R for bioinformatics, data summarisation and statistics

HUST Bioinformatics course series

Wei-Hua Chen (CC BY-NC 4.0)

25 September, 2019

section 1: TOC

前情提要

- basic plot functions
- basic ggplot2
- special letters
- equations
- advanced ggplot2

本次提要

- data summarisation functions (vector data)
 - median, mean, sd, quantile, summary
- 图形化的 data summarisation (two-D data/ tibble/ table)
 - dot plot
 - smooth
 - linear regression
 - correlation & variance explained
 - groupping & bar/ box/ plots
- statistics
 - parametric tests
 - t-test
 - one way ANNOVA
 - two way ANNOVA
 - linear regression
 - model / prediction / coefficients
 - non-parametric comparison



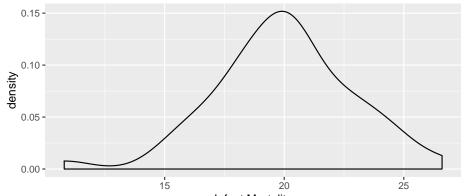
section 2: vector summarisation

vector data

distribution

```
library(tidyverse);
ggplot( swiss, aes( x = Infant.Mortality ) ) + geom_density() +
    ggtitle("Swiss Fertility and Socioeconomic Indicators (1888) Data")
```

Swiss Fertility and Socioeconomic Indicators (1888) Data



describe normal distributions

可以用 mean 和 sd 来描述

- It's symmetrical.
- Mean and median are the same.
- Most common values are near the mean; less common values are farther from it.
- Standard deviation marks the distance from the mean to the inflection point.

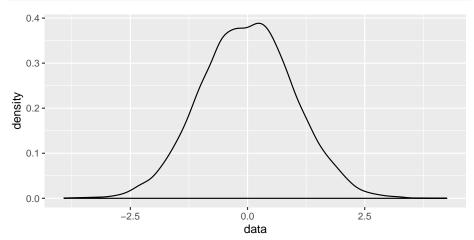
$$(mean + 1 * sd) >= 68%$$

 $(mean + 2 * sd) >= 95%$ 的数据



functions to generate random normal distrubions

```
# 生成 10000 个随机数字, 使其 mean = 0, sd = 1, 且为 normal distribution ...
x <- rnorm(10000, mean = 0, sd = 1);
ggplot( data.frame( data = x ), aes( data ) ) + geom_density( );
```



More to read: http://uc-r.github.io/generating_random_numbers/

other functions to generate random normal distributions

注意,以下函数中的 q, p, x 需要自行提供

```
# generate CDF probabilities for value(s) in vector q
pnorm(q, mean = 0, sd = 1)

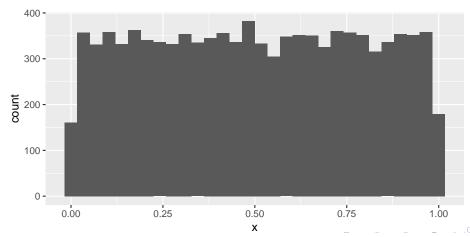
# generate quantile for probabilities in vector p
qnorm(p, mean = 0, sd = 1)

# generate density function probabilites for value(s) in vector x
dnorm(x, mean = 0, sd = 1)
```

其它规律的 distributions

uniform distributions

```
x <- runif( 10000 ); ## random numbers of uniform distributions between 0 and 1
ggplot( data.frame( dat = x ), aes( x ) ) + geom_histogram();</pre>
```



uniform distribution 的各种函数

注: 以下函数中的 n 需要自行决定

```
# generate n random numbers between 0 and 25
runif(n, min = 0, max = 25)

# generate n random numbers between 0 and 25 (with replacement)
sample(0:25, n, replace = TRUE)

# generate n random numbers between 0 and 25 (without replacement)
sample(0:25, n, replace = FALSE)
```

other distributions, cont.

