2.287"	"1.081"
[5,]	NA	"3"	"-0.103 "	"0.856"	"0.999"	"-1.787"	"1.581"
[6,]	"1"	"2"	"-0.417 "	"0.919"	"0.965"	"-2.225"	"1.392"
[7,]	NA	"3"	"0.083 "	"0.919"	"1.000"	"-1.725"	"1.892"
[8,]	"2"	"3"	"0.500 "	"1.204"	"0.972"	"-1.868"	"2.868"

```
dataset = anova(data0$job_title_s_data,data0$work.life_balance,tr_matrix_old=dataset)
```

```
[1] "diff $data0work.life_balance by $data0job_title_s_data"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"2.532"	"3"	"0.844"	"1.388"	"0.252"	NA
[3,]	"组内"	"54.133"	"89"	"0.608"	NA	NA	NA
[4,]	"总计"	"56.665"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.833"	"3"	"89"	"0.479"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:6] -0.4 -0.941 -0.941 -0.542 -0.542 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ sigma : Named num [1:6] 0.329 0.784 0.784 0.842 0.842 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ tstat : Named num [1:6] -1.213 -1.2 -1.2 -0.643 -0.643 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ pvalues : num [1:6] 0.59 0.598 0.598 0.907 0.907 ...
..- attr(*, "error")= num 3.29e-05
$ type : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	-3.995e-01	3.294e-01	-1.213	0.590
2 - 0 == 0	-9.412e-01	7.845e-01	-1.200	0.598
3 - 0 == 0	-9.412e-01	7.845e-01	-1.200	0.598
2 - 1 == 0	-5.417e-01	8.424e-01	-0.643	0.907
3 - 1 == 0	-5.417e-01	8.424e-01	-0.643	0.907
3 - 2 == 0	1.665e-15	1.103e+00	0.000	1.000

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"-0.400 "	"0.329"	"0.590"	"-1.047"	"0.248"
[4,]	NA	"2"	"-0.941 "	"0.784"	"0.598"	"-2.484"	"0.602"
[5,]	NA	"3"	"-0.941 "	"0.784"	"0.598"	"-2.484"	"0.602"
[6,]	"1"	"2"	"-0.542 "	"0.842"	"0.907"	"-2.198"	"1.115"
[7,]	NA	"3"	"-0.542 "	"0.842"	"0.907"	"-2.198"	"1.115"
[8,]	"2"	"3"	"0.000 "	"1.103"	"1.000"	"-2.169"	"2.169"

```
dataset = anova(data0$job_title_s_data,data0$well.being,tr_matrix_old=dataset)
```

```
[1] "diff $data0well.being by $data0job_title_s_data"
[1] "model"
      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
[1,] ""      "平方和" "自由度" "均方" "F"      "显著性" NA
[2,] "组间" "0.645" "3"      "0.215" "0.235" "0.871" NA
[3,] "组内" "81.330" "89"     "0.914" NA      NA      NA
[4,] "总计" "81.976" "92"     NA      NA      NA      NA
[1] "leveneTest_result"
      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
[1,] "莱文统计" "自由度1" "自由度2" "显著性" NA      NA      NA
[2,] "1.038"     "3"      "89"     "0.380" NA      NA      NA
[1] "tukey_comparison"
```

Warning in RET\$pfunction("adjusted", ...): lower == upper

List of 7

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:6] -0.173 -0.506 -0.506 -0.333 -0.333 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ sigma      : Named num [1:6] 0.404 0.962 0.962 1.033 1.033 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ tstat       : Named num [1:6] -0.427 -0.526 -0.526 -0.323 -0.323 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ pvalues      : num [1:6] 0.97 0.946 0.946 0.987 0.987 ...
..- attr(*, "error")= num 6.41e-06
$ type         : chr "single-step"
```

```
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	-0.1725	0.4038	-0.427	0.970
2 - 0 == 0	-0.5059	0.9615	-0.526	0.946
3 - 0 == 0	-0.5059	0.9615	-0.526	0.946
2 - 1 == 0	-0.3333	1.0325	-0.323	0.987
3 - 1 == 0	-0.3333	1.0325	-0.323	0.987
3 - 2 == 0	0.0000	1.3519	0.000	1.000

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"-0.173 "	"0.404"	"0.970"	"-0.967"	"0.622"
[4,]	NA	"2"	"-0.506 "	"0.962"	"0.946"	"-2.397"	"1.385"
[5,]	NA	"3"	"-0.506 "	"0.962"	"0.946"	"-2.397"	"1.385"
[6,]	"1"	"2"	"-0.333 "	"1.033"	"0.987"	"-2.364"	"1.697"
[7,]	NA	"3"	"-0.333 "	"1.033"	"0.987"	"-2.364"	"1.697"
[8,]	"2"	"3"	"0.000 "	"1.352"	"1.000"	"-2.659"	"2.659"

```
dataset = anova(data0$work_experience,data0$job_effectiveness,tr_matrix_old=dataset)
```

```
[1] "diff $data0job_effectiveness by $data0work_experience"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"25.011"	"3"	"8.337"	"18.536"	"0.000"	NA
[3,]	"组内"	"40.031"	"89"	"0.450"	NA	NA	NA
[4,]	"总计"	"65.042"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.926"	"3"	"89"	"0.431"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:6] 1.417 2.325 1.988 0.908 0.571 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ sigma : Named num [1:6] 0.4 0.406 0.427 0.157 0.205 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ tstat : Named num [1:6] 3.54 5.73 4.66 5.77 2.79 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ pvalues : num [1:6] 2.97e-03 4.17e-07 1.08e-04 3.73e-07 2.83e-02 ...
```

```
..- attr(*, "error")= num 0.000815
$ type      : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

Fit: aov(formula = y ~ x, data = data)

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	1.4167	0.3996	3.545	0.00297 **
2 - 0 == 0	2.3250	0.4061	5.725	< 0.001 ***
3 - 0 == 0	1.9881	0.4267	4.659	< 0.001 ***
2 - 1 == 0	0.9083	0.1574	5.771	< 0.001 ***
3 - 1 == 0	0.5714	0.2047	2.791	0.02827 *
3 - 2 == 0	-0.3369	0.2171	-1.552	0.38918

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"1.417 **"	"0.400"	"0.003"	"0.631"	"2.203"
[4,]	NA	"2"	"2.325 ***"	"0.406"	"0.000"	"1.526"	"3.124"
[5,]	NA	"3"	"1.988 ***"	"0.427"	"0.000"	"1.149"	"2.827"
[6,]	"1"	"2"	"0.908 ***"	"0.157"	"0.000"	"0.599"	"1.218"
[7,]	NA	"3"	"0.571 *"	"0.205"	"0.028"	"0.169"	"0.974"
[8,]	"2"	"3"	"-0.337 "	"0.217"	"0.389"	"-0.764"	"0.090"

```
dataset = anova(data0$work_experience,data0$work.life_balance,tr_matrix_old=dataset)
```

```
[1] "diff $data0work.life_balance by $data0work_experience"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"7.023"	"3"	"2.341"	"4.197"	"0.008"	NA
[3,]	"组内"	"49.643"	"89"	"0.558"	NA	NA	NA
[4,]	"总计"	"56.665"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.119"	"3"	"89"	"0.949"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:6] 0.806 1.258 1.155 0.452 0.349 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ sigma : Named num [1:6] 0.445 0.452 0.475 0.175 0.228 ...
```



```

..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ tstat      : Named num [1:6] 1.81 2.78 2.43 2.58 1.53 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ pvalues    : num [1:6] 0.2556 0.0293 0.0709 0.0493 0.4024 ...
..- attr(*, "error")= num 0.000974
$ type       : chr "single-step"
- attr(*, "class")= chr "mtest"

```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.8062	0.4450	1.811	0.2556
2 - 0 == 0	1.2583	0.4522	2.782	0.0293 *
3 - 0 == 0	1.1548	0.4752	2.430	0.0709 .
2 - 1 == 0	0.4522	0.1753	2.580	0.0493 *
3 - 1 == 0	0.3486	0.2280	1.529	0.4024
3 - 2 == 0	-0.1036	0.2417	-0.428	0.9711

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间" ""	
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.806 "	"0.445"	"0.256"	"-0.069"	"1.681"
[4,]	NA	"2"	"1.258 *"	"0.452"	"0.029"	"0.369"	"2.148"
[5,]	NA	"3"	"1.155 "	"0.475"	"0.071"	"0.220"	"2.089"
[6,]	"1"	"2"	"0.452 *"	"0.175"	"0.049"	"0.107"	"0.797"
[7,]	NA	"3"	"0.349 "	"0.228"	"0.402"	"-0.100"	"0.797"
[8,]	"2"	"3"	"-0.104 "	"0.242"	"0.971"	"-0.579"	"0.372"

```
dataset = anova(data0$work_experience,data0$well.being,tr_matrix_old=dataset)
```

```
[1] "diff $data0well.being by $data0work_experience"
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"10.387"	"3"	"3.462"	"4.304"	"0.007"	NA
[3,]	"组内"	"71.589"	"89"	"0.804"	NA	NA	NA
[4,]	"总计"	"81.976"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.776"	"3"	"89"	"0.510"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```
$ qfunction      :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:6] 0.861 1.475 1.101 0.614 0.241 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ sigma         : Named num [1:6] 0.534 0.543 0.571 0.21 0.274 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ tstat         : Named num [1:6] 1.61 2.716 1.93 2.92 0.879 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ pvalues       : num [1:6] 0.3566 0.0347 0.2059 0.0202 0.8026 ...
..- attr(*, "error")= num 0.000881
$ type          : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.8605	0.5344	1.610	0.3566
2 - 0 == 0	1.4750	0.5431	2.716	0.0347 *
3 - 0 == 0	1.1012	0.5706	1.930	0.2059
2 - 1 == 0	0.6145	0.2105	2.920	0.0202 *
3 - 1 == 0	0.2407	0.2738	0.879	0.8026
3 - 2 == 0	-0.3738	0.2903	-1.288	0.5517

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.861 "	"0.534"	"0.357"	"-0.191"	"1.912"
[4,]	NA	"2"	"1.475 *"	"0.543"	"0.035"	"0.407"	"2.543"
[5,]	NA	"3"	"1.101 "	"0.571"	"0.206"	"-0.021"	"2.223"
[6,]	"1"	"2"	"0.614 *"	"0.210"	"0.020"	"0.201"	"1.028"
[7,]	NA	"3"	"0.241 "	"0.274"	"0.803"	"-0.298"	"0.779"
[8,]	"2"	"3"	"-0.374 "	"0.290"	"0.552"	"-0.945"	"0.197"

```
dataset = anova(data0$marital_status,data0$job_effectiveness,tr_matrix_old=dataset)
```

```
[1] "diff $data0job_effectiveness by $data0marital_status"
[1] "model"
      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
[1,] ""      "平方和" "自由度" "均方"  "F"      "显著性" NA
[2,] "组间"  "11.549" "4"      "2.887" "4.750" "0.002" NA
[3,] "组内"  "53.493" "88"     "0.608" NA      NA      NA
[4,] "总计"  "65.042" "92"     NA      NA      NA      NA
[1] "leveneTest_result"
      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
[1,] "莱文统计" "自由度1" "自由度2" "显著性" NA      NA      NA
```

```
[2,] "1.835"      "4"      "88"      "0.129" NA    NA    NA
[1] "tukey_comparison"
List of 7
 $ pfunction      :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
 $ qfunction      :function (conf.level, adjusted = TRUE, ...)
 $ coefficients: Named num [1:10] 0.479 0.788 0.926 0.426 0.309 ...
 ..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
 $ sigma         : Named num [1:10] 0.257 0.194 0.314 0.418 0.242 ...
 ..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
 $ tstat         : Named num [1:10] 1.87 4.06 2.95 1.02 1.28 ...
 ..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
 $ pvalues       : num [1:10] 0.324078 0.000896 0.02929 0.834869 0.689899 ...
 ..- attr(*, "error")= num 0.000111
 $ type          : chr "single-step"
 - attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

Fit: aov(formula = y ~ x, data = data)

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.47950	0.25678	1.867	0.324078
2 - 0 == 0	0.78843	0.19419	4.060	0.000896 ***
3 - 0 == 0	0.92593	0.31384	2.950	0.029290 *
4 - 0 == 0	0.42593	0.41771	1.020	0.834869
2 - 1 == 0	0.30893	0.24211	1.276	0.689899
3 - 1 == 0	0.44643	0.34555	1.292	0.679929
4 - 1 == 0	-0.05357	0.44203	-0.121	0.999945
3 - 2 == 0	0.13750	0.30196	0.455	0.990146
4 - 2 == 0	-0.36250	0.40886	-0.887	0.893571
4 - 3 == 0	-0.50000	0.47744	-1.047	0.821095

