

Introduction to R

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Pocket calculator

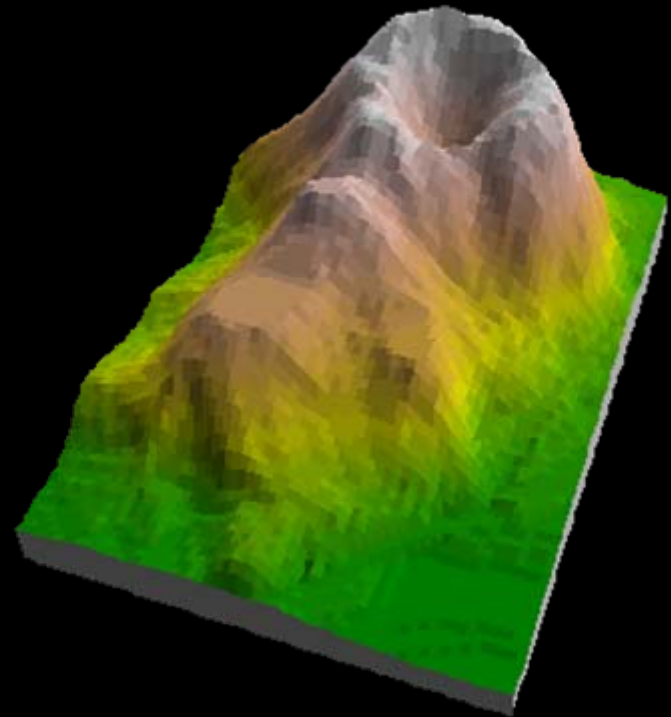
Data manipulation

Graphics

Pocket calculator

Data manipulation

Graphics



Pocket calculator

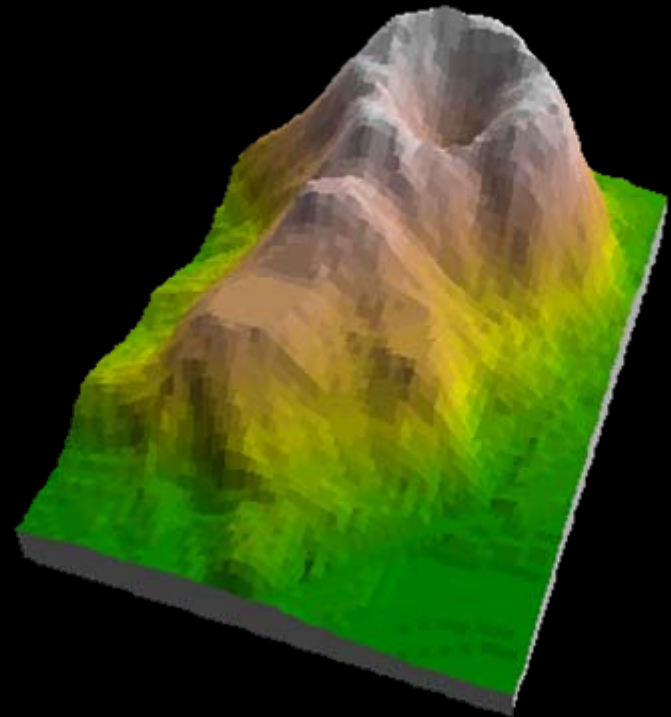
Data manipulation

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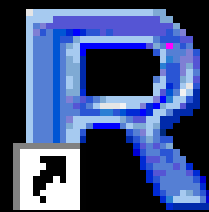
Statistical modelling

Computationally
Intensive Statistics

Programming



F:\valdar\TuesdayMorning\



R is a big calculator

Simple operations

`3*4`

`20 + 4`

`20/4`

`9^2`

`log(100, base=10)`

`2.5e2 / 2.5`

R is a big calculator

Simple operations

`3*4`

`20 + 4`

`20/4`

`9^2`

`log(100, base=10)`

`2.5e2 / 2.5`

Expressions

`(3*4)/(20+4)`

`3*(4/20)+4`

`(5/9)*(48-32)`

R is a big calculator

Variables

```
temp.f <- 48
```

```
temp.f
```

```
(5/9)*(temp.f-32)
```

```
temp.c <- (5/9)*(temp.f-32)
```

```
temp.c
```