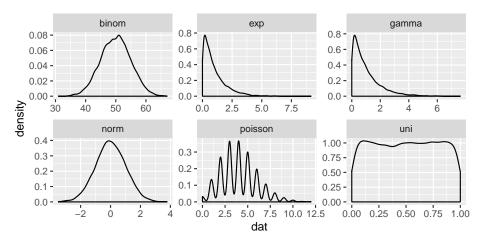
```
n <- 10000;
uni <- tibble( dat = runif(n), type = "uni" );
norm <- tibble( dat = rnorm(n), type = "norm" );
binom <- tibble( dat = rbinom(n, size = 100, prob = 0.5), type = "binom" );
poisson <- tibble( dat = rpois(n, lambda = 4), type = "poisson" );
exp <- tibble( dat = rexp(n, rate = 1) , type = "exp");
gamma <- tibble( dat = rgamma(n, shape = 1) , type = "gamma");

combined <- bind_rows( uni, norm, binom, poisson, exp, gamma );

plot1 <-
ggplot( combined , aes( dat ) ) + geom_density() +
facet_wrap( *type, ncol = 3, scales = "free");</pre>
```

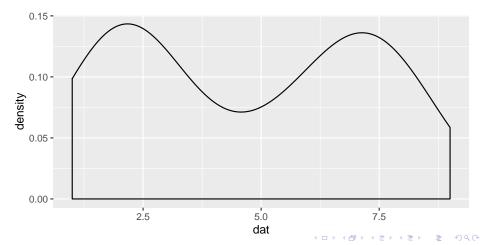
other distributions, plot

plot1;



non-parametric distribution

```
## votes on people's desire to visit
bi <- c(7, 3, 2, 1, 7, 3, 4, 5, 7, 6, 2, 2, 1, 3, 7, 2, 6, 8, 2, 7, 2, 2, 1,
3, 5, 8, 2, 6, 7, 8, 6, 2, 8, 7, 9, 2, 7, 5, 1, 8, 8, 2, 3, 7, 3, 8);
ggplot( data.frame( dat = bi ), aes(dat)) + geom_density();
```



量化描述数据

使用以下同名函数

mean: aka average, is the sum of all of the numbers in the data set divided by the size of the data set.

median: The median is the value that is in the middle when the numbers in a data set are sorted in increasing order.

sd: standard deviation

var: measures how far a set of numbers are spread out

range: 取值范围

note: from: https://www.ai-therapy.com/psychology-statistics/
descriptive/mean-mode-median

量化描述函数

```
mean( norm$dat );
median( norm$dat );
## mode( norm$dat ); ## ???

sd(norm$dat);
var(norm$dat);
range(norm$dat);
```

quantile and summary

```
quantile( norm$dat );
##
           0%
                      25% 50% 75%
                                                        100%
## -3.477057e+00 -6.630663e-01 7.636815e-05 6.881583e-01 3.777296e+00
## quantile 还接受其它参数
quantile( norm$dat, probs = seq(0, 1, length = 11));
                      10% 20%
                                             30%
                                                         40%
## -3.477057e+00 -1.290048e+00 -8.363559e-01 -5.200171e-01 -2.500945e-01
##
          50%
                      60%
                                 70%
                                             80%
## 7.636815e-05 2.546565e-01 5.338732e-01 8.579359e-01 1.312853e+00
##
         100%
## 3.777296e+00
## summaru ...
summary( norm$dat );
      Min. 1st Qu. Median Mean 3rd Qu. Max.
## -3.477057 -0.663066 0.000076 0.005601 0.688158 3.777296
## summaru 也可应用干非数值
summary( combined$type );
```

Class

Mode

Length

##

summary, cont.

```
## summary 可应用于整个表格; 相当于对每列进行 summary ...
summary( combined );
```

```
## dat type
## Min. :-3.4771 Length:60000
## 1st Qu.: 0.3157 Class :character
## Median : 0.9422 Mode :character
## Mean : 9.4104
## 3rd Qu.: 4.0000
## Max. :67.0000
```

table 函数

##

返回 vector 当中 unique 值和它们的出现次数

```
table( combined$type );
```

```
## binom exp gamma norm poisson uni
## 10000 10000 10000 10000 10000 10000
```

** 注 **: table 还接受 data.frame 作为输入,比如 table(combined)。请自行尝试并理解结果

section 3: two column data: part 1

数据介绍: a numeric vector and a facterial vector

此类数据,通常一列是数值,另一列是分组信息,如下例:

```
data.fig3a <- read_csv( file = "data/talk10/nc2015_data_for_fig3a.csv" );
head( data.fig3a[ c("tai", "trans.at") ] ); ## 只显示有用的两列</pre>
```

A tibble: 6 x 2

数据介绍,cont.

tai: 表达量的一种计算方式, 1 == lowest, 5 == highest

trans.at: A - T 碱基使用偏好;

假说:

