Feuille de Travaux Dirigés nº 4 Les graphiques et R

Les exemples de cette feuille de travaux dirigées sont tirés de l'aide du logiciel R

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1 Contenu

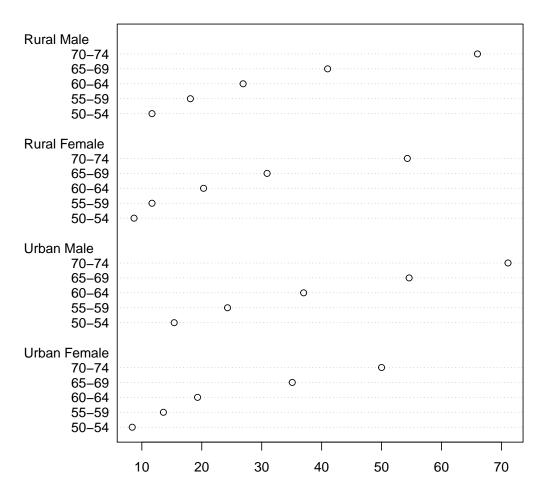
Nous allons nous intéresser à différents types de représentations graphiques adaptées à la nature des variables que nous souhaitons représenter.

```
library(graphics)
```

2 dotchart

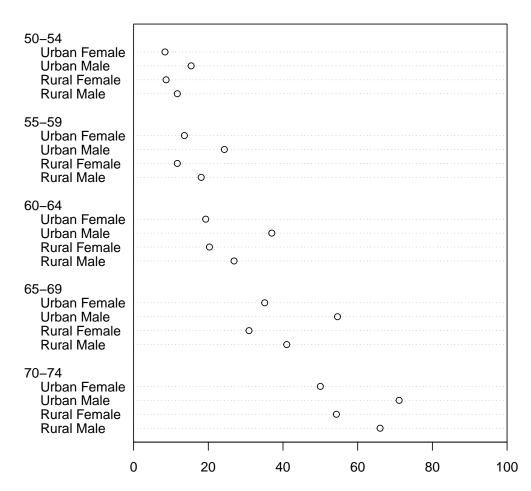
```
dotchart(VADeaths, main = "Death Rates in Virginia - 1940")
```

Death Rates in Virginia - 1940



```
op <- par(xaxs="i")
dotchart(t(VADeaths), main = "Death Rates in Virginia - 1940", xlim = c(0,100))</pre>
```

Death Rates in Virginia - 1940

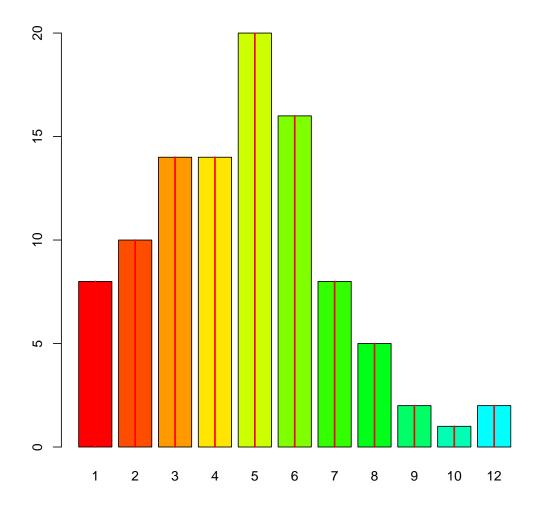


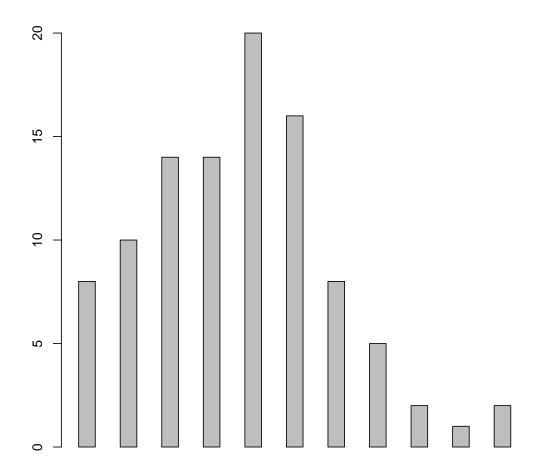
```
par(op)
```

3 barplot

```
require(grDevices)
tN <- table(Ni <- stats::rpois(100, lambda=5))</pre>
```

```
r <- barplot(tN, col=rainbow(20))
lines(r, tN, type='h', col='red', lwd=2)</pre>
```