R is a big calculator

Variables

```
temp.f <- 48  
temp.f  
(5/9)*(temp.f-32)  
temp.c <- (5/9)*(temp.f-32)  
temp.c
```

Write your own functions

```
will.fun <- function(x) { (5/9)*(x-32) }  
will.fun(48)  
will.fun(temp.f)  
thing <- will.fun(26)
```

Data types

```
class(will.fun)
```

```
class(temp.c)
```

```
my.name <- "Will"
```

```
class(my.name)
```

Data types

```
class(will.fun)
```

```
class(temp.c)
```

```
my.name <- "Will"
```

```
class(my.name)
```

Vectors

```
a <- c("Will", "Ben")
```

```
x <- c(1,2,3)
```

```
sum(x)
```

```
mean(x)
```

```
range(x)
```

```
var(x)
```

```
1:3
```

Matrices

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} 5 & 1 & 1 \\ 5 & 2 & 14 \\ 5 & 3 & -1 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} 7 \\ 8 \\ 9 \end{bmatrix}$$

```
A <- matrix(1:9, byrow=TRUE, ncol=3)
B <- rbind( c(5,1,1), c(5,2,14), c(5,3,-1) )
x <- c(7,8,9)
```

Matrices

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} 5 & 1 & 1 \\ 5 & 2 & 14 \\ 5 & 3 & -1 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} 7 \\ 8 \\ 9 \end{bmatrix}$$

```
A <- matrix(1:9, byrow=TRUE, ncol=3)
B <- rbind( c(5,1,1), c(5,2,14), c(5,3,-1) )
x <- c(7,8,9)
```

```
t(A)
```

```
diag(A)
```

```
A * B
```

```
A %*% B
```

```
A + 1
```

```
median(x)
```

```
range(B)
```

```
max(x)
```

```
B.inv <- solve(B)
```

```
3*A
```

Matrices

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} 5 & 1 & 1 \\ 5 & 2 & 14 \\ 5 & 3 & -1 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} 7 \\ 8 \\ 9 \end{bmatrix}$$

Subscripting is easy

`A[1,1]`

`A[3,2]`

`x[2]`

`A[3,]`

`A[,3]`

`A[3, c(1,2)]`

`A[, c(1,2)]`

`A[2,2] <- 1975`

Exercises 1

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} 11 & 12 \\ 21 & 22 \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} 4 \\ 5.2 \end{bmatrix}$$

1) $(3.5 + 6) \times (7.9)^2$

2) $\mathbf{A} + \mathbf{B}$

3) $\bar{x}$

4) Convert $\mathbf{B}$ from Fahrenheit to Centigrade

5) $\mathbf{Ax}$

*6) $\mathbf{A}^{-1}\mathbf{Bx}$

Answers 1

```
A <- matrix(1:4, byrow=TRUE, ncol=2)
B <- cbind(c(11,21), c(12,22))
x <- c(4, 5.2)
```

1) $(3.5+6)*((7.9)^2)$

2) $A + B$

3) $\text{mean}(x)$

4) $(5/9)*(B-32)$

or

$\text{will.fun}(B)$

5) $A \%*\% x$

6) $\text{solve}(A) \%*\% B \%*\% x$

Random numbers

```
rnorm(10)
```

```
rnorm(10, mean=5, sd=2)
```

```
rnorm(100, mean=5, sd=2)
```

```
?rnorm
```

Random numbers

```
rnorm(10)
```

```
rnorm(10, mean=5, sd=2)
```

```
rnorm(100, mean=5, sd=2)
```

```
library(lattice)
```

```
histogram( rnorm(100, mean=5, sd=2) )
```