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.479 "	"0.257"	"0.324"	"-0.026"	"0.985"
[4,]	NA	"2"	"0.788 ***"	"0.194"	"0.001"	"0.406"	"1.170"
[5,]	NA	"3"	"0.926 *"	"0.314"	"0.029"	"0.309"	"1.543"
[6,]	NA	"4"	"0.426 "	"0.418"	"0.835"	"-0.396"	"1.247"
[7,]	"1"	"2"	"0.309 "	"0.242"	"0.690"	"-0.167"	"0.785"
[8,]	NA	"3"	"0.446 "	"0.346"	"0.680"	"-0.233"	"1.126"
[9,]	NA	"4"	"-0.054 "	"0.442"	"1.000"	"-0.923"	"0.816"
[10,]	"2"	"3"	"0.138 "	"0.302"	"0.990"	"-0.456"	"0.731"
[11,]	NA	"4"	"-0.363 "	"0.409"	"0.894"	"-1.167"	"0.442"
[12,]	"3"	"4"	"-0.500 "	"0.477"	"0.821"	"-1.439"	"0.439"

```
dataset = anova(data0$marital_status,data0$work.life_balance,tr_matrix_old=dataset)
```

```
[1] "diff $data0work.life_balance by $data0marital_status"
```

```
[1] "model"
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
```

```
[1,] ""      "平方和" "自由度" "均方" "F"      "显著性" NA
```

```
[2,] "组间" "9.248" "4"      "2.312" "4.291" "0.003" NA
```

```
[3,] "组内" "47.417" "88"      "0.539" NA      NA      NA
```

```
[4,] "总计" "56.665" "92"      NA      NA      NA      NA
```

```
[1] "leveneTest_result"
```

```
      [,1]      [,2]      [,3]      [,4]      [,5] [,6] [,7]
```

```
[1,] "莱文统计" "自由度1" "自由度2" "显著性" NA      NA      NA
```

```
[2,] "0.586" "4"      "88"      "0.674" NA      NA      NA
```

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```
$ qfunction :function (conf.level, adjusted = TRUE, ...)
```

```
$ coefficients: Named num [1:10] 0.476 0.74 0.659 0.315 0.264 ...
```

```
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ sigma : Named num [1:10] 0.242 0.183 0.295 0.393 0.228 ...
```

```
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ tstat : Named num [1:10] 1.97 4.05 2.23 0.8 1.16 ...
```

```
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ pvalues : num [1:10] 0.27354 0.00093 0.16613 0.92413 0.76006 ...
```

```
..- attr(*, "error")= num 8.19e-05
```

```
$ type : chr "single-step"
```

```
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.47553	0.24175	1.967	0.27354
2 - 0 == 0	0.73981	0.18283	4.046	0.00093 ***
3 - 0 == 0	0.65856	0.29548	2.229	0.16613
4 - 0 == 0	0.31481	0.39327	0.800	0.92413
2 - 1 == 0	0.26429	0.22794	1.159	0.76006
3 - 1 == 0	0.18304	0.32533	0.563	0.97827
4 - 1 == 0	-0.16071	0.41617	-0.386	0.99475
3 - 2 == 0	-0.08125	0.28430	-0.286	0.99837
4 - 2 == 0	-0.42500	0.38494	-1.104	0.79110
4 - 3 == 0	-0.34375	0.44951	-0.765	0.93507

```
----
```

```
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
(Adjusted p values reported -- single-step method)
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
```

```
[1,] ""      ""      ""      ""      ""      "95%置信区间" ""
```

	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[2,]	"0"	"1"	"0.476 "	"0.242"	"0.274"	"0.000"	"0.951"
[4,]	NA	"2"	"0.740 ***"	"0.183"	"0.001"	"0.380"	"1.099"
[5,]	NA	"3"	"0.659 "	"0.295"	"0.166"	"0.077"	"1.240"
[6,]	NA	"4"	"0.315 "	"0.393"	"0.924"	"-0.459"	"1.088"
[7,]	"1"	"2"	"0.264 "	"0.228"	"0.760"	"-0.184"	"0.713"
[8,]	NA	"3"	"0.183 "	"0.325"	"0.978"	"-0.457"	"0.823"
[9,]	NA	"4"	"-0.161 "	"0.416"	"0.995"	"-0.979"	"0.658"
[10,]	"2"	"3"	"-0.081 "	"0.284"	"0.998"	"-0.640"	"0.478"
[11,]	NA	"4"	"-0.425 "	"0.385"	"0.791"	"-1.182"	"0.332"
[12,]	"3"	"4"	"-0.344 "	"0.450"	"0.935"	"-1.228"	"0.540"

```
dataset = anova(data0$marital_status,data0$well.being,tr_matrix_old=dataset)
```

```
[1] "diff $data0well.being by $data0marital_status"
```

```
[1] "model"
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
```

```
[1,] ""          "平方和" "自由度" "均方"    "F"          "显著性" NA
```

```
[2,] "组间" "16.836" "4"      "4.209" "5.686" "0.000" NA
```

```
[3,] "组内" "65.139" "88"     "0.740" NA      NA      NA
```

```
[4,] "总计" "81.976" "92"     NA      NA      NA      NA
```

```
[1] "leveneTest_result"
```

```
      [,1]      [,2]      [,3]      [,4]      [,5] [,6] [,7]
```

```
[1,] "莱文统计" "自由度1" "自由度2" "显著性" NA      NA      NA
```

```
[2,] "1.065"    "4"      "88"      "0.379" NA      NA      NA
```

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```
$ qfunction :function (conf.level, adjusted = TRUE, ...)
```

```
$ coefficients: Named num [1:10] 0.575 0.992 0.83 0.236 0.417 ...
```

```
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ sigma : Named num [1:10] 0.283 0.214 0.346 0.461 0.267 ...
```

```
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ tstat : Named num [1:10] 2.031 4.631 2.396 0.512 1.561 ...
```

```
..- attr(*, "names")= chr [1:10] "1 - 0" "2 - 0" "3 - 0" "4 - 0" ...
```

```
$ pvalues : num [1:10] 0.244041 0.000102 0.116209 0.984655 0.506516 ...
```

```
..- attr(*, "error")= num 0.000135
```

```
$ type : chr "single-step"
```

```
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.5754	0.2834	2.031	0.244041
2 - 0 == 0	0.9924	0.2143	4.631	0.000102 ***
3 - 0 == 0	0.8299	0.3463	2.396	0.116209
4 - 0 == 0	0.2361	0.4609	0.512	0.984655

```

2 - 1 == 0    0.4170    0.2672    1.561 0.506516
3 - 1 == 0    0.2545    0.3813    0.667 0.959661
4 - 1 == 0   -0.3393    0.4878   -0.696 0.953290
3 - 2 == 0   -0.1625    0.3332   -0.488 0.987236
4 - 2 == 0   -0.7563    0.4512   -1.676 0.433943
4 - 3 == 0   -0.5938    0.5269   -1.127 0.778477
---

```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"0.575 "	"0.283"	"0.244"	"0.018"	"1.133"
[4,]	NA	"2"	"0.992 ***"	"0.214"	"0.000"	"0.571"	"1.414"
[5,]	NA	"3"	"0.830 "	"0.346"	"0.116"	"0.149"	"1.511"
[6,]	NA	"4"	"0.236 "	"0.461"	"0.985"	"-0.670"	"1.143"
[7,]	"1"	"2"	"0.417 "	"0.267"	"0.507"	"-0.109"	"0.942"
[8,]	NA	"3"	"0.254 "	"0.381"	"0.960"	"-0.496"	"1.004"
[9,]	NA	"4"	"-0.339 "	"0.488"	"0.953"	"-1.299"	"0.620"
[10,]	"2"	"3"	"-0.162 "	"0.333"	"0.987"	"-0.818"	"0.493"
[11,]	NA	"4"	"-0.756 "	"0.451"	"0.434"	"-1.644"	"0.131"
[12,]	"3"	"4"	"-0.594 "	"0.527"	"0.778"	"-1.630"	"0.442"

```
dataset = anova(data0$partners_work,data0$job_effectiveness,tr_matrix_old=dataset)
```

```
[1] "diff $data0job_effectiveness by $data0partners_work"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
--	------	------	------	------	------	------	------

[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
------	----	-------	-------	------	-----	-------	----

[2,]	"组间"	"4.569"	"3"	"1.523"	"2.241"	"0.089"	NA
------	------	---------	-----	---------	---------	---------	----

[3,]	"组内"	"60.473"	"89"	"0.679"	NA	NA	NA
------	------	----------	------	---------	----	----	----

[4,]	"总计"	"65.042"	"92"	NA	NA	NA	NA
------	------	----------	------	----	----	----	----

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
--	------	------	------	------	------	------	------

[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
------	--------	--------	--------	-------	----	----	----

[2,]	"0.367"	"3"	"89"	"0.777"	NA	NA	NA
------	---------	-----	------	---------	----	----	----

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```
$ qfunction :function (conf.level, adjusted = TRUE, ...)
```

```
$ coefficients: Named num [1:6] 1.065 0.315 0.165 -0.75 -0.9 ...
```

```
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```

```
$ sigma : Named num [1:6] 0.489 0.186 0.385 0.498 0.602 ...
```

```
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```

```
$ tstat : Named num [1:6] 2.178 1.695 0.428 -1.505 -1.495 ...
```

```
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```

```
$ pvalues : num [1:6] 0.123 0.307 0.971 0.412 0.418 ...
```

```
..- attr(*, "error")= num 0.000517
```

```
$ type : chr "single-step"
```

```
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	1.0648	0.4890	2.178	0.123
2 - 0 == 0	0.3148	0.1857	1.695	0.307
3 - 0 == 0	0.1648	0.3853	0.428	0.971
2 - 1 == 0	-0.7500	0.4984	-1.505	0.412
3 - 1 == 0	-0.9000	0.6020	-1.495	0.418
3 - 2 == 0	-0.1500	0.3973	-0.378	0.979

(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"1.065 "	"0.489"	"0.123"	"0.103"	"2.026"
[4,]	NA	"2"	"0.315 "	"0.186"	"0.307"	"-0.051"	"0.680"
[5,]	NA	"3"	"0.165 "	"0.385"	"0.971"	"-0.593"	"0.923"
[6,]	"1"	"2"	"-0.750 "	"0.498"	"0.412"	"-1.730"	"0.230"
[7,]	NA	"3"	"-0.900 "	"0.602"	"0.418"	"-2.084"	"0.284"
[8,]	"2"	"3"	"-0.150 "	"0.397"	"0.979"	"-0.931"	"0.631"

```
dataset = anova(data0$partners_work,data0$work.life_balance,tr_matrix_old=dataset)
```

```
[1] "diff $data0work.life_balance by $data0partners_work"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
--	------	------	------	------	------	------	------

[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
------	----	-------	-------	------	-----	-------	----

[2,]	"组间"	"5.662"	"3"	"1.887"	"3.294"	"0.024"	NA
------	------	---------	-----	---------	---------	---------	----

[3,]	"组内"	"51.003"	"89"	"0.573"	NA	NA	NA
------	------	----------	------	---------	----	----	----

[4,]	"总计"	"56.665"	"92"	NA	NA	NA	NA
------	------	----------	------	----	----	----	----

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
--	------	------	------	------	------	------	------

[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
------	--------	--------	--------	-------	----	----	----

[2,]	"2.057"	"3"	"89"	"0.112"	NA	NA	NA
------	---------	-----	------	---------	----	----	----

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
```

```
$ qfunction :function (conf.level, adjusted = TRUE, ...)
```

```
$ coefficients: Named num [1:6] 1.1019 0.3814 0.0185 -0.7204 -1.0833 ...
```

```
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```

```
$ sigma : Named num [1:6] 0.449 0.171 0.354 0.458 0.553 ...
```

```
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```

```
$ tstat : Named num [1:6] 2.4538 2.236 0.0523 -1.574 -1.9596 ...
```

```
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```

```
$ pvalues : num [1:6] 0.0657 0.1081 0.9999 0.372 0.1916 ...
```

```
..- attr(*, "error")= num 0.000923
```

```
$ type : chr "single-step"
```

```
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	1.10185	0.44904	2.454	0.0657 .
2 - 0 == 0	0.38142	0.17058	2.236	0.1081
3 - 0 == 0	0.01852	0.35387	0.052	0.9999
2 - 1 == 0	-0.72043	0.45772	-1.574	0.3720
3 - 1 == 0	-1.08333	0.55284	-1.960	0.1916
3 - 2 == 0	-0.36290	0.36483	-0.995	0.7328

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported — single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"1.102 "	"0.449"	"0.066"	"0.219"	"1.985"
[4,]	NA	"2"	"0.381 "	"0.171"	"0.108"	"0.046"	"0.717"
[5,]	NA	"3"	"0.019 "	"0.354"	"1.000"	"-0.677"	"0.715"
[6,]	"1"	"2"	"-0.720 "	"0.458"	"0.372"	"-1.621"	"0.180"
[7,]	NA	"3"	"-1.083 "	"0.553"	"0.192"	"-2.171"	"0.004"
[8,]	"2"	"3"	"-0.363 "	"0.365"	"0.733"	"-1.080"	"0.355"

```
dataset = anova(data0$partners_work,data0$well.being,tr_matrix_old=dataset)
```

```
[1] "diff $data0well.being by $data0partners_work"
```

```
[1] "model"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	"平方和"	"自由度"	"均方"	"F"	"显著性"	NA
[2,]	"组间"	"8.226"	"3"	"2.742"	"3.309"	"0.024"	NA
[3,]	"组内"	"73.750"	"89"	"0.829"	NA	NA	NA
[4,]	"总计"	"81.976"	"92"	NA	NA	NA	NA