sous-titre

```
barplot(VADeaths, plot = FALSE)
## [1] 0.7 1.9 3.1 4.3
```

```
barplot(VADeaths, plot = FALSE, beside = TRUE)

## [,1] [,2] [,3] [,4]

## [1,] 1.5 7.5 13.5 19.5

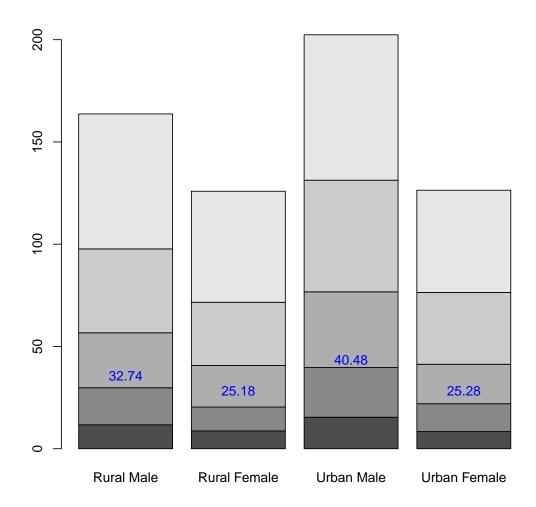
## [2,] 2.5 8.5 14.5 20.5

## [3,] 3.5 9.5 15.5 21.5

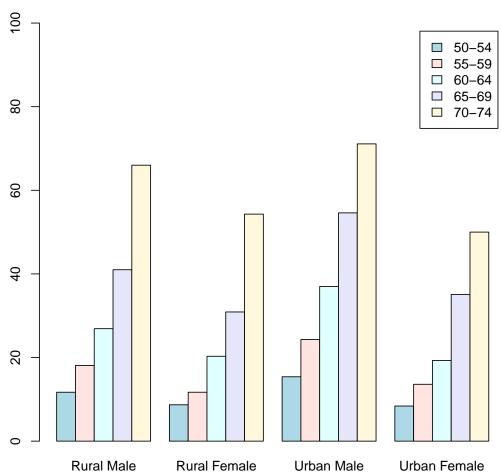
## [4,] 4.5 10.5 16.5 22.5

## [5,] 5.5 11.5 17.5 23.5
```

```
mp <- barplot(VADeaths)
tot <- colMeans(VADeaths)
text(mp, tot + 3, format(tot), xpd = TRUE, col = "blue")</pre>
```



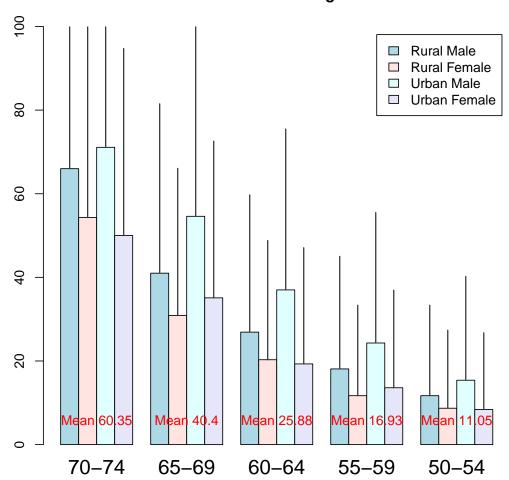




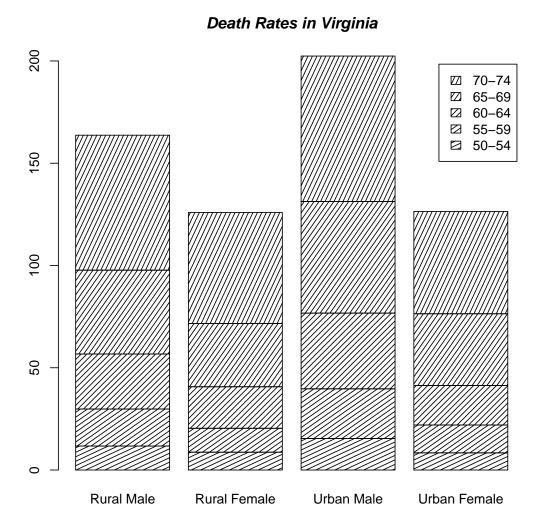
```
hh <- t(VADeaths)[, 5:1]
mybarcol <- "gray20"</pre>
```

```
mtext(side = 1, at = colMeans(mp), line = -2,
    text = paste("Mean", formatC(colMeans(hh))), col = "red")
```