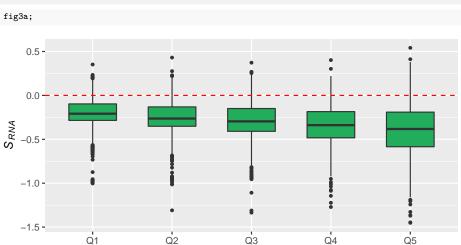
- de novo synthesis cost of A is higher than T,
- therefore highly expressed genes will tend to use T than A when possible.
- ◎ 因此,在高表达的基因当中,A T 的差值会变大(更负)。

data source: Chen et al, Nature Communications, 2016

boxplot

```
fig3a <-
  ggplot( data.fig3a, aes( factor(tai), trans.at ) ) +
    geom_boxplot( fill = "#22AD5C", linetype = 1 ,outlier.size = 1, width = 0.6) +
    xlab( "tAI group" ) +
    ylab( expression( paste( italic(S[RNA]) ) ) ) +
    scale_x_discrete(breaks= 1:5 , labels= paste("Q", 1:5, sep = "") ) +
    geom_hline( yintercept = 0, colour = "red", linetype = 2);</pre>
```

show the plot



说明:

① 此种情况下,我们通常只看 median 值的趋势;

tAl group

another example plot

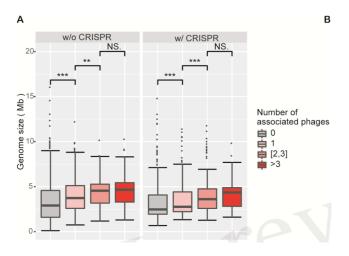
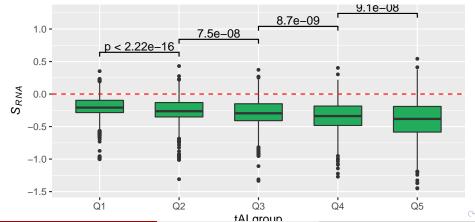


Figure 1: a figure from an in-press manuscript

how to add significance indicators to plot??

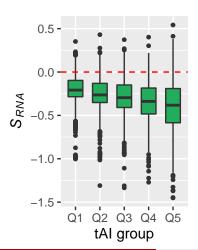
ggplot2 的扩展包,geom_signif;如果第一次使用,请先安装。



boxplot 画图注意事项

正确的画法为: 要高瘦, 不要矮胖!!!!

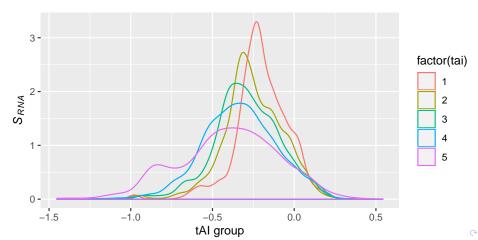
fig3a;



此类数据的另一种可视化方式

density plot; 但在此例中, 不如 boxplot 好。

```
ggplot( data.fig3a, aes( trans.at, colour = factor(tai) ) ) + geom_density( ) +
xlab( "tAI group" ) + ylab( expression( paste( italic(S[RNA]) ) ) );
```



section 4: two column data: part 2

数据介绍: two numerical vectors

多用于描述两组(量化)数据之间的关系;

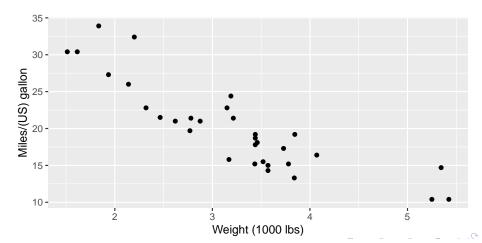
以 mtcars 为例:

```
head(mtcars);
```

```
mpg cyl disp hp drat
                                              wt qsec vs am gear carb
                    21.0
## Mazda RX4
                           6 160 110 3.90 2.620 16.46 0
## Mazda RX4 Wag
                    21.0
                              160 110 3.90 2.875 17.02 0 1
                              108 93 3.85 2.320 18.61 1 1 4
258 110 3.08 3.215 19.44 1 0 3
## Datsun 710
                    22.8
                    21.4
## Hornet 4 Drive
                              360 175 3.15 3.440 17.02 0 0
## Hornet Sportabout 18.7 8
## Valiant
                    18.1
                              225 105 2.76 3.460 20.22 1 0
```