Simulation

```
args(sample)
```

```
sample(c("Heads", "Tails"), size=1)
```

Simulation

```
args(sample)
```

```
sample(c("Heads", "Tails"), size=1)
```

0 allele does nothing, 1 makes you taller

```
sample(0:1, size=1)
```

```
sample(0:1, size=2, replace=TRUE)
```

```
sum( sample(0:1, size=2, replace=TRUE ) )
```

Simulation

```
args(sample)
```

```
sample(c("Heads", "Tails"), size=1)
```

0 allele does nothing, 1 makes you taller

```
sample(0:1, size=1)
```

```
sample(0:1, size=2, replace=TRUE)
```

```
sum( sample(0:1, size=2, replace=TRUE ) )
```

Function to make a fake genotype

```
make.genotype <- function(){ sum(sample(0:1,  
size=2, replace=TRUE) )}
```

```
make.genotype()
```

Simulating the genetic basis of height

```
gmat <- matrix(ncol=5, nrow=100)  
gmat
```

Simulating the genetic basis of height

```
gmat <- matrix(ncol=5, nrow=100)
gmat
for (person in 1:100) {
  for (gene in 1:5) {
    gmat[person, gene] <- make.genotype()
  }
}
gmat
```

Simulating the genetic basis of height

```
gmat <- matrix(ncol=5, nrow=100)
gmat
for (person in 1:100) {
  for (gene in 1:5) {
    gmat[person, gene] <- make.genotype()
  }
}
gmat
head(gmat)
```

Simulating the genetic basis of height

```
gmat <- matrix(ncol=5, nrow=100)
```

```
gmat
```

```
for (person in 1:100) {
```

```
  for (gene in 1:5) {
```

```
    gmat[person, gene] <- make.genotype()
```

```
  }}
```

```
gmat
```

```
head(gmat)
```

Genotype of person 1:

```
gmat[1,]
```

Simulating the genetic basis of height

```
gmat <- matrix(ncol=5, nrow=100)
```

```
gmat
```

```
for (person in 1:100) {
```

```
  for (gene in 1:5) {
```

```
    gmat[person, gene] <- make.genotype()
```

```
  }}
```

```
gmat
```

```
head(gmat)
```

Genotype of person 1:

```
gmat[1,]
```

Height of person 1:

```
gmat[1,] %*% c( 1, 1, 1, 1, 1)
```

```
gmat[1,] %*% c( 3, 5, 2, 3, 1)
```

Simulating the genetic basis of height

All heights

```
gmat %*% c( 3, 5, 2, 3, 1)
```

```
histogram(gmat %*% c( 3, 5, 2, 3, 1))
```

Further reading

Books:

Introductory Statistics with R

by Peter Dalgaard

Modern Applied Statistics with S

by Venables and Ripley

Web sites:

<http://www.r-project.org/>

<http://wiki.r-project.org/rwiki/doku.php>

<http://news.gmane.org/gmane.comp.lang.r.general>

Graph gallery:

<http://addictedtor.free.fr/graphiques/>

	A	B	C	D	E	F	G	H	I
1	sex	weight	country	age	snp.A	snp.B	family	phenotype	status
2	F	91.29	Europe	42	0	1	Family3	-2.6134576	FALSE
3	M	154.22	US	42	1	2	Family6	-7.7237116	FALSE
4	F	95.74	US	40	2	0	Family1	3.2361416	TRUE
5	F	110.25	Australia	30	0	1	Family10	-0.026201	TRUE
6	M	170.47	Australia	36	0	2	Family1	-11.132778	FALSE
7	M	127.57	US	45	1	1	Family6	-3.6016196	FALSE
8	M	125.56	Europe	44	1	1	Family9	-7.0376314	FALSE

data.frame

	A	B	C	D	E	F	G	H	I
1	sex	weight	country	age	snp.A	snp.B	family	phenotype	status
2	F	91.29	Europe	42	0	1	Family3	-2.6134576	FALSE
3	M	154.22	US	42	1	2	Family6	-7.7237116	FALSE
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6	M	170.47	Australia	36	0	2	Family1	-11.132778	FALSE
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8	M	125.56	Europe	44	1	1	Family9	-7.0376314	FALSE