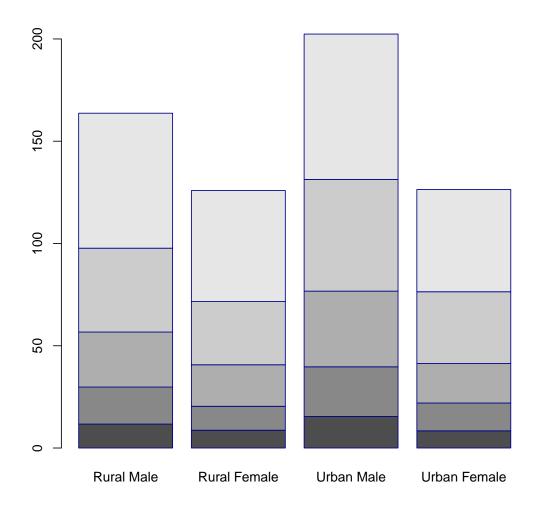
Death Rates in Virginia



Faked upper 2*sigma error bars

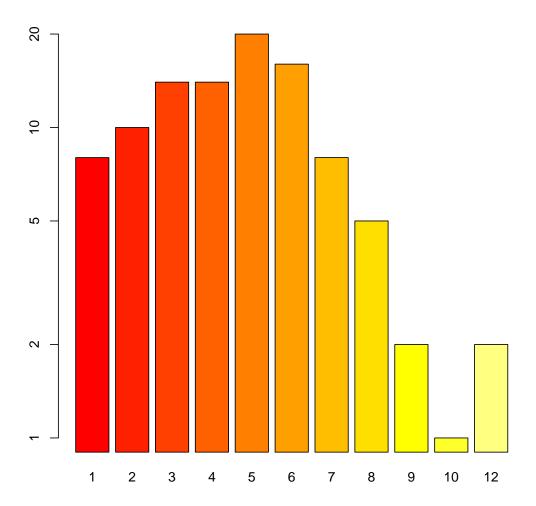


barplot(VADeaths, border = "dark blue")

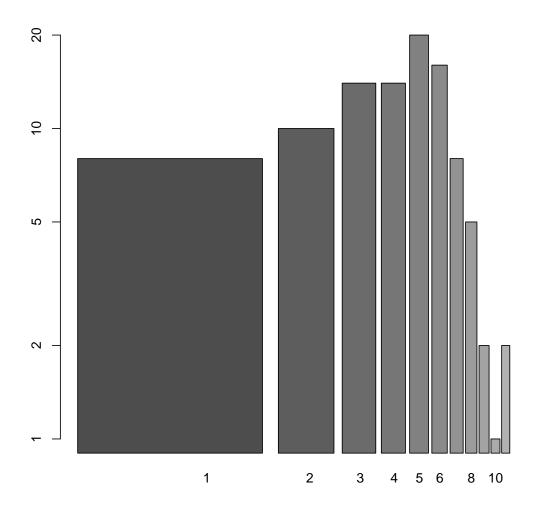


Échelles logarithmiques

```
barplot(tN, col=heat.colors(12), log = "y")
```

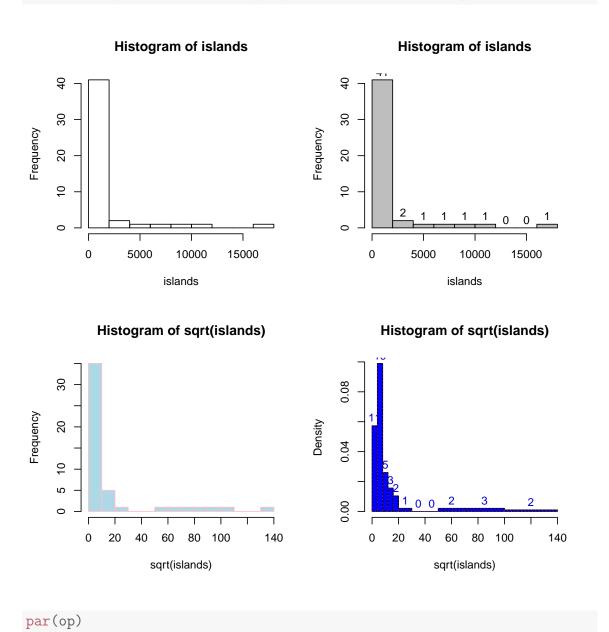


barplot(tN, col=gray.colors(20), log = "xy")