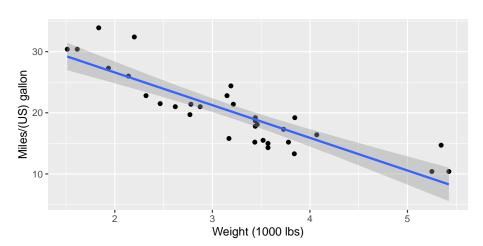
查看重量与燃油效率之间的关系

```
plotcars <-
   ggplot( mtcars, aes( x = wt, y = mpg ) ) +
      geom_point() + xlab( "Weight (1000 lbs)" ) + ylab( "Miles/(US) gallon" );
plotcars;</pre>
```



smooth, 减少噪音

```
plotcars + geom_smooth( method = "lm" ); ## default is lowess
```



作业: 计算 correlation 并做图

用 talk09 中介绍的方法添加类似于第 43 页的两个公式

expression(R^2) variance of x explained by y ...

cor ## 0.7528328

当趋势不明显时,可以按另一组数据分组

这里还以 mtcars 为例。

两种分组(binning)方法 equal-distance, equal-size binning

举例:

ntile 函数的参数

... *tile 函数都是 equal size

```
## ntile 的结果
table( mtcars2$group1 );
```

```
##
## 1 2 3 4
## 8 8 8 8
```

cut 函数

按指定的间隔 (breaks) 对数据进行分割。

```
##
## [1.51,2.49] (2.49,3.47] (3.47,4.45] (4.45,5.42]
## 8 13 8 3
```

使用方法:

```
cut(x, ...)
# S3 method for default
cut(x, breaks, labels = NULL,
   include.lowest = FALSE, right = TRUE, dig.lab = 3,
   ordered_result = FALSE, ...)
```

cut 示例

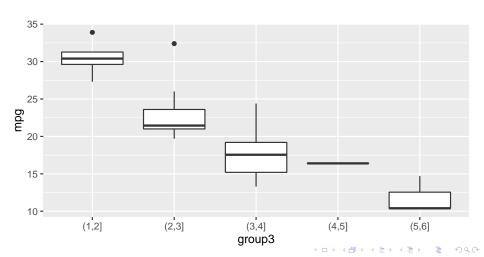
不仅可用于 equal distance, 还可以用于任意间距

```
mtcars3 <- mtcars2 %>%
  mutate( group3 = cut( mtcars$wt, breaks = c(0,1,2,3,4,5,6) ) ) ;
table(mtcars3$group3);
```

```
##
## (0,1] (1,2] (2,3] (3,4] (4,5] (5,6]
## 0 4 8 16 1 3
```

分组后的数据适合用 boxplot

```
ggplot( mtcars3, aes( group3, mpg ) ) +
  geom_boxplot();
```



小结

目前讲述了以下内容:

一维数据

table, summary, range, quantile, mean, median ...

二维数据

- boxplot
- point plot
- correlation
- 分组: equal distance, equal size binning ...

section 5: parametric tests

parametric tests

- 包括:
 - t-test
 - analysis of variance
 - linear regression
- ② 数据有较明确的分布 (e.g. normal distribution), 或假设数据有明确的分布; 当假设不成立时, 检测会无效;
- 更灵敏 (相比 nonparametric test), p-value 更低

more to read: http://rcompanion.org/handbook/I_01.html

适用性

适用于

- 数量化性状,比如:身高、体重、产量、污染值
- 整数值: 成绩、年龄、每天步数

不适用于

- 其它 count data 或者 discrete data;
- 或者有太多趋向于 min 或 max 的值
- 百分比或比例

详见: http://rcompanion.org/handbook/index.html

需要的 packages

需要的 packages

```
## chooseCRANmirror()
if(!require(psych)) {
   install.packages("psych");
}
if( !require(rcompanion) ) {
   install.packages("rcompanion");
}
library(psych);
library(rcompanion)
```

数据

注意 source() 函数的用法

```
source("data/talk10/input_data1.R"); ## 装入 Data data.frame ...
str(Data);
```

```
## 'data.frame': 26 obs. of 5 variables:
## $ Student: Factor w/ 26 levels "a","b","c","d",..: 1 2 3 4 5 6 7 8 9 10 ...
## $ Sex : Factor w/ 2 levels "female", "male": 1 1 1 1 1 1 2 2 2 2 ...
## $ Teacher: Factor w/ 3 levels "Catbus", "Satsuki",..: 1 1 1 1 1 1 1 1 1 1 ...
## $ Steps : int 8000 9000 10000 7000 6000 8000 7000 5000 9000 7000 ...
## $ Rating : int 7 10 9 5 4 8 6 5 10 8 ...
```