```
[1] "leveneTest_result"
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	"莱文统计"	"自由度1"	"自由度2"	"显著性"	NA	NA	NA
[2,]	"0.466"	"3"	"89"	"0.707"	NA	NA	NA

```
[1] "tukey_comparison"
```

```
List of 7
```

```
$ pfunction :function (type = c("univariate", "adjusted", p.adjust.methods), ...)
$ qfunction :function (conf.level, adjusted = TRUE, ...)
$ coefficients: Named num [1:6] 1.31 0.477 0.177 -0.833 -1.133 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ sigma : Named num [1:6] 0.54 0.205 0.426 0.55 0.665 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
$ tstat : Named num [1:6] 2.426 2.325 0.416 -1.514 -1.705 ...
..- attr(*, "names")= chr [1:6] "1 - 0" "2 - 0" "3 - 0" "2 - 1" ...
```



```
$ pvalues      : num [1:6] 0.0701 0.0887 0.973 0.4062 0.3021 ...
.- attr(*, "error")= num 0.000518
$ type        : chr "single-step"
- attr(*, "class")= chr "mtest"
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	1.3102	0.5400	2.426	0.0701 .
2 - 0 == 0	0.4769	0.2051	2.325	0.0887 .
3 - 0 == 0	0.1769	0.4255	0.416	0.9730
2 - 1 == 0	-0.8333	0.5504	-1.514	0.4062
3 - 1 == 0	-1.1333	0.6648	-1.705	0.3021
3 - 2 == 0	-0.3000	0.4387	-0.684	0.8934

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]
[1,]	""	""	""	""	""	"95%置信区间"	""
[2,]	"group"	"group"	"平均值差值"	"标准差"	"显著性"	"下限"	"上限"
[3,]	"0"	"1"	"1.310 "	"0.540"	"0.070"	"0.248"	"2.372"
[4,]	NA	"2"	"0.477 "	"0.205"	"0.089"	"0.073"	"0.880"
[5,]	NA	"3"	"0.177 "	"0.426"	"0.973"	"-0.660"	"1.014"
[6,]	"1"	"2"	"-0.833 "	"0.550"	"0.406"	"-1.916"	"0.249"
[7,]	NA	"3"	"-1.133 "	"0.665"	"0.302"	"-2.441"	"0.174"
[8,]	"2"	"3"	"-0.300 "	"0.439"	"0.893"	"-1.163"	"0.563"

```
saveExcel(dataset,"ANOVA")
rm(anova)
```

TWO-WAY ANOVA 双因素方差分析步骤

LSD.test and Bonferroni, duncan.test

```
library(car)
library(multcomp)
library(knitr)
library(xtable)
library(agricolae)
library(multcomp)
library(multcompView)
library(emmeans)
## 按性别分析因变量
x0 <- factor( as.character(data0$age) )
x1 <- factor( as.character(data0$gender) )
y <- data0$job_effectiveness
```

```
data = data.frame(x0,x1,y)
model <- aov(y ~ x0 * x1, data=data)
model <- aov(y ~ x0 * x1, data=data)
data_1 = data.frame(x0=data0$age,x1=data0$gender,y=data0$job_effectiveness)
model_1 <- aov(y ~ x0 * x1, data=data_1)

# 如果ANOVA拒绝原假设, 执行Fisher's LSD
#if (summary(model)$'Pr(>F)')[1] < 0.05) {
  print("~~~~~ LSD.test")
}
```

```
[1] "~~~~~ LSD.test"
```

```
pairwise_results <- LSD.test(model_1, "x0", p.adj="bonferroni") #bonferroni#对p值进行修正
pairwise_results
```

```
$statistics
```

```
      MSerror Df      Mean      CV
0.6866993 89 4.083333 20.29404
```

```
$parameters
```

```
      test p.adjusted name.t ntr alpha
Fisher-LSD bonferroni      x0  5  0.05
```

```
$means
```

```
      y      std  r      se      LCL      UCL  Min  Max   Q25 Q50   Q75
0 3.416667 0.6373774  9 0.2762244 2.867815 3.965519 2.25 4.00 3.0000 3.5 4.000
1 3.943182 0.8861763 22 0.1766737 3.592135 4.294229 2.00 5.50 3.0625 4.0 4.500
2 4.318182 0.8552977 33 0.1442535 4.031553 4.604810 2.00 5.75 4.0000 4.5 5.000
3 4.202381 0.7608344 21 0.1808313 3.843073 4.561689 3.00 5.50 3.7500 4.0 4.750
4 3.937500 0.7165144  8 0.2929802 3.355355 4.519645 3.00 5.25 3.4375 4.0 4.125
```

```
$comparison
```

```
NULL
```

```
$groups
```

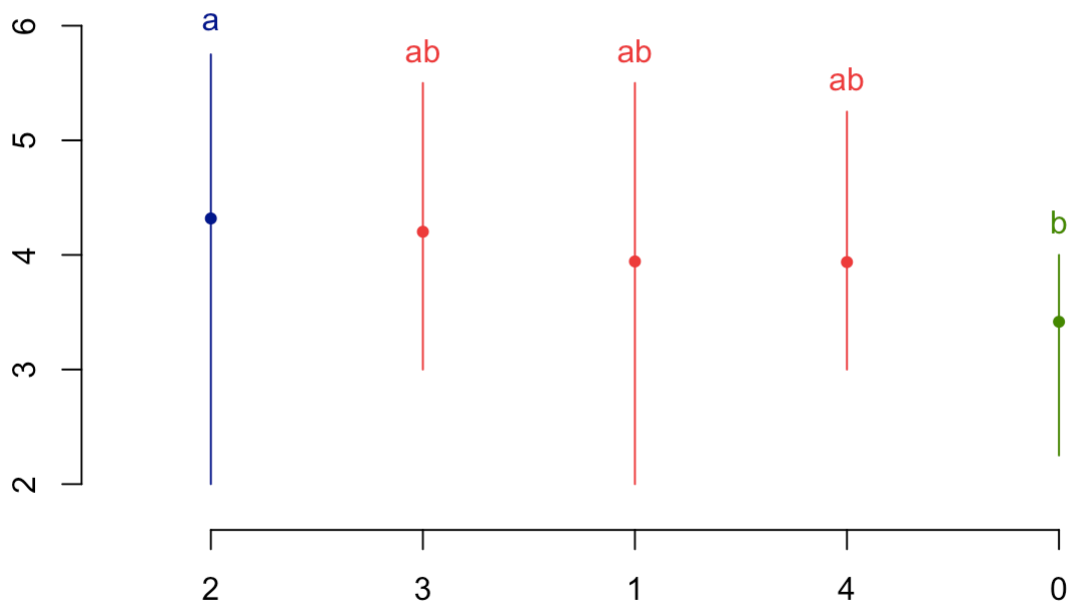
```
      y groups
2 4.318182    a
3 4.202381   ab
1 3.943182   ab
4 3.937500   ab
0 3.416667    b
```

```
attr(,"class")
```

```
[1] "group"
```

```
plot(pairwise_results)
```

Groups and Range



```
print("~~~~~")
```

```
[1] "~~~~~"
```

```
#}
```

```
leveneTest_result = leveneTest(model)
```

```
# 进行图基事后比较
```

```
tukey_comparison0 <- glht(model, linfct = mcp(x0 = "Tukey"))
```

Warning in mcp2matrix(model, linfct = linfct): covariate interactions found -- default contrast might be inappropriate

```
tukey_comparison1 <- glht(model, linfct = mcp(x1 = "Tukey"))
```

Warning in mcp2matrix(model, linfct = linfct): covariate interactions found -- default contrast might be inappropriate

```
tuk <- TukeyHSD(model)
```

```
duncan <- duncan.test(model, 'x0')
```

```
# 进行邓尼特事后比较, 其中"control"是对照组
```

```
#dunnett_comparison <- glht(model, linfct = mcp(x = "Dunnett"))
```

```
#xtable(summary(model), type = "html", digits = 2, width = "600px", include.rownames =
```

```
#kable(summary(model), format = "html")
summary(model)
```

```

      Df Sum Sq Mean Sq F value Pr(>F)
x0      4   6.72   1.6800   2.443  0.053 .
x1      1   1.14   1.1355   1.651  0.202
x0:x1    4   0.11   0.0275   0.040  0.997
Residuals 83  57.08   0.6877
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(leveneTest_result)
```

```

      Df      F value      Pr(>F)
Min.   : 9.0   Min.   :0.9691   Min.   :0.4715
1st Qu.:27.5   1st Qu.:0.9691   1st Qu.:0.4715
Median :46.0   Median :0.9691   Median :0.4715
Mean    :46.0   Mean    :0.9691   Mean    :0.4715
3rd Qu.:64.5   3rd Qu.:0.9691   3rd Qu.:0.4715
Max.    :83.0   Max.    :0.9691   Max.    :0.4715
      NA's :1      NA's :1
```

```
summary(tukey_comparison0)
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

Fit: aov(formula = y ~ x0 * x1, data = data)

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	0.5000	0.4414	1.133	0.779
2 - 0 == 0	0.9125	0.4146	2.201	0.179
3 - 0 == 0	0.8077	0.4364	1.851	0.338
4 - 0 == 0	0.6667	0.6056	1.101	0.796
2 - 1 == 0	0.4125	0.3028	1.362	0.640
3 - 1 == 0	0.3077	0.3320	0.927	0.879
4 - 1 == 0	0.1667	0.5353	0.311	0.998
3 - 2 == 0	-0.1048	0.2954	-0.355	0.996
4 - 2 == 0	-0.2458	0.5134	-0.479	0.988
4 - 3 == 0	-0.1410	0.5311	-0.266	0.999

(Adjusted p values reported -- single-step method)

```
summary(tukey_comparison1)
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = y ~ x0 * x1, data = data)
```

Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t)
1 - 0 == 0	-0.1875	0.5563	-0.337	0.737

(Adjusted p values reported -- single-step method)

```
summary(tuk)
```

	Length	Class	Mode
x0	40	-none-	numeric
x1	4	-none-	numeric
x0:x1	180	-none-	numeric

```
duncan
```

```
$statistics
```

MSerror	Df	Mean	CV
0.6876646	83	4.083333	20.3083

```
$parameters
```

test name	t	ntr	alpha
Duncan	x0	5	0.05

```
$duncan
```

```
NULL
```

```
$means
```

	y	std	r	se	Min	Max	Q25	Q50	Q75
0	3.416667	0.6373774	9	0.2764185	2.25	4.00	3.0000	3.5	4.000
1	3.943182	0.8861763	22	0.1767979	2.00	5.50	3.0625	4.0	4.500
2	4.318182	0.8552977	33	0.1443548	2.00	5.75	4.0000	4.5	5.000
3	4.202381	0.7608344	21	0.1809584	3.00	5.50	3.7500	4.0	4.750
4	3.937500	0.7165144	8	0.2931861	3.00	5.25	3.4375	4.0	4.125

```
$comparison
```

```
NULL
```

```
$groups
```

	y	groups
2	4.318182	a
3	4.202381	a
1	3.943182	ab
4	3.937500	ab
0	3.416667	b

```
attr("class")
```

```
[1] "group"
```

```
#multcomp_plot(tukey_comparison)
```

```
#summary(dunnett_comparison)
```