4 hist





```
require(utils) # for str
str(hist(islands, breaks=12, plot= FALSE)) #-> 10 (~= 12) breaks

## List of 6
## $ breaks : num [1:10] 0 2000 4000 6000 8000 10000 12000 14000 16000 18000
## $ counts : int [1:9] 41 2 1 1 1 1 0 0 1
## $ density : num [1:9] 4.27e-04 2.08e-05 1.04e-05 1.04e-05 1.04e-05 ...
## $ mids : num [1:9] 1000 3000 5000 7000 9000 11000 13000 15000 17000
```

```
## $ xname : chr "islands"
## $ equidist: logi TRUE
## - attr(*, "class") = chr "histogram"

str(hist(islands, breaks=c(12,20,36,80,200,1000,17000), plot = FALSE))

## List of 6
## $ breaks : num [1:7] 12 20 36 80 200 1000 17000

## $ counts : int [1:6] 12 11 8 6 4 7

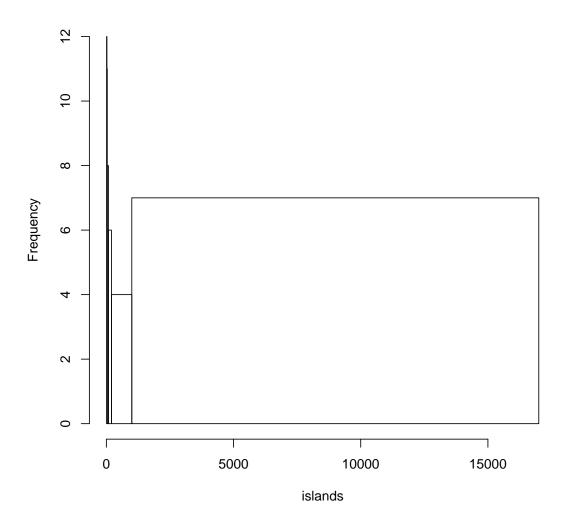
## $ density : num [1:6] 0.03125 0.014323 0.003788 0.001042 0.000104 ...
## $ mids : num [1:6] 16 28 58 140 600 9000

## $ xname : chr "islands"

## $ equidist: logi FALSE
## - attr(*, "class") = chr "histogram"
```

```
hist(islands, breaks=c(12,20,36,80,200,1000,17000), freq = TRUE,
    main = "WRONG histogram") # and warning
## Warning in plot.histogram(r, freq = freq1, col = col, border =
border, angle = angle, : the AREAS in the plot are wrong - rather use
    'freq = FALSE'
```

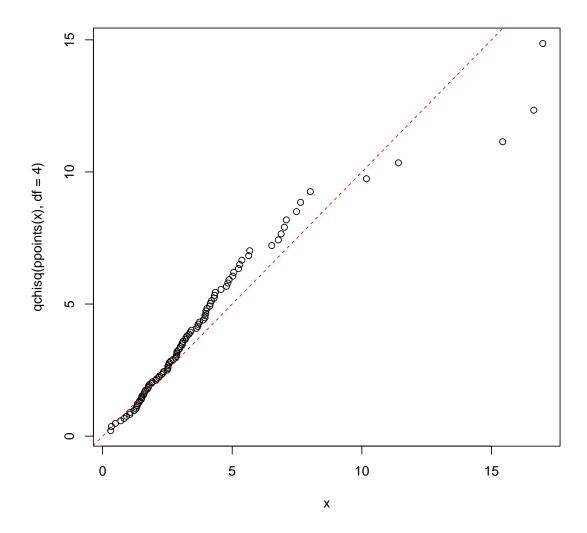
WRONG histogram



```
require(stats)
set.seed(14)
x <- rchisq(100, df = 4)</pre>
```

Comparing data with a model distribution should be done with qqplot()!