检查数据

library(psych)

```
##
## Attaching package: 'psych'
## The following objects are masked from 'package:ggplot2':
##
## %+%, alpha
```

headTail(Data); ## psych 包提供的函数

```
##
      Student
                 Sex Teacher Steps Rating
## 1
            a female Cathus
                              8000
            b female Catbus
## 2
                              9000
                                       10
            c female Catbus 10000
            d female Cathus 7000
         <NA> <NA>
                     <NA>
                               . . .
                                      . . .
## 23
            w male Totoro
                              6000
                                        8
## 24
            x male
                      Totoro
                              8000
                                       10
## 25
                male
                      Totoro
                              7000
## 26
                male Totoro
                              7000
                                        7
```

查看数据, cont.

```
## 其它常用函数
str(Data)
## 'data.frame': 26 obs. of 5 variables:
```

```
## $ Student: Factor w/ 26 levels "a","b","c","d",..: 1 2 3 4 5 6 7 8 9 10 ...
## $ Sex : Factor w/ 2 levels "female","male": 1 1 1 1 1 1 2 2 2 2 ...
## $ Teacher: Factor w/ 3 levels "Catbus","Satsuki",..: 1 1 1 1 1 1 1 1 1 1 1 ...
## $ Steps : int 8000 9000 10000 7000 8000 7000 5000 9000 7000 ...
## $ Rating : int 7 10 9 5 4 8 6 5 10 8 ...
```

```
summary(Data)
```

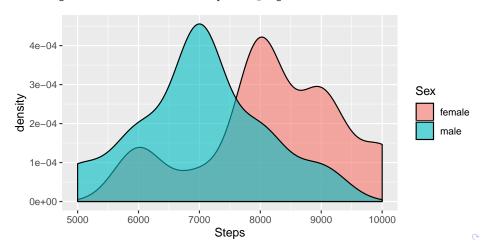
```
Student
                 Sex Teacher
                                       Steps
                                                    Rating
##
         . 1
            female:15 Cathus:10
                                   Min. : 5000
                                                 Min. : 4.000
     : 1
## b
              male :11
                        Satsuki: 7
                                   1st Qu.: 7000 1st Qu.: 7.000
##
     : 1
                        Totoro : 9
                                   Median: 8000
                                                 Median: 8.000
     . 1
                                   Mean : 7692
                                                 Mean : 7.615
                                   3rd Qu.: 8750
##
       : 1
                                                 3rd Qu.: 9.000
                                   Max. :10000
##
        : 1
                                                 Max. :10.000
   (Other):20
```

parametric test 的要求

- 随机取样
- ② 值或 residuals 为正态分布; residules 是指观察值与预测值 (mean) 之 差

数据的分布

Warning: Width not defined. Set with `position_dodge(width = ?)`



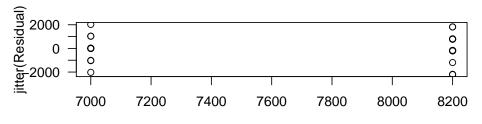
parametric test 的要求, cont.

● 有相同的 variance

```
M1 = mean(Data$Steps[Data$Sex=="female"])
M2 = mean(Data$Steps[Data$Sex=="male"])

Data$Mean[Data$Sex=="female"] = M1
Data$Mean[Data$Sex=="male"] = M2

Data$Residual = Data$Steps - Data$Mean
plot(jitter(Residual) ~ Mean, data = Data, las = 1);
```



how to detect outlier ??

一个很模糊的定义: Outliers are extreme values that fall a long way outside of the other observations. For example, in a normal distribution, outliers may be values on the tails of the distribution.

对于 normal distribution,通常 mean +- 2 or 3 * sd

对于 non-parametric distribution (注: IRQ 计算可使用同名函数: IRQ):

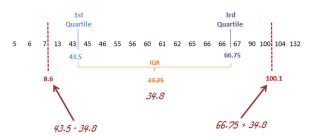
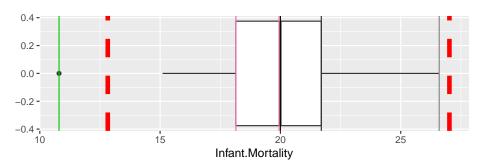