相关性分析

Pearson(皮尔逊)相关系数

```
library(corrplot)
```

corrplot 0.92 loaded

```
getCol <-function(data){
  correlation_result0 <- corr.test(data)
  r_value = correlation_result0$r
  p_value = correlation_result0$p
  # 使用 for 循环遍历矩阵
  for (i in 1:nrow(p_value)) {
    for (j in 1:ncol(p_value)) {
      # 访问矩阵元素
      #保留上三角
      if(i>1&&j<i){
        r_value[i, j] = NA
        next
      }
      c = ""
      if( p_value[i,j]<0.001){
        c="***"
      }else if( p_value[i,j]<0.01){
        c="**"
      }else if( p_value[i,j]<0.05){
        c="*"
      }
      r_value[i, j]=format(as.numeric(r_value[i, j]), digits = 3)
      #r_value[i, j]=round(as.numeric(r_value[i, j]),3)
      #r_value[i, j]=paste(as.character(r_value[i, j]),c)
      r_value[i, j]=paste(r_value[i, j],c)
    }
  }

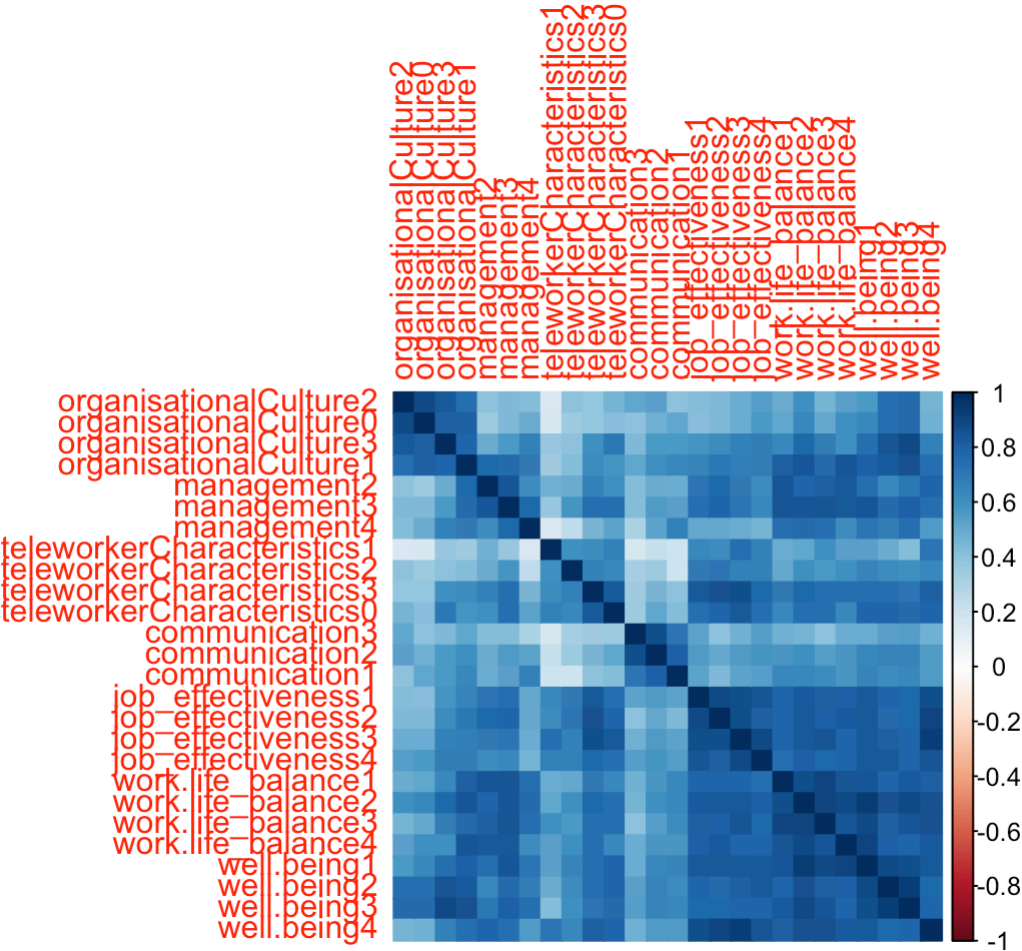
  return(r_value)
}

data <- data0[c(cols_independent,cols_dependent)]
r_value = getCol(data)
kable(r_value)
```

	organisationalCulture2	organisationalCulture0	organisationalCulture3	organisationalCulture1
organisationalCulture2	1 ***	0.889 ***	0.839 ***	0.755 ***
organisationalCulture0	NA	1 ***	0.816 ***	0.813 ***
organisationalCulture3	NA	NA	1 ***	0.798 ***
organisationalCulture1	NA	NA	NA	1 ***

	organisationalCulture2	organisationalCulture0	organisationalCulture3	organi
management2	NA	NA	NA	NA
management3	NA	NA	NA	NA
management4	NA	NA	NA	NA
teleworkerCharacteristics1	NA	NA	NA	NA
teleworkerCharacteristics2	NA	NA	NA	NA
teleworkerCharacteristics3	NA	NA	NA	NA
teleworkerCharacteristics0	NA	NA	NA	NA
communication3	NA	NA	NA	NA
communication2	NA	NA	NA	NA
communication1	NA	NA	NA	NA
job_effectiveness1	NA	NA	NA	NA
job_effectiveness2	NA	NA	NA	NA
job_effectiveness3	NA	NA	NA	NA
job_effectiveness4	NA	NA	NA	NA
work.life_balance1	NA	NA	NA	NA
work.life_balance2	NA	NA	NA	NA
work.life_balance3	NA	NA	NA	NA
work.life_balance4	NA	NA	NA	NA
well.being1	NA	NA	NA	NA
well.being2	NA	NA	NA	NA
well.being3	NA	NA	NA	NA
well.being4	NA	NA	NA	NA

```
correlation_result0 = cor(data)
corrplot(correlation_result0, method = "color")
```



```
saveExcel(r_value,"correlation Pearson")

data <- data0[all_var_calculated]
r_value = getCol(data)
kable(r_value)
```

	teleworkerCharacteristics	communication	management	organisationalCulture
teleworkerCharacteristics	1 ***	0.561 ***	0.639 ***	0.522 ***
communication	NA	1 ***	0.586 ***	0.534 ***
management	NA	NA	1 ***	0.639 ***
organisationalCulture	NA	NA	NA	1 ***
job_effectiveness	NA	NA	NA	NA
work.life_balance	NA	NA	NA	NA
well.being	NA	NA	NA	NA

```
correlation_result0 = cor(data)
corrplot(correlation_result0, method = "color")
```




```
saveExcel(r_value,"correlation Pearson2")
#data <- data0[cols_dependent]
#r_value = getCol(data)
#kable(r_value)
#correlation_result1 = cor(data)
#corrplot(correlation_result1, method = "color")
```

回归分析

一元回归分析

略... ## 多元回归分析 VIF 共线性诊断 Durbin-Watson (DW): 容差值 (Tolerance) 是VIF的倒数, 即 $Tolerance = 1/VIF$ Durbin-Watson (DW): 在“2”附近不存在序列相关, 非伪回归方程; • 小于“2”存在正自相关; • 大于“2”存在负自相关

```
library(ggplot2)
library(car)
library(lmtest)
```

Loading required package: zoo

Attaching package: 'zoo'

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

as.Date, as.Date.numeric

```
#environment, jobCharacteristics, teleworkerCharacteristics, communication, management
#job_effectiveness, work.life_balance, well.being
#"environment","jobCharacteristics","teleworkerCharacteristics","communication","manag
#data = data0[, c("environment","jobCharacteristics","teleworkerCharacteristics","comm
data = data0[, c("teleworkerCharacteristics","communication","management","organisatio
mylm<-function(...,data=NULL){
  # 因变 ~ 自变量1 + 自变量2 + ... ,
  #model <- lm(work.life_balance ~ teleworkerCharacteristics + communication + managem
  model <- lm(..., data = data)
  # 查看模型摘要, 获取回归系数、标准误、t值和p值等信息
  sum = summary(model)
  print(sum)
  #result = cor(data) #变量间如果相关性为1,则不能进行vif验证
  #print(result)
  vif = vif(model)
  #容差值 (Tolerance) 是VIF的倒数, 即Tolerance = 1/VIF
  Tolerance = 1/vif
  print(vif)
  print(Tolerance)
  # Durbin-Watson (DW): 德宾沃森
  dw_test <- dwtest(model)
  print(dw_test)
  #coef(model)
  #predict(model)
  AIC(model)
  BIC(model)
  #plot(model$resid)
  # 输出模型的详细结果
  print(model)
  # 预测新数据点的mpg值
  # 假设我们有一个新的数据点, 马力为120, 车重为3
  #newdata <- data.frame(environment = 5.2, jobCharacteristics = 6)
  #predictions <- predict(model, newdata)
  #print(predictions)
  # 绘制回归拟合线
  # 首先, 安装并加载所需的绘图包
  # 创建散点图并添加拟合线
  #ggplot(mtcars, aes(x = environment, y = job_effectiveness, color = factor(cyl))) +
  #  geom_point() +
  #  geom_smooth(method = lm, se = FALSE, formula = job_effectiveness ~ environment) +
  #  labs(title = "Regression of mpg on hp", x = "Horsepower", y = "Miles/(US) gallon"
}
```

```
mylm(work.life_balance ~ teleworkerCharacteristics + communication + management + orga
```

Call:

```
lm(formula = ..1, data = data)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.65736	-0.13461	-0.01356	0.11863	0.52079

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.69842	0.12330	5.664	1.83e-07 ***
teleworkerCharacteristics	0.21022	0.03445	6.102	2.75e-08 ***
communication	0.11732	0.03723	3.151	0.00222 **
management	0.40776	0.03077	13.252	< 2e-16 ***
organisationalCulture	0.08599	0.02089	4.116	8.65e-05 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.2194 on 88 degrees of freedom

Multiple R-squared: 0.9252, Adjusted R-squared: 0.9218

F-statistic: 272.2 on 4 and 88 DF, p-value: < 2.2e-16

teleworkerCharacteristics	communication	management
1.884931	1.750454	2.288005
organisationalCulture		
1.839745		
teleworkerCharacteristics	communication	management
0.5305233	0.5712803	0.4370619
organisationalCulture		
0.5435536		

Durbin-Watson test

data: model

DW = 2.0762, p-value = 0.6304

alternative hypothesis: true autocorrelation is greater than 0

Call:

lm(formula = ..1, data = data)

Coefficients:

(Intercept)	teleworkerCharacteristics
0.69842	0.21022
communication	management
0.11732	0.40776
organisationalCulture	
0.08599	

```
mylm(job_effectiveness ~ teleworkerCharacteristics + communication + management + orga
```

Call:

lm(formula = ..1, data = data)

Residuals:

Min	1Q	Median	3Q	Max
-----	----	--------	----	-----

-0.63471 -0.13687 0.01613 0.14390 0.57467

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	0.33096	0.14615	2.265	0.02600	*
teleworkerCharacteristics	0.58346	0.04084	14.287	< 2e-16	***
communication	0.24359	0.04413	5.520	3.39e-07	***
management	0.10474	0.03647	2.872	0.00511	**
organisationalCulture	0.05452	0.02476	2.202	0.03029	*

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.2601 on 88 degrees of freedom
Multiple R-squared: 0.9085, Adjusted R-squared: 0.9043
F-statistic: 218.4 on 4 and 88 DF, p-value: < 2.2e-16

teleworkerCharacteristics	communication	management
1.884931	1.750454	2.288005
organisationalCulture		
1.839745		
teleworkerCharacteristics	communication	management
0.5305233	0.5712803	0.4370619
organisationalCulture		
0.5435536		

Durbin-Watson test

data: model
DW = 1.7299, p-value = 0.08962
alternative hypothesis: true autocorrelation is greater than 0

Call:
lm(formula = ..1, data = data)

Coefficients:	(Intercept)	teleworkerCharacteristics
	0.33096	0.58346
	communication	management
	0.24359	0.10474
	organisationalCulture	
	0.05452	

```
mylm(well.being ~ teleworkerCharacteristics + communication + management + organisatio
```

Call:
lm(formula = ..1, data = data)

Residuals:	Min	1Q	Median	3Q	Max
	-0.59375	-0.10624	-0.01171	0.10151	0.42853

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-0.01850	0.09900	-0.187	0.852
teleworkerCharacteristics	0.37731	0.02766	13.640	< 2e-16 ***
communication	0.17162	0.02989	5.741	1.32e-07 ***
management	0.25245	0.02470	10.219	< 2e-16 ***
organisationalCulture	0.23158	0.01677	13.808	< 2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.1762 on 88 degrees of freedom

Multiple R-squared: 0.9667, Adjusted R-squared: 0.9652

F-statistic: 638.3 on 4 and 88 DF, p-value: < 2.2e-16

teleworkerCharacteristics	communication	management
1.884931	1.750454	2.288005
organisationalCulture		
1.839745		
teleworkerCharacteristics	communication	management
0.5305233	0.5712803	0.4370619
organisationalCulture		
0.5435536		

Durbin-Watson test

data: model

DW = 1.9892, p-value = 0.4782

alternative hypothesis: true autocorrelation is greater than 0

Call:

lm(formula = ..1, data = data)

Coefficients:

(Intercept)	teleworkerCharacteristics
-0.0185	0.3773
communication	management
0.1716	0.2525
organisationalCulture	
0.2316	

```
#mylm(work.life_balance ~ teleworkerCharacteristics + communication + management ,data
#mylm(work.life_balance ~ teleworkerCharacteristics + communication + organisationalCu
#mylm(work.life_balance ~ teleworkerCharacteristics + management + organisationalCult
#mylm(work.life_balance ~ communication + management + organisationalCulture,data=data
```

多元回归可始化

```
library(plotly)
```

Attaching package: 'plotly'

The following object is masked from 'package:ggplot2':

last_plot

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

rename, style

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

select

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

filter

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

layout

```
library(reshape2)
library(tidyverse)
```

— Attaching core tidyverse packages — tidyverse 2.0.0 —

✓ dplyr	1.1.4	✓ readr	2.1.5
✓ forcats	1.0.0	✓ stringr	1.5.1
✓ lubridate	1.9.3	✓ tibble	3.2.1
✓ purrr	1.0.2	✓ tidyr	1.3.1

— Conflicts — tidyverse_conflicts() —

* purrr::%@%()	masks memisc::%@%()
* ggplot2::%+%()	masks psych::%+%()
* ggplot2::alpha()	masks psych::alpha()
* lubridate::as.interval()	masks memisc::as.interval()
* dplyr::collect()	masks memisc::collect()
* dplyr::filter()	masks plotly::filter(), stats::filter()
* lubridate::is.interval()	masks memisc::is.interval()
* dplyr::lag()	masks stats::lag()
* dplyr::recode()	masks car::recode(), memisc::recode()
* dplyr::rename()	masks plotly::rename(), memisc::rename()
* dplyr::select()	masks plotly::select(), MASS::select()
* purrr::some()	masks car::some()
* dplyr::syms()	masks ggplot2::syms(), memisc::syms()
* tibble::view()	masks memisc::view()

i Use the conflicted package (<<http://conflicted.r-lib.org/>>) to force all conflicts to become errors