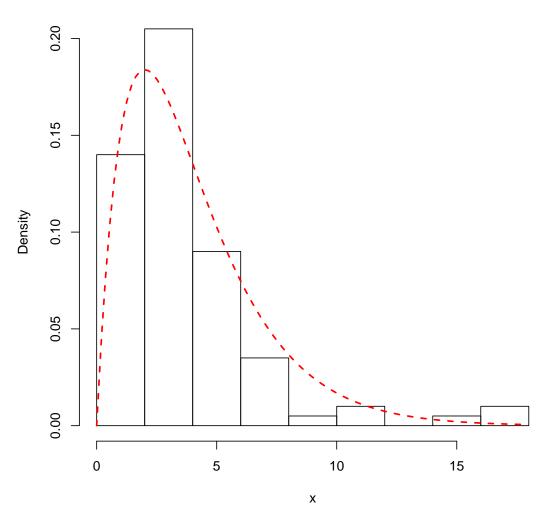
```
qqplot(x, qchisq(ppoints(x), df = 4)); abline(0,1, col = 2, lty = 2)
```



if you really insist on using hist () \dots :

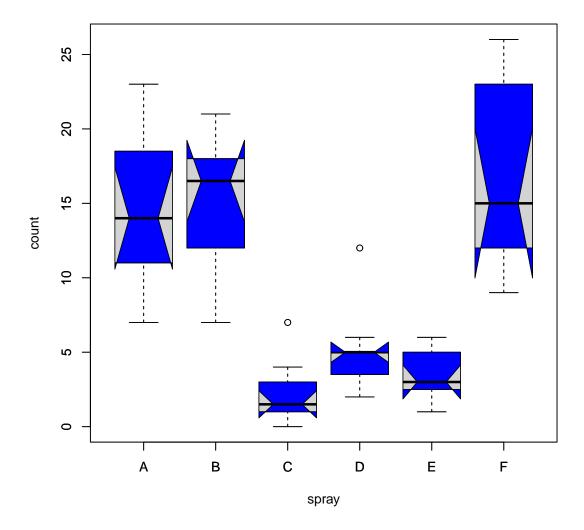
```
hist(x, freq = FALSE, ylim = c(0, 0.2))
curve(dchisq(x, df = 4), col = 2, lty = 2, lwd = 2, add = TRUE)
```



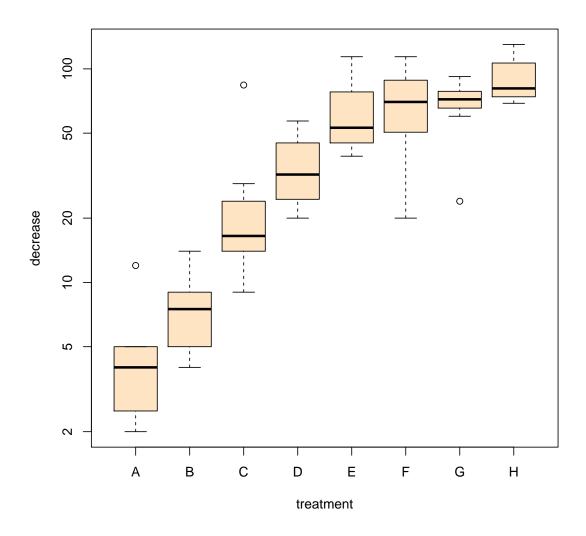


5 Boîtes à moustaches

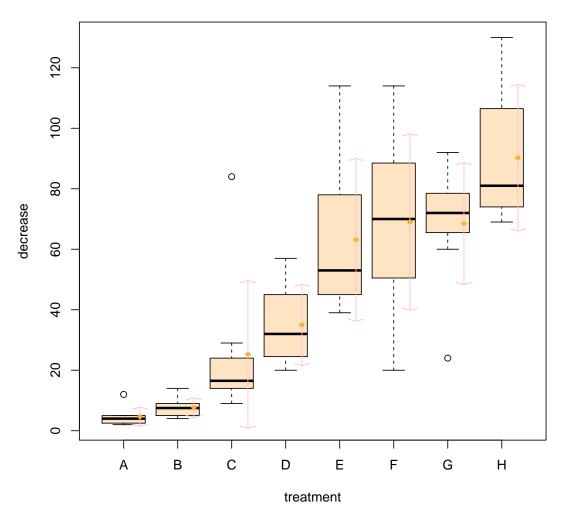
5.1 boxplot d'une formule



The last command add notches : If the notches of two plots do not overlap this is 'strong evidence' that the two medians differ.