Figure 2: Tukey's method for outlier detection

an example of outlier values



other methods for detect outliers

- 1 https://conversionxl.com/blog/outliers/
- 1 https://www.r-bloggers.com/ 2 outlier-detection-and-treatment-with-r/
- https://machinelearningmastery.com/ how-to-identify-outliers-in-your-data/
- https://www.amazon.com/dp/1461463955
- https://www.itl.nist.gov/div898/handbook/eda/section3/eda35h.htm

one sample t-test

检测分布是否与预期一致;比如:男生每天的步数是否显著区别于 1 万

```
with( Data, t.test( Steps[ Sex == "male" ], mu = 1000 ) );
```

```
##
## One Sample t-test
##
## data: Steps[Sex == "male"]
## t = 18.166, df = 10, p-value = 5.484e-09
## alternative hypothesis: true mean is not equal to 1000
## 95 percent confidence interval:
## 6264.07 7735.93
## sample estimates:
## mean of x
## 7000
```

two samples t-test

with(Data, t.test(Steps ~ Sex))

比较 sd 和 mean ,可应用于正态分布。几种使用方法:

```
##
## Welch Two Sample t-test
##
## data: Steps by Sex
## t = 2.6424, df = 22.816, p-value = 0.01461
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 260.1421 2139.8579
## sample estimates:
## mean in group female mean in group male
## 8200 7000
```

two sample t-test 使用方法 2

```
##
## Welch Two Sample t-test
##
## data: Steps[Sex == "male"] and Steps[Sex == "female"]
## t = -2.6424, df = 22.816, p-value = 0.01461
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -2139.8579 -260.1421
## sample estimates:
## mean of x mean of y
## 7000 8200
```

two sample t test 检测结果

```
res <- with( Data, t.test( Steps ~ Sex ) );
str(res):
## List of 10
## $ statistic : Named num 2.64
## ..- attr(*, "names")= chr "t"
## $ parameter : Named num 22.8
    ..- attr(*, "names")= chr "df"
## $ p.value : num 0.0146
## $ conf.int : num [1:2] 260 2140
## ..- attr(*, "conf.level")= num 0.95
## $ estimate : Named num [1:2] 8200 7000
    ..- attr(*, "names")= chr [1:2] "mean in group female" "mean in group male"
## $ null.value : Named num 0
## ..- attr(*, "names")= chr "difference in means"
## $ stderr : num 454
## $ alternative: chr "two.sided"
## $ method : chr "Welch Two Sample t-test"
   $ data.name : chr "Steps by Sex"
   - attr(*, "class")= chr "htest"
```

paired two sample t test

例如:辅导前后的学生成绩:

```
source("data/talk10/input_data2.R");
head(scores);
```

```
## 1 Before a 65
## 2 Before b 75
## 3 Before c 86
## 4 Before d 69
## 5 Before e 60
```

Time Student Score

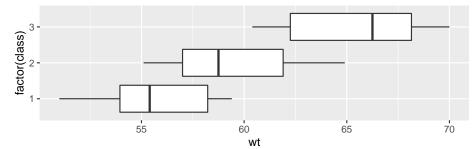
paired two sample t test

```
scores.wide <- scores %>% spread( Time, Score );
head(scores.wide. n = 3):
     Student After Before
                77
                       65
               98
                       75
## 3
                92
                    86
with( scores.wide, t.test( After, Before, paired = T ) );
##
   Paired t-test
##
## data: After and Before
## t = 3.8084, df = 9, p-value = 0.004163
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 4.141247 16.258753
## sample estimates:
## mean of the differences
##
                      10.2
```

one way ANOVA

ANOVA: similar to indepedent t-test, but can be applied to multiple groups

比如: 3 个班学生的体重



one way ANOVA, cont.

```
library(FSA); ## 如果没有这个句, 请先安装 ...
## ## FSA v0.8.25. See citation('FSA') if used in publication.
## ## Run fishR() for related website and fishR('IFAR') for related book.
##
## Attaching package: 'FSA'
## The following object is masked from 'package:psych':
##
##
       headtail
with( wts. Summarize( wt ~ class. digits = 3 ) ):
```

```
## class n mean sd min Q1 median Q3 max
## 1 1 20 55.490 2.812 51.0 53.95 55.40 58.225 59.4
## 2 2 20 59.250 3.059 55.1 57.00 58.75 61.900 64.9
## 3 3 20 65.395 3.193 60.4 62.25 66.25 68.150 70.0
```

linear model

两个问题:

● 组间有显著区别吗?

```
model <- lm( wt ~ class, data = wts );
anova( model );</pre>
```

ANOVA

summary(model);