```
library(tidymodels)
```

Registered S3 method overwritten by 'parsnip':

method from
autoplot.glmnet ggfortify

— Attaching packages — tidymodels 1.1.1 —

✓ broom	1.0.5	✓ rsample	1.2.0
✓ dials	1.2.1	✓ tune	1.1.2
✓ infer	1.0.6	✓ workflows	1.1.4
✓ modeldata	1.3.0	✓ workflowsets	1.0.1
✓ parsnip	1.2.0	✓ yardstick	1.3.0
✓ recipes	1.0.10		

— Conflicts ————— tidymodels_conflicts() —

```

✖ purrr::%@%()      masks memisc::%@%()
✖ ggplot2::%+%()    masks psych::%+%()
✖ scales::alpha()   masks ggplot2::alpha(), psych::alpha()
✖ dplyr::collect()  masks memisc::collect()
✖ scales::discard() masks purrr::discard()
✖ dplyr::filter()   masks plotly::filter(), stats::filter()
✖ recipes::fixed()  masks stringr::fixed()
✖ dplyr::lag()       masks stats::lag()
✖ dplyr::recode()    masks car::recode(), memisc::recode()
✖ dplyr::rename()    masks plotly::rename(), memisc::rename()
✖ dplyr::select()    masks plotly::select(), MASS::select()
✖ purrr::some()      masks car::some()
✖ yardstick::spec() masks readr::spec()
✖ recipes::step()    masks stats::step()
✖ dplyr::syms()       masks ggplot2::syms(), memisc::syms()
✖ tibble::view()     masks memisc::view()
• Use tidymodels_prefer() to resolve common conflicts.

```

```

library(plotly)
#install.packages("kernlab")
library(kernlab)

```

Attaching package: 'kernlab'

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

alpha

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

cross

The following object is masked from 'package:ggplot2':

alpha

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

alpha

```

#install.packages("pracma")
library(pracma) #为了在曲面上显示网格线

```

Attaching package: 'pracma'

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

cross, eig, size

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

cross

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

logit

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

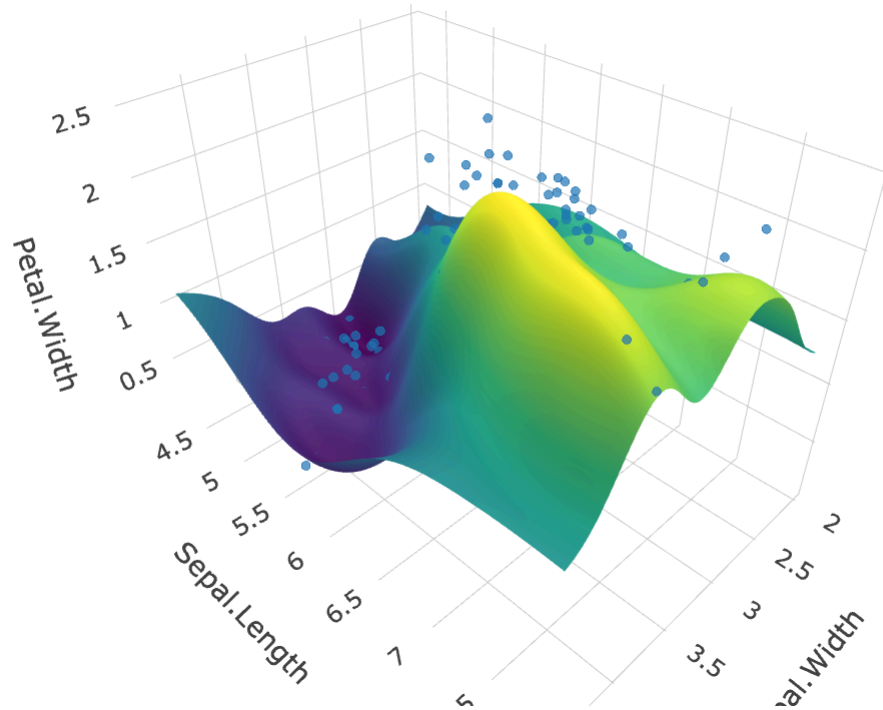
logit, polar

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

Reshape

```
data(iris)
#选择自变量和因变量
mesh_size <- .02
margin <- 0
X <- iris %>% select(Sepal.Width, Sepal.Length)
y <- iris %>% select(Petal.Width)
model <- svm_rbf(cost = 1.0) %>%
  set_engine("kernlab") %>%
  set_mode("regression") %>%
  fit(Petal.Width ~ Sepal.Width + Sepal.Length, data = iris)
x_min <- min(X$Sepal.Width) - margin
x_max <- max(X$Sepal.Width) - margin
y_min <- min(X$Sepal.Length) - margin
y_max <- max(X$Sepal.Length) - margin
xrange <- seq(x_min, x_max, mesh_size)
yrange <- seq(y_min, y_max, mesh_size)
xy <- meshgrid(x = xrange, y = yrange)
xx <- xy$X
yy <- xy$Y
dim_val <- dim(xx)
xx1 <- matrix(xx, length(xx), 1)
yy1 <- matrix(yy, length(yy), 1)
final <- cbind(xx1, yy1)
pred <- model %>%
  predict(final)

pred <- pred$.pred
pred <- matrix(pred, dim_val[1], dim_val[2])
fig <- plot_ly(iris, x = ~Sepal.Width, y = ~Sepal.Length, z = ~Petal.Width ) %>%
  add_markers(size = 5) %>%
  add_surface(x=xrange, y=yrange, z=pred, alpha = 0.65, type = 'mesh3d', name = 'pred_
fig
```

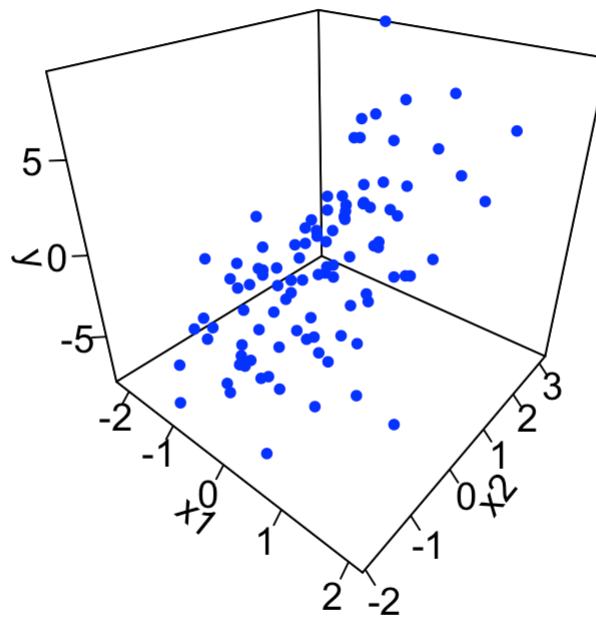



```
library(plot3D)
set.seed(123)
n <- 100
x1 <- rnorm(n)
x2 <- rnorm(n)
y <- 2*x1 + 3*x2 + rnorm(n)

data <- data.frame(x1, x2, y)
model <- lm(y ~ x1 + x2, data = data)
# 创建一个网格
grid_x1 <- seq(min(data$x1), max(data$x1), length.out = 50)
grid_x2 <- seq(min(data$x2), max(data$x2), length.out = 50)
grid <- expand.grid(x1 = grid_x1, x2 = grid_x2)

# 预测网格上的y值
grid$y_pred <- predict(model, newdata = grid)

# 绘制三维散点图
scatter3D(data$x1, data$x2, data$y, pch = 20, colvar = NULL, col = "blue",
          xlab = "x1", ylab = "x2", zlab = "y", theta = 40, phi = 30,
          ticktype = "detailed", cex.lab = 1.2, cex.axis = 1.2)
```



```
# 绘制三维拟合曲面
#surf3D(grid_x1, grid_x2, matrix(grid$y_pred, nrow = length(grid_x2)), add = TRUE, c

library(plot3D)

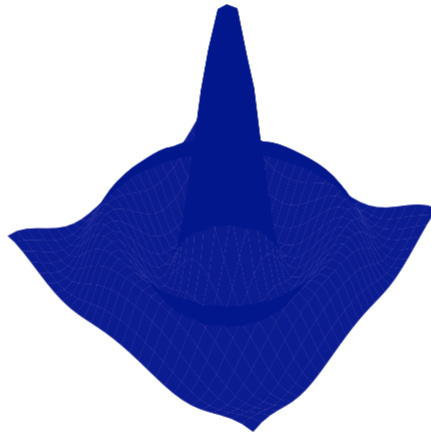
# 创建网格
x <- seq(-10, 10, length.out = 30)
y <- seq(-10, 10, length.out = 30)

# 创建网格矩阵
x_grid <- outer(x, y, Vectorize(function(x, y) x))
y_grid <- outer(x, y, Vectorize(function(x, y) y))

# 计算z值 (这里使用一个简单的函数作为例子)
z_grid <- outer(x, y, function(x, y) sin(sqrt(x^2 + y^2)) / sqrt(x^2 + y^2))

# 使用surf3D绘制3D曲面图
surf3D(x = x_grid, y = y_grid, z = z_grid, colvar = NULL,
       color = "green", alpha = 0.8, xlab = "X", ylab = "Y", zlab = "Z",
       main = "3D Surface Plot")
```

3D Surface Plot



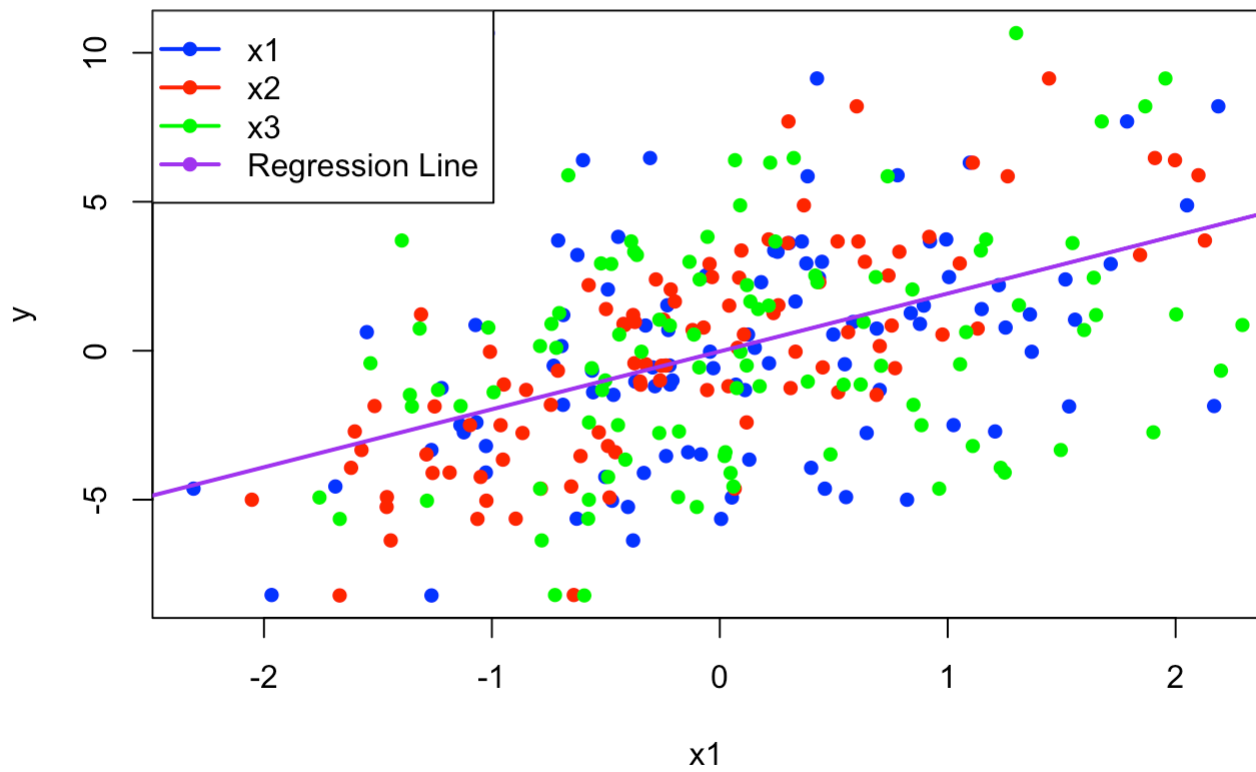
```
# 创建示例数据
set.seed(123)
x1 <- rnorm(100)
x2 <- rnorm(100)
x3 <- rnorm(100)
y <- 2*x1 + 3*x2 + 1.5*x3 + rnorm(100)