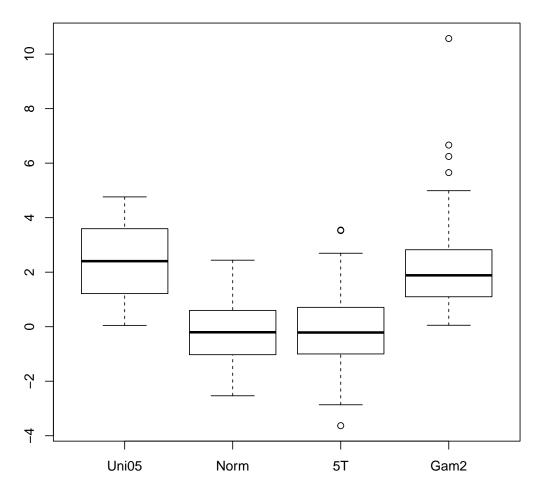




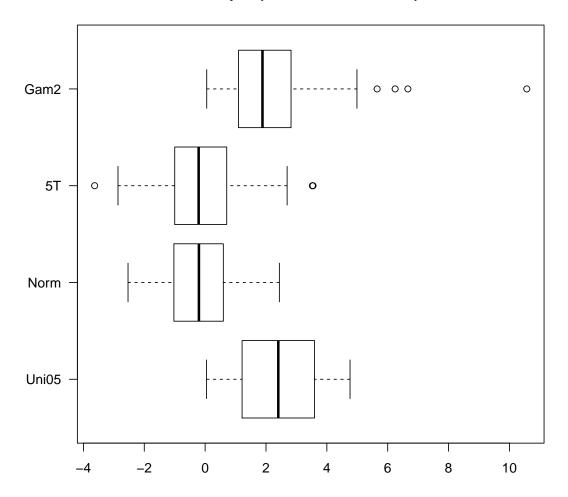
5.2 boxplot d'une matrice

```
boxplot(as.data.frame(mat),
    main = "boxplot(as.data.frame(mat), main = ...)")
```



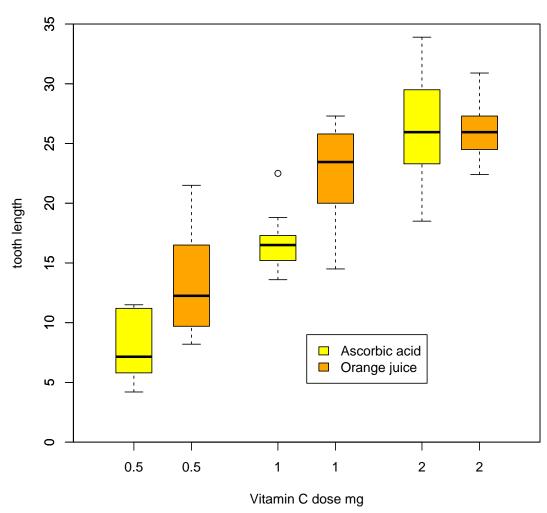


boxplot(*, horizontal = TRUE)



Using 'at = ' and adding boxplots – example idea by Roger Bivand :

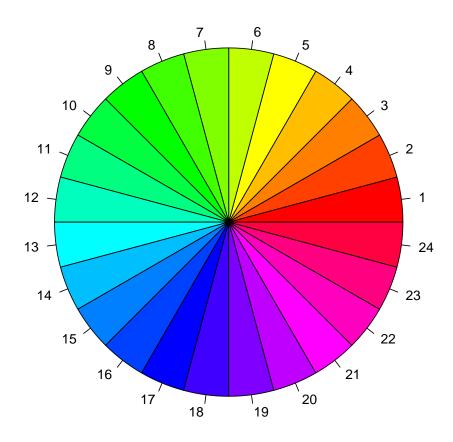
Guinea Pigs' Tooth Growth



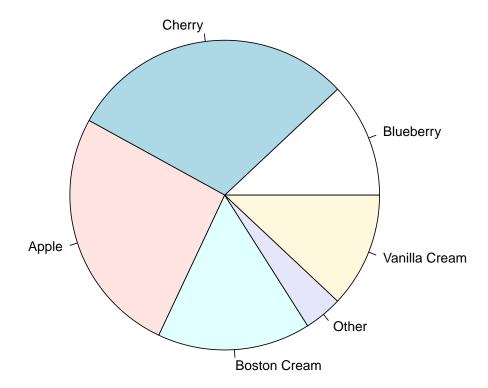
6 pie

require(grDevices)

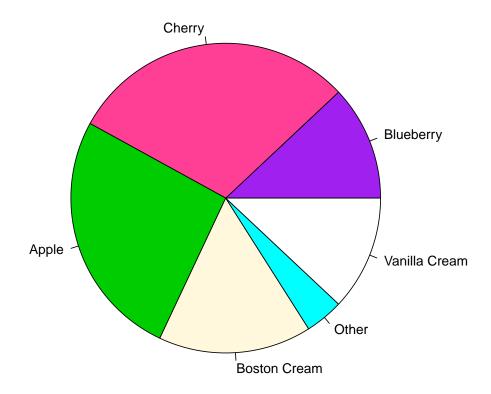
pie(rep(1, 24), col = rainbow(24), radius = 0.9)