② 分组对变量的贡献(r-square, aka. variance explained)

```
##
## Call:
## lm(formula = wt ~ class, data = wts)
## Residuals:
      Min
              1Q Median
                              3Q
                                    Max
## -4.9450 -2.7700 -0.2187 2.5419 5.0025
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 50.1400 1.0429 48.08 < 2e-16 ***
              4.9525 0.4828 10.26 1.2e-14 ***
## class
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.053 on 58 degrees of freedom
## Multiple R-squared: 0.6447, Adjusted R-squared: 0.6386
## F-statistic: 105.2 on 1 and 58 DF, p-value: 1.195e-14
```

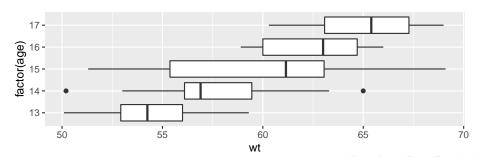
其中的值都是什么意思???

one way ANOVA with blocks

同时有多个因素影响体重时,哪些才是主要的?

```
wts2 <- bind_rows(
    tibble( class = 1, age = sample( 13:15, 20, replace = T ), wt = sample( seq(50, 60, by = 0.1
    tibble( class = 2, age = sample( 14:16, 20, replace = T ), wt = sample( seq(55, 65, by = 0.1
    tibble( class = 3, age = sample( 15:17, 20, replace = T ), wt = sample( seq(60, 70, by = 0.1
);

ggplot(wts2, aes( factor( age ), wt ) ) + geom_boxplot() + coord_flip();</pre>
```



one way ANOVA with blocks, cont.

one way ANOVA with blocks, 各个 factor 的重要性??

```
library(relaimpo);
res3 <- calc.relimp( wt ~ factor(class) + age, data = wts2 );
res3$R2; ## A R2
## [1] 0.7048711
res3$1mg: ## 每个因素的贡献;
## factor(class)
                         age
      0.4920945 0.2127766
## 测试 rela 参数:
res4 <- calc.relimp( wt ~ factor(class) + age, data = wts2, rela = T);
res4$R2: ## A R2
## [1] 0.7048711
res4$1mg; ## 每个因素的贡献;
```

factor(class) age ## 0.6981341 0.3018659

two way ANOVA

一个变量受另外两个因素影响的分;比如上例中 体重受 年级和 年龄的影响。

年级和 年龄至少有 4 个 unique combinations.

Summarize(wt ~ age + class, data = wts2, digits=3);

实际上,上面的 block test 可以认为是 two-way ANOVA 分析

```
Q1 median
    age class n
                   mean
                           sd min
                                                    03
## 1
     13
            1 12 54.575 2.784 50.1 52.925 54.25 56.000 59.3
## 2
     14
            1 4 54,650 3,760 50,2 52,300 54,95 57,300 58,5
## 3
            1 4 52.875 1.307 51.3 52.425 52.85 53.300 54.5
            2 7 59,386 3,621 55,7 56,600 58,10 61,850 65,0
            2 8 59.200 3.365 55.3 56.525 59.00 62.425 62.9
## 5
     15
## 6
     16
            2 5 60.300 1.594 58.9 59.400 60.00 60.200 63.0
## 7
     15
            3 8 64.062 3.034 60.1 61.375 64.45 65.725 69.1
            3 4 65.025 0.655 64.6 64.675
## 8
     16
                                          64.75 65.100 66.0
## 9
     17
            3 8 65.088 3.158 60.3 63.075
                                          65.40 67.275 69.0
```

two way ANOVA, cont.

```
model3 <- lm( wt ~ class + age + class:age, data = wts2);
anova( model3 );</pre>
```

其中: class 和 age 称为 main effects, class:age 称为 interaction effects

relative importance of interactions

```
res5 <- calc.relimp( wt ~ factor(class) + age + factor(class):age, data = wts2);
res5$R2; ## 总 R2

## [1] 0.7131791

res5$lmg; ## 每个因素的贡献;

## factor(class) factor(class):age age age age 0.492094503 0.008308066 0.212776574
```

more to read about the interaction effects:

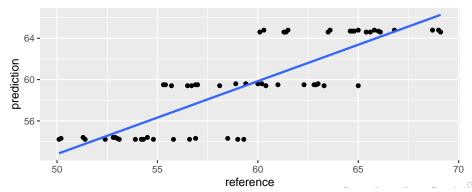
http://oak.ucc.nau.edu/rh232/courses/EPS625/Handouts/Two-way%20ANOVA/Understanding%20the%20Two-way%20ANOVA.pdf