# 拟合多元回归模型
model <- lm(y ~ x1 + x2 + x3)

# 绘制多元回归模型
plot(y ~ x1, col="blue", pch=16, xlab="x1", ylab="y")
points(x2, y, col="red", pch=16)
points(x3, y, col="green", pch=16)
abline(model, col="purple", lwd=2)
```

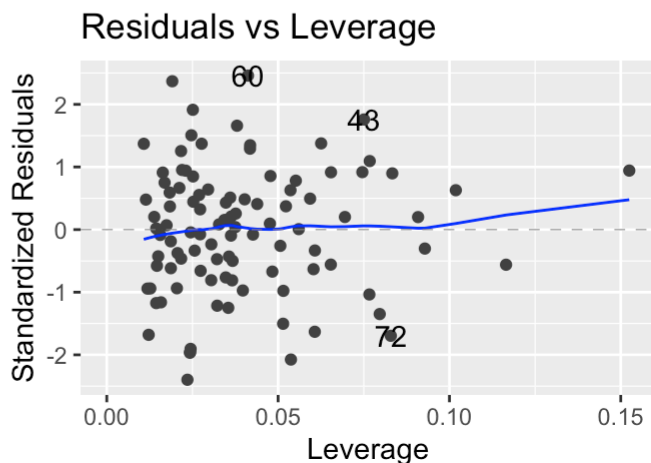
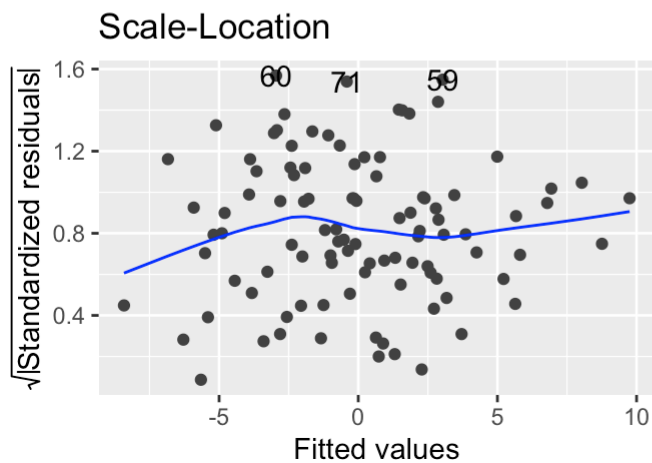
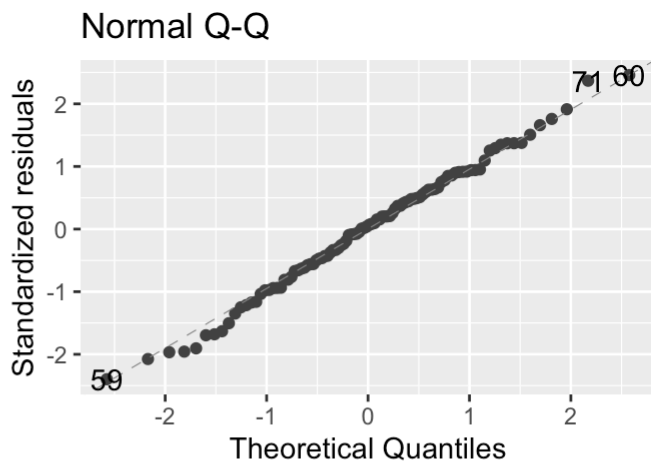
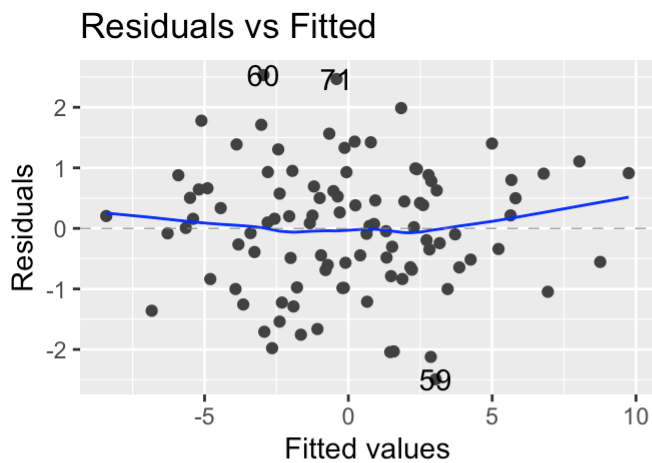
Warning in abline(model, col = "purple", lwd = 2): only using the first two of 4 regression coefficients

```
legend("topleft", legend=c("x1", "x2", "x3", "Regression Line"),
      col=c("blue", "red", "green", "purple"), pch=16, lwd=2)
```

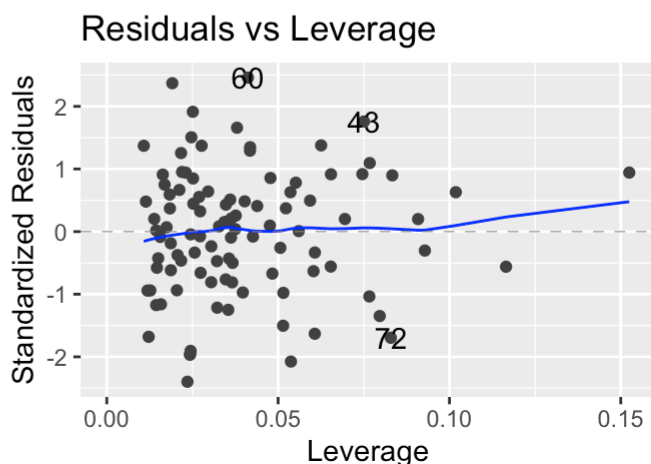
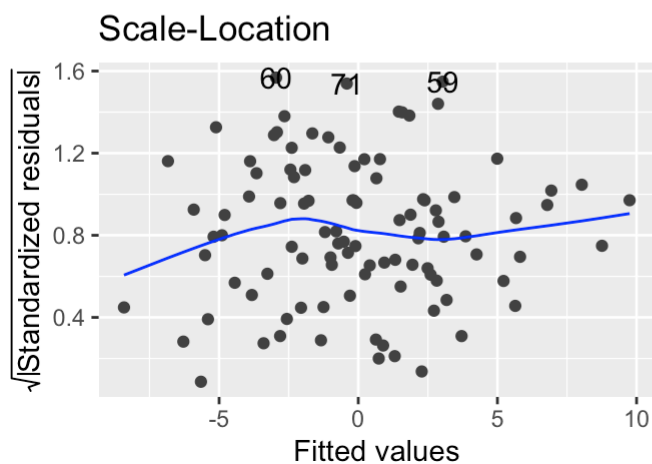
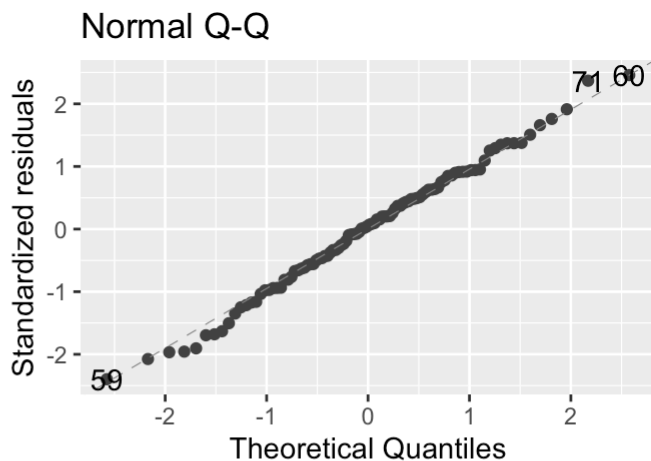
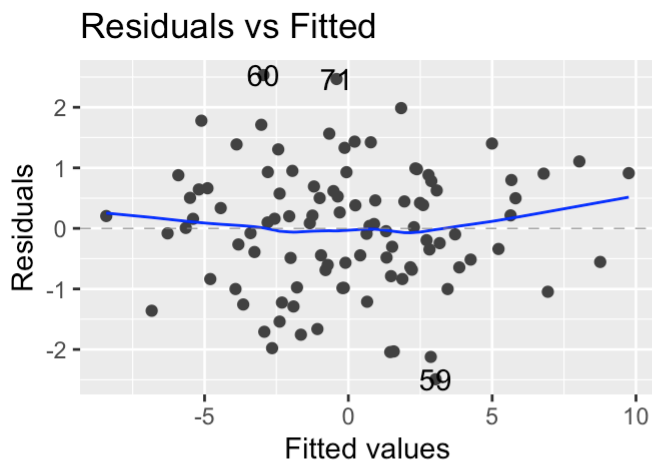


```
library(ggfortify)

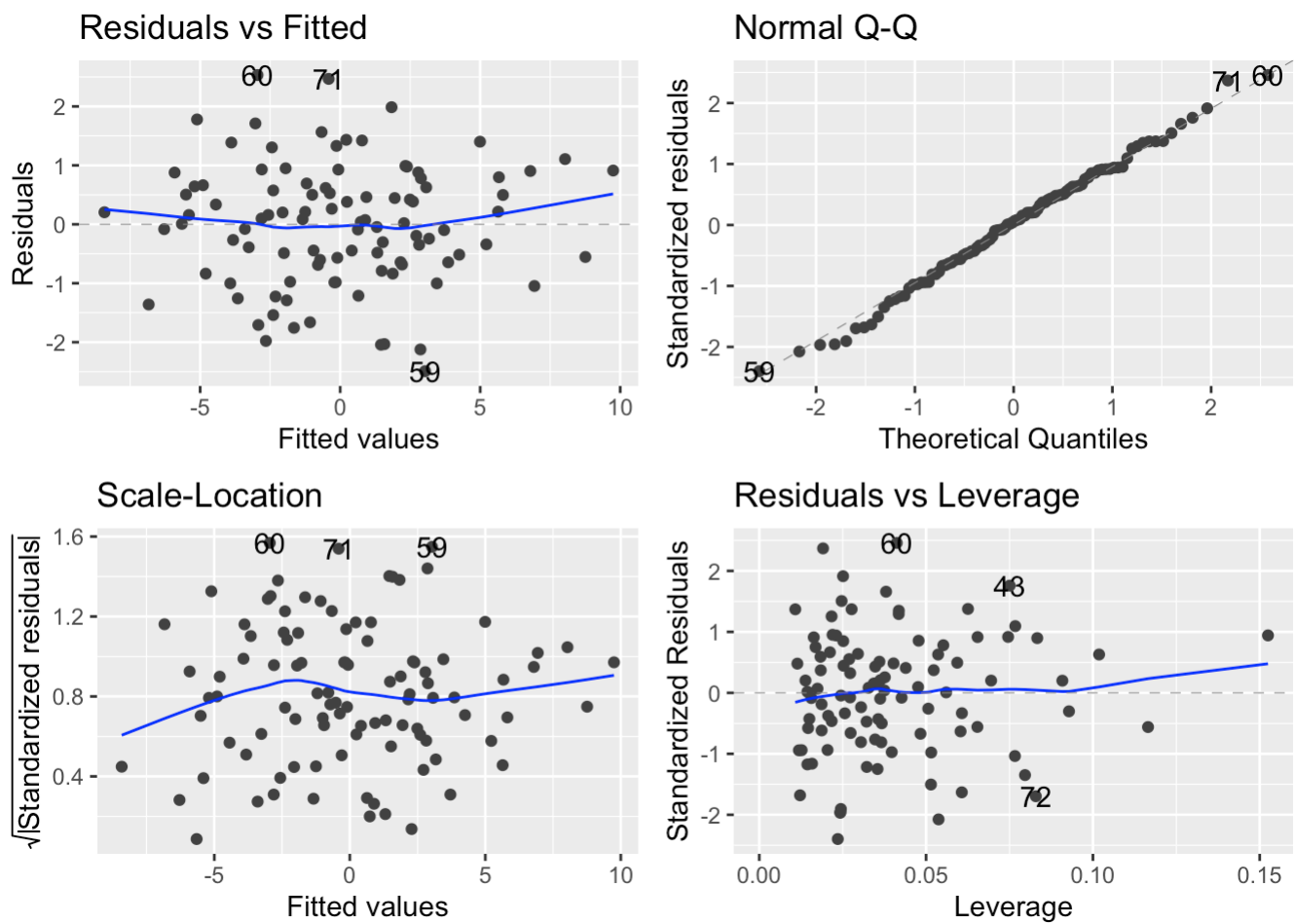
autoplot(model)
```



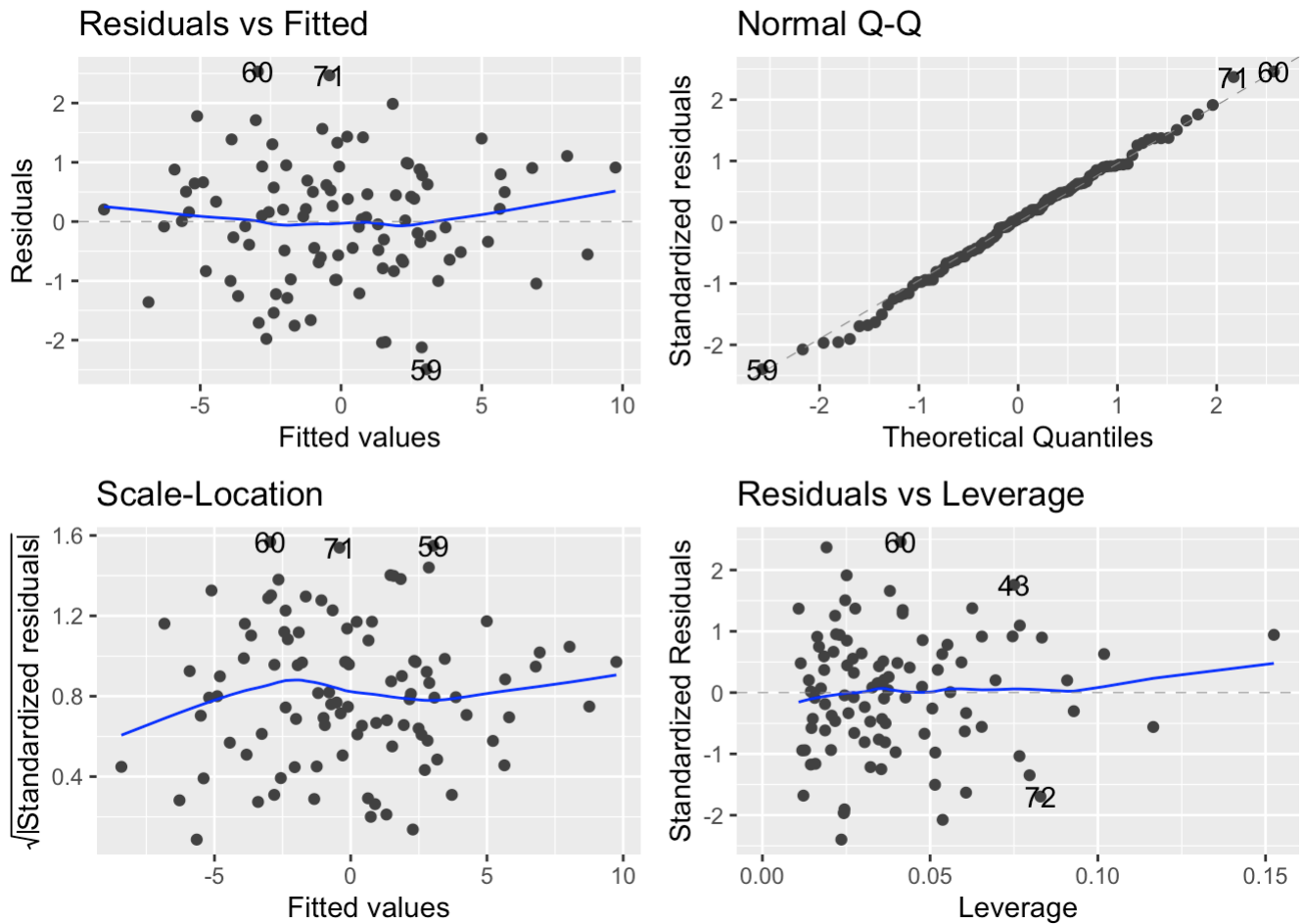
```
autoplot(model, type = "resid")
```