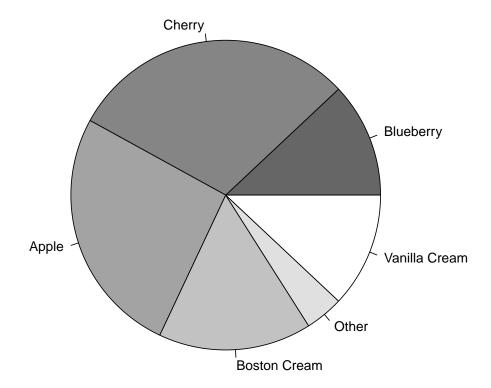


```
pie(pie.sales) # default colours
```

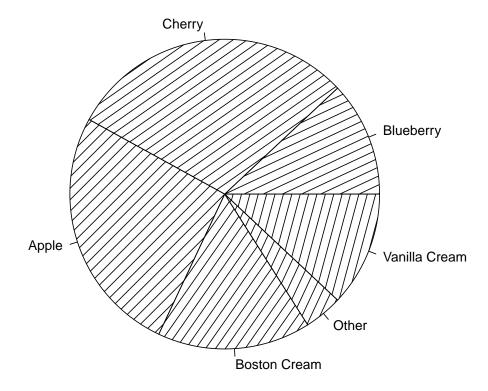


```
pie(pie.sales,
     col = c("purple", "violetred1", "green3", "cornsilk", "cyan", "white"))
```



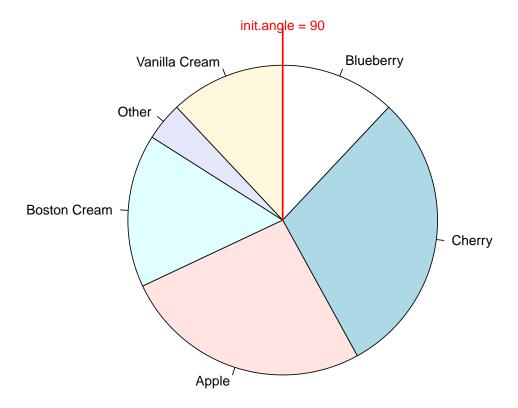


pie(pie.sales, density = 10, angle = 15 + 10 * 1:6)



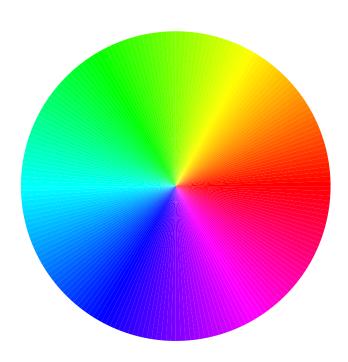
```
pie(pie.sales, clockwise=TRUE, main="pie(*, clockwise=TRUE)")
segments(0,0, 0,1, col= "red", lwd = 2)
text(0,1, "init.angle = 90", col= "red")
```

pie(*, clockwise=TRUE)



```
n <- 200
pie(rep(1,n), labels="", col=rainbow(n), border=NA,
    main = "Rainbow")</pre>
```

Rainbow



7 Tableaux de contingence

7.1 balloonplot

```
library(gdata)
## gdata: read.xls support for 'XLS' (Excel 97-2004) files ENABLED.
##
## gdata: read.xls support for 'XLSX' (Excel 2007+) files ENABLED.
##
## Attaching package: 'gdata'
## The following object is masked from 'package:stats':
##
## nobs
```

```
## The following object is masked from 'package:utils':
##

## object.size
## The following object is masked from 'package:base':
##

## startsWith

library(gtools)
library(gplots)

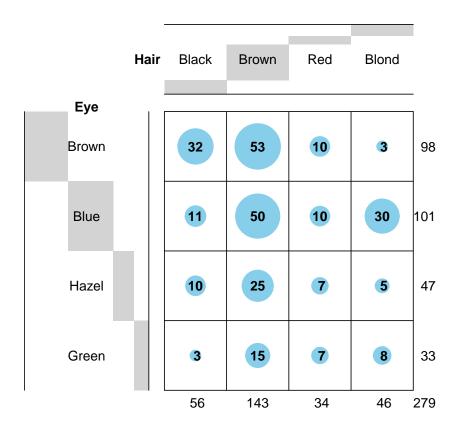
##

## Attaching package: 'gplots'
## The following object is masked from 'package:stats':
##

## lowess
```