用模型进行预测 predict

```
model2 = lm(formula = wt ~ class + age, data = wts2)
```

```
newdata <- wts2 %>% dplyr::select( class, age );
wt.predicted <- predict( model2, newdata );

dat <- data.frame( reference = wts2$wt, prediction = wt.predicted );
ggplot( dat , aes( x = reference, y = prediction ) ) + geom_point() +
    geom_smooth( method = "lm", se = F );</pre>
```



prediction 与 original data 的 correlation 是多少??

```
with( dat, cor.test( prediction, reference ) )$estimate;
         cor
## 0.8395383
## R ^ 2
with( dat, cor.test( prediction, reference ) ) $estimate ^2;
         cor
## 0.7048245
## 正好是 model2 的 r.squred ...
summary( model2 )$r.squared;
```

[1] 0.7048245

手动计算 prediction

在一个 linear model 中, wt = intercept + a * class + b * age 而 intercept, a, b 的值分别为:

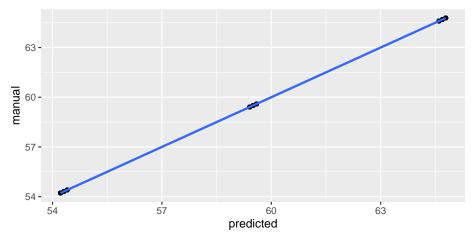
```
## (Intercept) class age
## 47.91315057 5.09584697 0.09304419

predicted2 <-
    paras[1] + paras["age"] * wts2$age + paras["class"] * wts2$class;

plot <-
    ggplot( data.frame( predicted = wt.predicted, manual = predicted2 ),
        aes( predicted, manual ) ) +
        geom_point() + geom_smooth( method = "lm", se = F );</pre>
```

show the plot

plot;



更多更方便的函数

以下函数也可以用于 multivariable analysis / multiple regression

```
fit <- lm(y ~ x1 + x2 + x3, data=mydata)
summary(fit) # show results

# Other useful functions
coefficients(fit) # model coefficients
confint(fit, level=0.95) # CIs for model parameters
fitted(fit) # predicted values
residuals(fit) # residuals
anova(fit) # anova table
vcov(fit) # covariance matrix for model parameters
influence(fit) # regression diagnostics</pre>
```

linear regression 注意事项

- ① 是 parametric test
- ② 假设变量之间独立(比如:年龄和班级之间没有关联)
- 4 homogeneity of variance

但实际上 ...

multivariable analysis

more to read:

- https://www.statmethods.net/stats/regression.html
- https://data.library.virginia.edu/
 getting-started-with-multivariate-multiple-regression/

```
# Multiple Linear Regression Example
fit <-lm(y ~x1 + x2 + x3, data=mydata)
summary(fit) # show results
# compare models
fit1 <- lm(y \sim x1 + x2 + x3 + x4, data=mydata)
fit2 < -lm(y ~ x1 + x2)
anova(fit1, fit2)
# K-fold cross-validation
library(DAAG)
cv.lm(df=mydata, fit, m=3) # 3 fold cross-validation
# Stepwise Regression; feature selection
library(MASS)
fit <-lm(v~x1+x2+x3.data=mvdata)
step <- stepAIC(fit, direction="both")</pre>
step$anova # display results
                                                              イロト イボト イミト イミト
```

extended reading

- repeated measures ANOVA:
 http://rcompanion.org/handbook/I_09.html, 同一变量、不同时间段的重复测量(对上例中学生的体重进行多次测量)
- correlation and linear regression: http://rcompanion.org/handbook/I_10.html
- onn-linear regression: https://www.amazon.com/ Statistical-Tools-Nonlinear-Regression-Statistics/dp/ 0387400818

更多详见: http://rcompanion.org/handbook/I_08.html

section 6: non-parametric test

wilcox.test and kruskal.test

```
# independent 2-group Mann-Whitney U Test
with( Data, wilcox.test( Steps ~ Sex ) );
## Warning in wilcox.test.default(x = c(8000L, 9000L, 10000L, 7000L, 6000L, :
## cannot compute exact p-value with ties
##
   Wilcoxon rank sum test with continuity correction
##
## data: Steps by Sex
## W = 127.5, p-value = 0.01773
## alternative hypothesis: true location shift is not equal to 0
# Kruskal Wallis Test One Way Anova by Ranks
with( Data, kruskal.test( Steps ~ Sex ) ):
```

```
##
## Kruskal-Wallis rank sum test
##
## data: Steps by Sex
## Kruskal-Wallis chi-squared = 5.7494, df = 1, p-value = 0.01649
```

作业与练习

作业与练习

练习本课堂所讲的内容 ...