data.frame

	A	B	C	D	E	F	G	H	I
1	sex	weight	country	age	snp.A	snp.B	family	phenotype	status
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8	M	125.56	Europe	44	1	1	Family9	-7.0376314	FALSE

```
data <- read.delim("F:valdar/disease.txt")
class(data)
head(data)
```

data.frame

	A	B	C	D	E	F	G	H	I
1	sex	weight	country	age	snp.A	snp.B	family	phenotype	status
2	F	91.29	Europe	42	0	1	Family3	-2.6134576	FALSE
3	M	154.22	US	42	1	2	Family6	-7.7237116	FALSE
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6	M	170.47	Australia	36	0	2	Family1	-11.132778	FALSE
7	M	127.57	US	45	1	1	Family6	-3.6016196	FALSE
8	M	125.56	Europe	44	1	1	Family9	-7.0376314	FALSE

	sex	weight	country	age	snp.A	snp.B	family	phenotype	status
1	F	91.29	Europe	42	0	1	Family3	-2.61345764	FALSE
2	M	154.22	US	42	1	2	Family6	-7.72371165	FALSE
3	F	95.74	US	40	2	0	Family1	3.23614160	TRUE
4	F	110.25	Australia	30	0	1	Family10	-0.02620101	TRUE
5	M	170.47	Australia	36	0	2	Family1	-11.13277756	FALSE
6	M	127.57	US	45	1	1	Family6	-3.60161956	FALSE

a variable with other variables hanging off it

```
str(data)
```

```
'data.frame':   400 obs. of  9 variables:
 $ sex       : Factor w/ 2 levels "F","M": 1 2 1 1 2 ...
 $ weight    : num   91.3 154.2  95.7 110.2 170.5 ...
 $ country   : Factor w/ 3 levels "Australia","Europe",...: 2 3 ...
 $ age       : int   42 42 40 30 36 45 44 38 40 34 ...
 $ snp.A     : int    0 1 2 0 0 1 1 0 2 1 ...
 $ snp.B     : int    1 2 0 1 2 1 1 0 1 1 ...
 $ family    : Factor w/ 10 levels "Family1",...: 4 7 ...
 $ phenotype: num   -2.6135 -7.7237  3.2361 ...
 $ status    : logi  FALSE FALSE  TRUE  TRUE FALSE FALSE ...
```

a variable with other variables hanging off it

```
str(data)
```

```
'data.frame':   400 obs. of  9 variables:
 $ sex       : Factor w/ 2 levels "F","M": 1 2 1 1 2 ...
 $ weight    : num   91.3 154.2  95.7 110.2 170.5 ...
 $ country   : Factor w/ 3 levels "Australia","Europe",...: 2 3 ...
 $ age       : int   42 42 40 30 36 45 44 38 40 34 ...
 $ snp.A     : int    0 1 2 0 0 1 1 0 2 1 ...
 $ snp.B     : int    1 2 0 1 2 1 1 0 1 1 ...
 $ family    : Factor w/ 10 levels "Family1",...: 4 7 ...
 $ phenotype: num   -2.6135 -7.7237  3.2361 ...
 $ status    : logi  FALSE FALSE  TRUE  TRUE FALSE FALSE ...
```

```
data$weight
```