```
autoplot(model, type = "fit")
```



```
autoplot(model, type = "conf")
```



中介效应分析

调节分析

共线性诊断不通过

取标准化值 ## SEM <https://lavaan.ugent.be/tutorial/> $a \rightarrow b \rightarrow c$ $a \rightarrow c$ b 为中介变量, 中介效应: $a \rightarrow b$ 的系数 * $b \rightarrow c$ 的系数 总效应 中介效应 + $a \rightarrow c$ 的系数

```
library(lavaan)
```

This is lavaan 0.6-17
lavaan is FREE software! Please report any bugs.

Attaching package: 'lavaan'

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

```
cor2cov
```

```
library(semPlot)
#SEM
model <- '
```

```

# 潜变量 =~ 测量指标1(既量表) + 测量指标2 + ...
teleworkerCharacteristics =~ teleworkerCharacteristics1 + teleworkerCharacteristics2 +
communication =~ communication1 + communication2 + communication3
management =~ management2 + management3 + management4
organisationalCulture =~ organisationalCulture1 + organisationalCulture2 + organisatio
#因变量
job_effectiveness =~ job_effectiveness1 + job_effectiveness2 + job_effectiveness3 + jo
work.life_balance =~ work.life_balance1 + work.life_balance2 + work.life_balance3 + wo
well.being =~ well.being1 + well.being2 + well.being3 + well.being4
#回归方程
# 因变量~ 自变量1+自变量2+...
#中介
work.life_balance ~ beta_work_tel*teleworkerCharacteristics + beta_work_com*communicat
well.being ~ beta_well_tel*teleworkerCharacteristics + beta_well_com*communication + b
#直接
job_effectiveness ~ beta_job_tel*teleworkerCharacteristics + beta_job_com*communicatio
#中介效应
indirect_work_job_tel:=beta_work_tel*beta_job_tel
indirect_work_job_com:=beta_work_com*beta_job_com
indirect_work_job_man:=beta_work_man*beta_job_man
indirect_work_job_org:=beta_work_org*beta_job_org

indirect_well_job_tel:=beta_well_tel*beta_job_tel
indirect_well_job_com:=beta_well_com*beta_job_com
indirect_well_job_man:=beta_well_man*beta_job_man
indirect_well_job_org:=beta_well_org*beta_job_org
#整体效应
all:=indirect_work_job_tel+indirect_work_job_com+indirect_work_job_man+indirect_work_j
'
result <- sem(model,data=data_src)

```

Warning in lav_object_post_check(object): lavaan WARNING: some estimated lv
variances are negative

```

#summary(result,standardized = TRUE)
summary(result,standardized=TRUE, fit.measures=TRUE) #后面画图后,显示不全

```

lavaan 0.6.17 ended normally after 140 iterations

Estimator	ML
Optimization method	NLMINB
Number of model parameters	70
Number of observations	93

Model Test User Model:

Test statistic	1235.326
Degrees of freedom	255
P-value (Chi-square)	0.000

Model Test Baseline Model:

Test statistic	4436.925
Degrees of freedom	300
P-value	0.000

User Model versus Baseline Model:

Comparative Fit Index (CFI)	0.763
Tucker-Lewis Index (TLI)	0.721

Loglikelihood and Information Criteria:

Loglikelihood user model (H0)	-1829.675
Loglikelihood unrestricted model (H1)	-1212.012
Akaike (AIC)	3799.350
Bayesian (BIC)	3976.632
Sample-size adjusted Bayesian (SABIC)	3755.658

Root Mean Square Error of Approximation:

RMSEA	0.203
90 Percent confidence interval - lower	0.192
90 Percent confidence interval - upper	0.215
P-value H ₀ : RMSEA ≤ 0.050	0.000
P-value H ₀ : RMSEA ≥ 0.080	1.000

Standardized Root Mean Square Residual:

SRMR	0.085
------	-------

Parameter Estimates:

Standard errors	Standard
Information	Expected
Information saturated (h1) model	Structured

Latent Variables:

	Estimate	Std.Err	z-value	P(> z)	Std.lv
teleworkerCharacteristics =~					
tlwrkrChrctrs1	1.000				0.781
tlwrkrChrctrs2	0.929	0.127	7.314	0.000	0.726
tlwrkrChrctrs3	1.271	0.141	8.993	0.000	0.993
communication =~					
communication1	1.000				0.910
communication2	1.026	0.081	12.669	0.000	0.934
communication3	1.210	0.108	11.189	0.000	1.101
management =~					
management2	1.000				1.023
management3	1.220	0.086	14.227	0.000	1.248
management4	0.920	0.091	10.097	0.000	0.941
organisationalCulture =~					
organstnlCltr1	1.000				1.336
organstnlCltr2	1.036	0.106	9.786	0.000	1.384
organstnlCltr3	1.076	0.084	12.745	0.000	1.437

organstnlCltr4	0.931	0.088	10.529	0.000	1.243
job_effectiveness =~					
job_effctvnss1	1.000				0.785
job_effctvnss2	1.087	0.067	16.321	0.000	0.853
job_effctvnss3	1.143	0.059	19.290	0.000	0.897
job_effctvnss4	0.955	0.058	16.425	0.000	0.749
work.life_balance =~					
work.lif_blnc1	1.000				0.682
work.lif_blnc2	1.213	0.062	19.535	0.000	0.828
work.lif_blnc3	1.191	0.064	18.484	0.000	0.813
work.lif_blnc4	1.099	0.067	16.368	0.000	0.750
well.being =~					
well.being1	1.000				0.823
well.being2	0.971	0.061	15.820	0.000	0.800
well.being3	1.362	0.078	17.411	0.000	1.121
well.being4	1.140	0.072	15.919	0.000	0.939
Std.all					

0.729
0.723
0.876

0.833
0.973
0.889

0.869
0.946
0.801

0.804
0.840
0.997
0.882

0.931
0.916
0.956
0.918

0.931
0.960
0.947
0.917

0.920
0.913
0.940
0.915

Regressions:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
work.life_balance ~						
tlc (bt_wrk_t)	0.219	0.059	3.740	0.000	0.251	0.251

cmm (bt_wrk_c)	0.014	0.032	0.434	0.664	0.019	0.019
mng (bt_wrk_m)	0.440	0.053	8.321	0.000	0.659	0.659
orC (bt_wrk_r)	0.084	0.024	3.579	0.000	0.165	0.165
well.being ~						
tlC (bt_wll_t)	0.398	0.067	5.987	0.000	0.378	0.378
cmm (bt_wll_c)	0.101	0.029	3.456	0.001	0.112	0.112
mng (bt_wll_m)	0.263	0.043	6.065	0.000	0.327	0.327
orC (bt_wll_r)	0.223	0.028	7.946	0.000	0.361	0.361
job_effectiveness ~						
tlC (bt_jb_t)	0.605	0.273	2.217	0.027	0.603	0.603
cmm (bt_jb_c)	0.092	0.046	1.975	0.048	0.106	0.106
mng (bt_jb_m)	-0.546	0.620	-0.881	0.379	-0.712	-0.712
orC (bt_jb_r)	-0.146	0.106	-1.383	0.167	-0.249	-0.249
w._ (bt_jb_wr)	0.936	1.453	0.644	0.520	0.814	0.814
wl. (bt_jb_wl)	0.378	0.279	1.353	0.176	0.396	0.396

Covariances:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	
teleworkerCharacteristics ~~						
communication	0.336	0.096	3.489	0.000	0.473	
management	0.613	0.129	4.768	0.000	0.767	
organistnlCltr	0.653	0.155	4.200	0.000	0.626	
communication ~~						
management	0.561	0.127	4.405	0.000	0.603	
organistnlCltr	0.704	0.165	4.256	0.000	0.579	
management ~~						
organistnlCltr	0.854	0.190	4.503	0.000	0.625	

Std.all

0.473

0.767

0.626

0.603

0.579

0.625

Variances:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
.tlwrkrChrctrs1	0.537	0.079	6.767	0.000	0.537	0.468
.tlwrkrChrctrs2	0.482	0.071	6.778	0.000	0.482	0.477
.tlwrkrChrctrs3	0.300	0.052	5.802	0.000	0.300	0.233
.communication1	0.364	0.061	5.974	0.000	0.364	0.305
.communication2	0.049	0.029	1.710	0.087	0.049	0.054
.communication3	0.323	0.062	5.185	0.000	0.323	0.210
.management2	0.339	0.055	6.162	0.000	0.339	0.244
.management3	0.184	0.040	4.629	0.000	0.184	0.106
.management4	0.496	0.077	6.456	0.000	0.496	0.359
.organstnlCltr1	0.978	0.146	6.677	0.000	0.978	0.354
.organstnlCltr2	0.799	0.121	6.607	0.000	0.799	0.294
.organstnlCltr3	0.014	0.028	0.508	0.611	0.014	0.007
.organstnlCltr4	0.441	0.069	6.439	0.000	0.441	0.222
.job_effctvnss1	0.094	0.015	6.155	0.000	0.094	0.132

.job_effctvnss2	0.139	0.022	6.305	0.000	0.139	0.161
.job_effctvnss3	0.075	0.013	5.627	0.000	0.075	0.085
.job_effctvnss4	0.105	0.017	6.292	0.000	0.105	0.158
.work.lif_blnc1	0.072	0.012	6.192	0.000	0.072	0.133
.work.lif_blnc2	0.059	0.011	5.612	0.000	0.059	0.079
.work.lif_blnc3	0.075	0.013	5.948	0.000	0.075	0.103
.work.lif_blnc4	0.106	0.017	6.316	0.000	0.106	0.158
.well.being1	0.124	0.017	7.341	0.000	0.124	0.154
.well.being2	0.127	0.017	7.342	0.000	0.127	0.166
.well.being3	0.166	0.023	7.294	0.000	0.166	0.117
.well.being4	0.171	0.023	7.342	0.000	0.171	0.163
tlwrkrChrctrst	0.610	0.149	4.094	0.000	1.000	1.000
communication	0.829	0.169	4.899	0.000	1.000	1.000
management	1.047	0.198	5.276	0.000	1.000	1.000
organistnlCltr	1.785	0.380	4.695	0.000	1.000	1.000
.job_effectvnss	-0.036	0.012	-2.979	0.003	-0.058	-0.058
.work.life_blnc	0.005	0.005	0.993	0.321	0.010	0.010
.well.being	-0.022	0.005	-4.637	0.000	-0.033	-0.033

Defined Parameters:

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
indrct_wrk_jb_	0.133	0.084	1.572	0.116	0.151	0.151
indrct_wrk_jb_	0.001	0.003	0.453	0.650	0.002	0.002
indrct_wrk_jb_	-0.240	0.289	-0.831	0.406	-0.470	-0.470
indrct_wrk_jb_	-0.012	0.010	-1.215	0.224	-0.041	-0.041
indrct_wll_jb_	0.241	0.129	1.871	0.061	0.228	0.228
indrct_wll_jb_	0.009	0.006	1.680	0.093	0.012	0.012
indrct_wll_jb_	-0.144	0.172	-0.838	0.402	-0.233	-0.233
indrct_wll_jb_	-0.033	0.024	-1.353	0.176	-0.090	-0.090
all	1.269	0.652	1.946	0.052	0.769	0.769