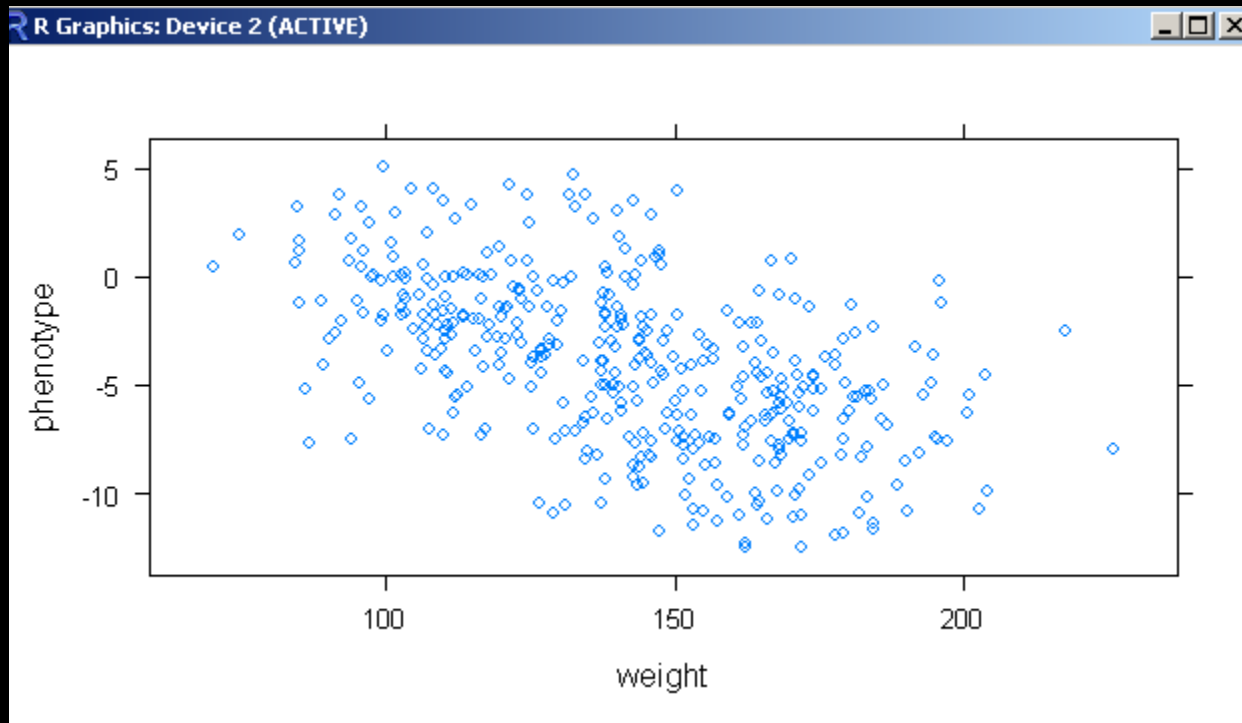
```
data$weight[3:10]
```

```
mean(data$weight)
```

```
library(lattice)
```