```
#chisq_result = chisq.test(result)
#summary(chisq_result)
#获取模型拟合参数
fits = fitMeasures(result)
fits
```

npar	fmin	chisq
70.000	6.642	1235.326
df	pvalue	baseline.chisq
255.000	0.000	4436.925
baseline.df	baseline.pvalue	cfi
300.000	0.000	0.763
tli	nnfi	rfi
0.721	0.721	0.672
nfi	pnfi	ifi
0.722	0.613	0.766
rni	logl	unrestricted.logl
0.763	-1829.675	-1212.012
aic	bic	ntotal
3799.350	3976.632	93.000
bic2	rmsea	rmsea.ci.lower
3755.658	0.203	0.192

rmsea.ci.upper	rmsea.ci.level	rmsea.pvalue
0.215	0.900	0.000
rmsea.close.h0	rmsea.notclose.pvalue	rmsea.notclose.h0
0.050	1.000	0.080
rmr	rmr_nomean	srmr
0.127	0.127	0.085
srmr_bentler	srmr_bentler_nomean	crmr
0.085	0.085	0.088
crmr_nomean	srmr_mplus	srmr_mplus_nomean
0.088	0.085	0.085
cn_05	cn_01	gfi
23.077	24.372	0.503
agfi	pgfi	mfi
0.366	0.394	0.005
ecvi		
14.788		

```
summary(fits)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-1829.675	0.085	0.558	325.208	12.752	4436.925

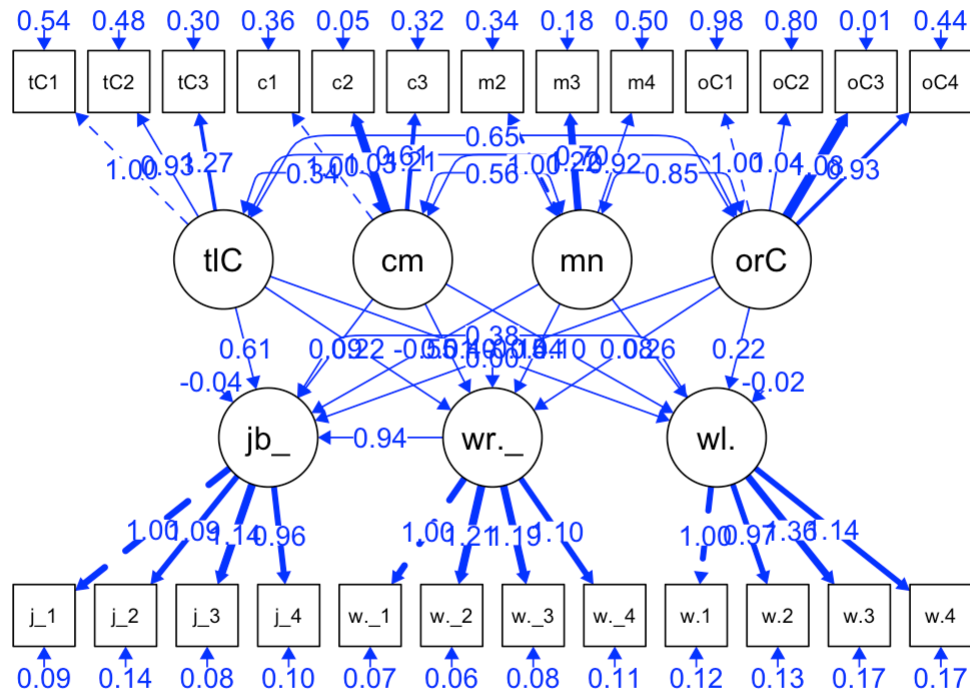
```
fits['rmsea']
```

```
rmsea
0.203317
```

```
#summary(result)
#模型概览
```

```
semPaths(result,
  what = "std.all", #"est",      # 显示估计的系数
  whatLabels = "est", # 标签显示估计的系数
  style="lisrel", #"lisrel"、"ram"、"dot"、"jgraph"等。默认为"lisrel"。
  intervals=T,
  node.color = "blue",
  estimlegend.cex = 1.5, # 系数字体大小
  edge.color = "blue",
  edge.label.cex = 1, # 调整标签大小
  fade = F,          # 不使用渐变效果
  layout="tree",
  rotation=1,
  nCharNodes = 0.05,
  residuals=T,
  curvePivot = TRUE) # 曲线在变量处弯曲
```

Warning in qqgraph::qqgraph(Edgelist, labels = nLab, bidirectional = Bidir, : The following arguments are not documented and likely not arguments of qqgraph and thus ignored: intervals; node.color; estimlegend.cex



```
# semPaths(fit, whatLabels="est.std", style="lisrel", edge.label.cex=1, layout="tree2")
```

```
model <- '
# 潜变量 =~ 测量指标1(既量表) + 测量指标2 + ...
environment =~ environment0 + environment1 + environment2 + environment3
jobCharacteristics =~ jobCharacteristics0 + jobCharacteristics1 + jobCharacteristics2
teleworkerCharacteristics =~ teleworkerCharacteristics0 + teleworkerCharacteristics1 +
communication =~ communication0 + communication1 + communication2 + communication3 + c
management =~ management0 + management1 + management2 + management3 + management4
organisationalCulture =~ organisationalCulture0 + organisationalCulture1 + organisatio
technology =~ technology0 + technology1 + technology2 + technology3 + technology4
asynchronousWork =~ asynchronousWork0 + asynchronousWork1 + asynchronousWork2 + asynch
#因变量
job_effectiveness =~ job_effectiveness1 + job_effectiveness2 + job_effectiveness3 + jo
work.life_balance =~ work.life_balance1 + work.life_balance2 + work.life_balance3 + wo
well.being =~ well.being1 + well.being2 + well.being3 + well.being4
#回归方程
# 因变量~ 自变量1+自变量2+...
#job_effectiveness ~ environment + jobCharacteristics + teleworkerCharacteristics + co
#work.life_balance ~ environment + jobCharacteristics + teleworkerCharacteristics + co
well.being ~ environment + jobCharacteristics + teleworkerCharacteristics + communicat
,
,
```

model: SEM模型的拟合对象, 通常是由sem()函数拟合后得到的对象。

what: 指定要显示的内容, 可以是"std" (标准化估计值)、"std.lv" (标准化潜变量)、"std.all" (标准化估计值)。

style: 图形风格, 可以是"lisrel"、"ram"、"dot"、"jgraph"等。默认为"lisrel"。

residuals: 是否显示残差。默认为TRUE。

intervals: 是否显示参数估计的置信区间。默认为FALSE。
 whatLabels: 是否显示节点标签。默认为FALSE。
 layout: 图形的布局。默认为circular (圆形布局)。
 rotation: 图形的旋转角度。默认为0。
 edge.label.cex: 边标签的大小。默认为1。
 edge.label.offset: 边标签的偏移。默认为0。
 edge.label: 是否显示边标签。默认为TRUE。
 edge.color: 边的颜色。默认为黑色。
 edge.width: 边的宽度。默认为1。
 edge.lwd: 边的线宽。默认为1。
 edge.curved: 边的曲率。默认为FALSE。
 edge.lty: 边的线型。默认为1。
 label.prop: 是否根据参数大小调整节点标签的大小。默认为FALSE。
 label.cex: 节点标签的大小。默认为1。
 label.offset: 节点标签的偏移。默认为0.5。
 node.color: 节点的颜色。默认为黑色。
 node.width: 节点的宽度。默认为0.3。
 node.size: 节点的大小。默认为2。
 curvePivot: 弯曲箭头的位置。默认为0.5。
 curveAngle: 弯曲箭头的角度。默认为60。
 curveArrowSize: 弯曲箭头的大小。默认为0.5。
 ,

[1] "\nmodel: SEM模型的拟合对象, 通常是由sem()函数拟合后得到的对象.\nwhat: 指定要显示的内容, 可以是\"std\" (标准化估计值)、\"std.lv\" (标准化潜变量)、\"std.all\" (标准化估计值和标准化潜变量)、\"std.nox\" (标准化估计值但不包括残差)等。默认值为\"std\".\nstyle: 图形风格, 可以是\"lisrel\"、\"ram\"、\"dot\"、\"jgraph\"等。默认为\"lisrel\".\nresiduals: 是否显示残差。默认为TRUE.\nintervals: 是否显示参数估计的置信区间。默认为FALSE.\nwhatLabels: 是否显示节点标签。默认为FALSE.\nlayout: 图形的布局。默认为circular (圆形布局)。\nrotation: 图形的旋转角度。默认为0.\nedge.label.cex: 边标签的大小。默认为1.\nedge.label.offset: 边标签的偏移。默认为0.\nedge.label: 是否显示边标签。默认为TRUE.\nedge.color: 边的颜色。默认为黑色.\nedge.width: 边的宽度。默认为1.\nedge.lwd: 边的线宽。默认为1.\nedge.curved: 边的曲率。默认为FALSE.\nedge.lty: 边的线型。默认为1.\nlabel.prop: 是否根据参数大小调整节点标签的大小。默认为FALSE.\nlabel.cex: 节点标签的大小。默认为1.\nlabel.offset: 节点标签的偏移。默认为0.5.\nnode.color: 节点的颜色。默认为黑色.\nnode.width: 节点的宽度。默认为0.3.\nnode.size: 节点的大小。默认为2.\ncurvePivot: 弯曲箭头的位置。默认为0.5.\ncurveAngle: 弯曲箭头的角度。默认为60.\ncurveArrowSize: 弯曲箭头的大小。默认为0.5.\n"

```

library(lavaan)

set.seed(123) # 设置随机种子以便结果可复现
n <- 100 # 样本量

# 创建潜在变量
eta1 <- rnorm(n) # 潜在变量1
eta2 <- rnorm(n) # 潜在变量2

# 创建观测变量
x1 <- eta1 + rnorm(n) # x1是eta1的指标
x2 <- eta1 + rnorm(n) # x2也是eta1的指标
y <- eta2 + rnorm(n) # y是eta2的指标

# 将数据整合到数据框中

```

```
data <- data.frame(x1, x2, y)

# 定义SEM模型
model <- '
  # 测量模型
  eta1 =~ x1 + x2
  eta2 =~ y
'

#在这个模型中, eta1 =~ x1 + x2表示eta1是由x1和x2测量的潜在变量, 而eta2 =~ y表示eta2是由y测量的潜

# 拟合模型
fit <- sem(model, data = data)

# 显示摘要信息
summary(fit)
```

lavaan 0.6.17 ended normally after 23 iterations

Estimator	ML
Optimization method	NLMINB
Number of model parameters	6
Number of observations	100

Model Test User Model:

Test statistic	0.000
Degrees of freedom	0

Parameter Estimates:

Standard errors	Standard
Information	Expected
Information saturated (h1) model	Structured

Latent Variables:

	Estimate	Std.Err	z-value	P(> z)
eta1 =~				
x1	1.000			
x2	0.908	0.840	1.082	0.279
eta2 =~				
y	1.000			

Covariances:

	Estimate	Std.Err	z-value	P(> z)
eta1 ~~				
eta2	-0.211	0.158	-1.331	0.183

Variances:

	Estimate	Std.Err	z-value	P(> z)
.x1	0.804	0.647	1.244	0.214
.x2	1.239	0.554	2.237	0.025
.y	0.000			

eta1	0.692	0.661	1.047	0.295
eta2	1.647	0.233	7.071	0.000

多层SEM模型

ESEM 探索性结构方程模型， Exploratory Structural Equation Modeling

是一种相对较新的方法，用于探索性的数据分析。这种方法结合了探索性因素分析（EFA）和验证性因素分析（CFA）的思想，同时融合了结构方程模型（SEM）的灵活性，旨在同时实现理论的探索与验证。

CFA

##其它 ###图形 <https://plotly.com/r/>