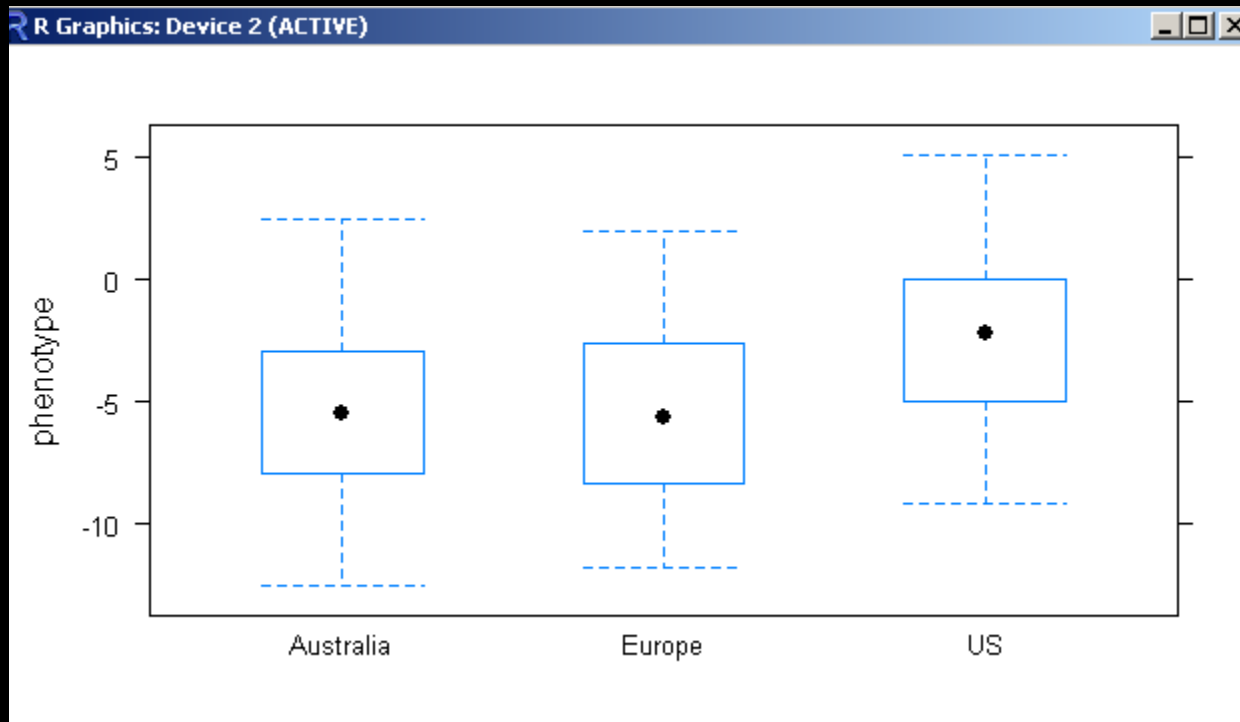
xyplot()

```
xyplot( phenotype ~ weight, data=data)
```



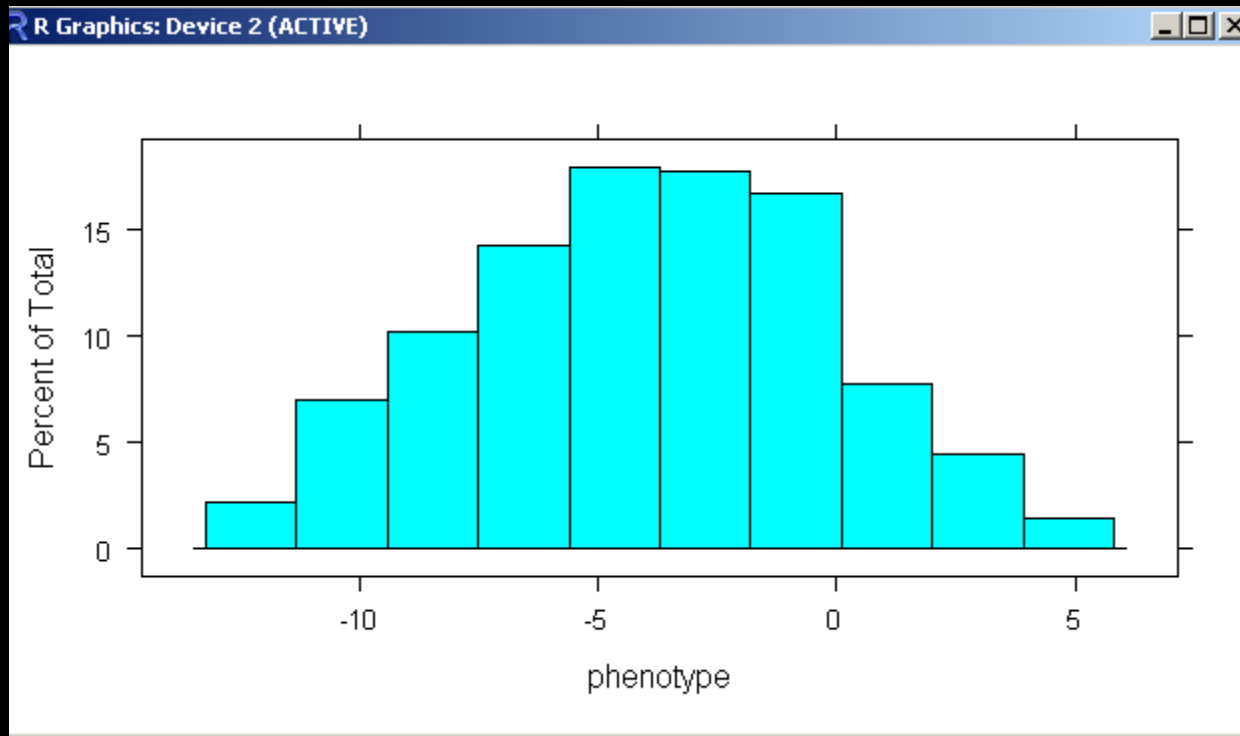
bwplot() , ie, box and whisker plot

```
bwplot( phenotype ~ country, data=data)
```



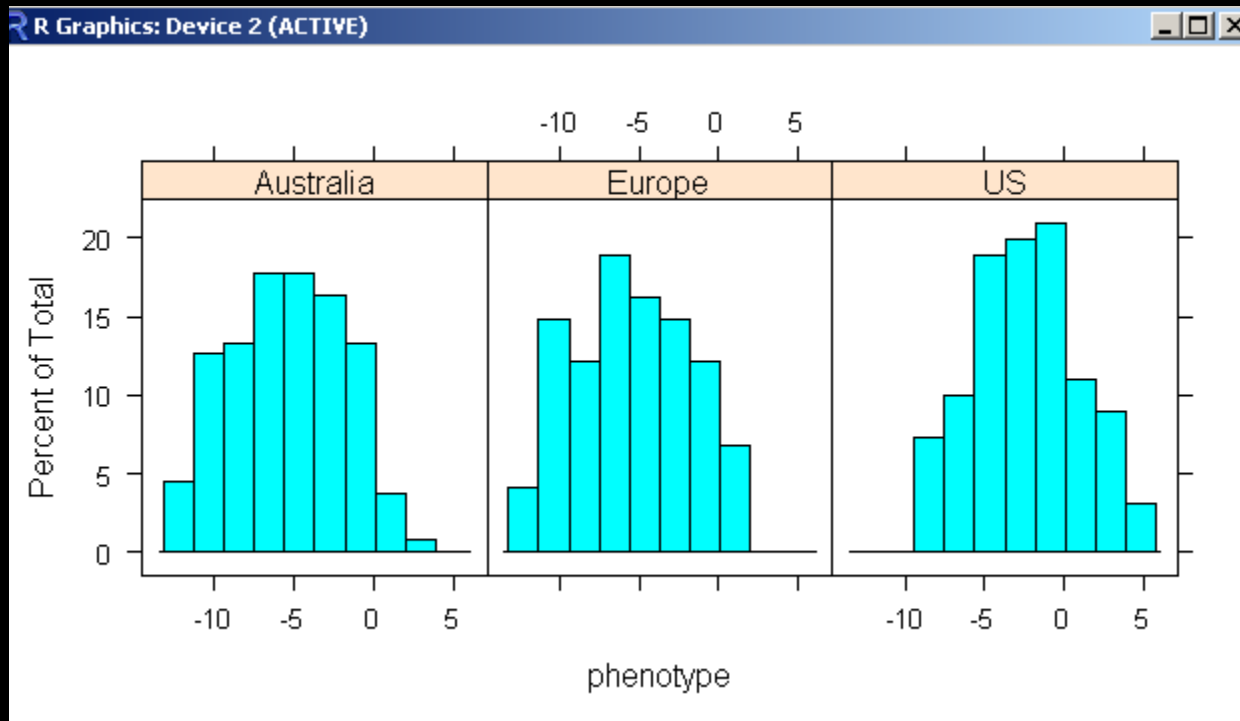
histogram()

```
histogram( ~ phenotype, data=data)
```



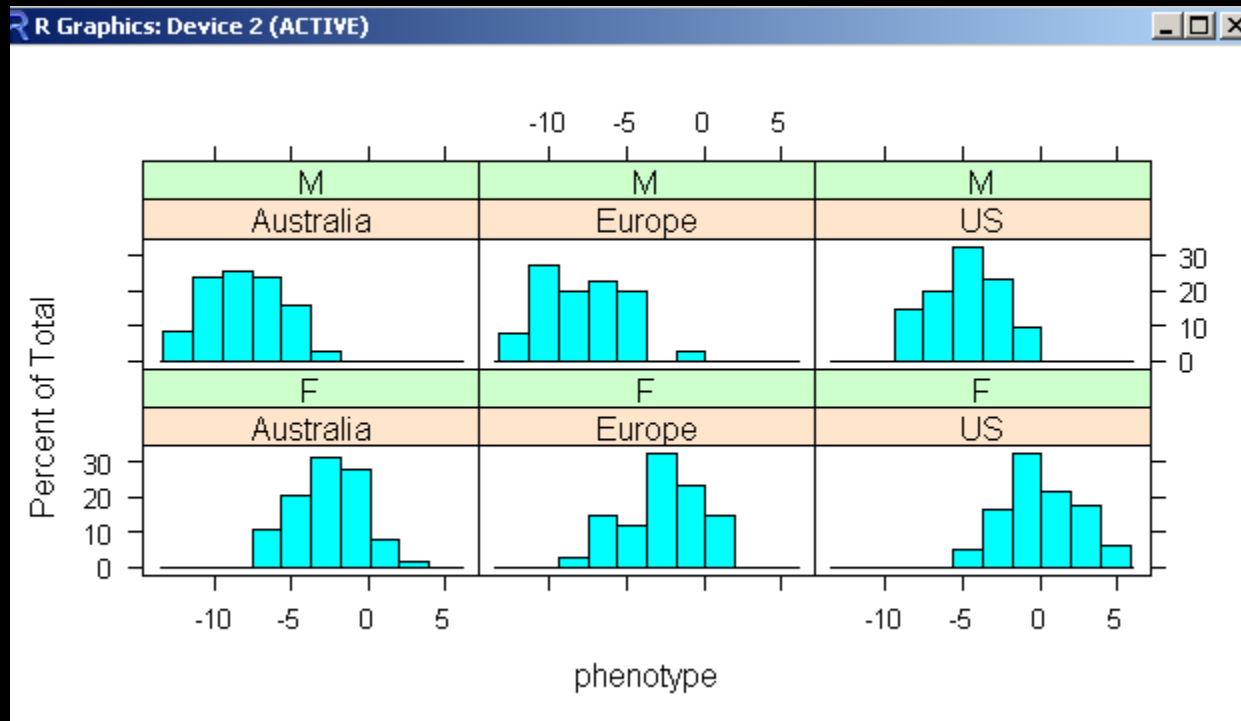
Conditioning

```
histogram( ~ phenotype | country, data=data)
```



Conditioning

```
histogram( ~ phenotype | sex * country, data=data)
```



Conditioning

Examples:

```
xyplot( phenotype ~ weight | sex, data=data)
```

```
bwplot( phenotype ~ country | snp.A, data=data)
```

Exercises 2

- 1) What do you think the status variable means and how is it defined? (*conditional histogram*)
- 2) How does snp.B affect the phenotype? (*cond. hist*)
- 3) Is there an effect of age on the phenotype? Are you sure? (*conditional xyplot*)
- 4) Does weight affect the phenotype? Does sex? Does sex affect weight? Does weight affect sex? What's going on here? (*conditional xyplot*)

Answers 2

- 1) `histogram( ~ phenotype | status, data=data)`
- 2) `xyplot(phenotype ~ snp.B, data=data)`
`histogram( ~ phenotype | snp.B, data=data)`
- 3) `xyplot(phenotype ~ age, data=data)`
`xyplot(phenotype ~ age | sex, data=data)`
- 4) `bwplot(phenotype ~ sex, data=data)`
`xyplot(phenotype ~ weight, data=data)`
`xyplot(phenotype ~ weight | sex, data=data)`

Answers 2: question 4