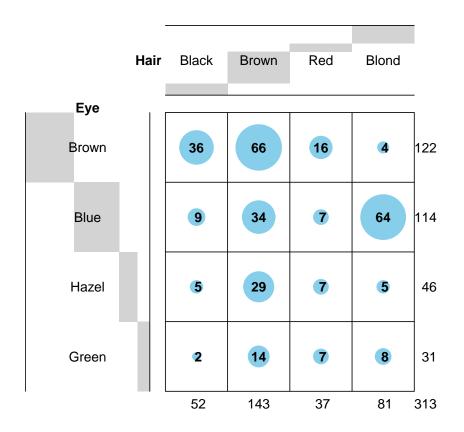
```
balloonplot(as.table(HairEyeColor[,,Sex="Male"]), dotsize = 10)
```

Balloon Plot for x by y. Area is proportional to Freq.



balloonplot(as.table(HairEyeColor[,,Sex="Female"]), dotsize = 10)

Balloon Plot for x by y. Area is proportional to Freq.



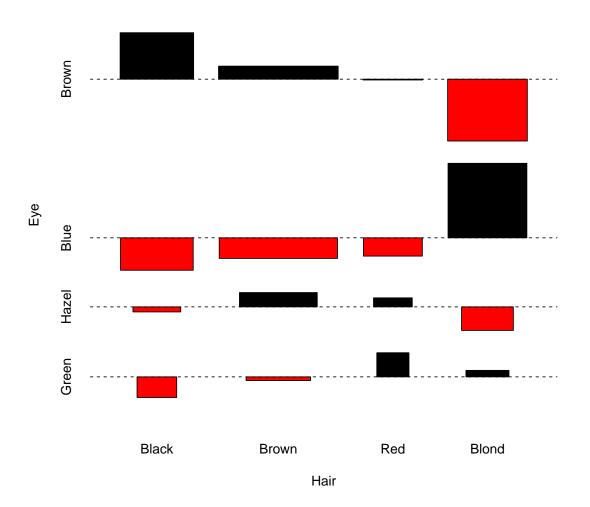
7.2 assocplot

Aggregate over sex :

```
x <- margin.table(HairEyeColor, c(1, 2))
x</pre>
```

```
assocplot(x, main = "Relation between hair and eye color")
```

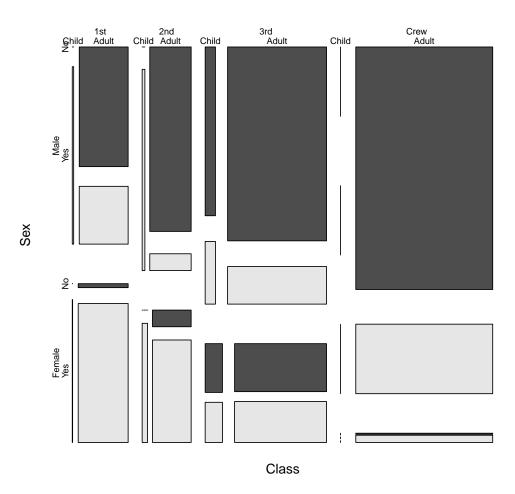
Relation between hair and eye color



7.3 mosaicplot

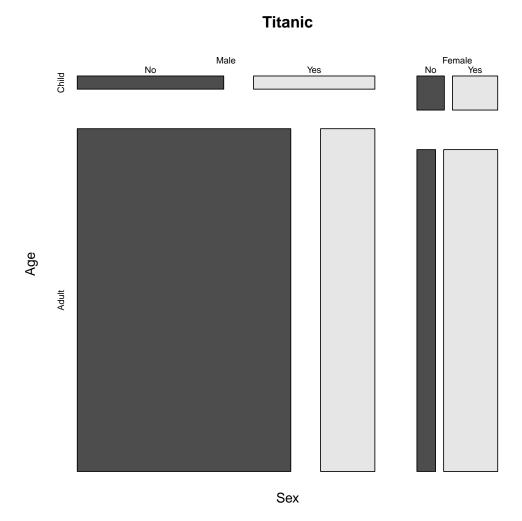
mosaicplot(Titanic, main = "Survival on the Titanic", color = TRUE)

Survival on the Titanic



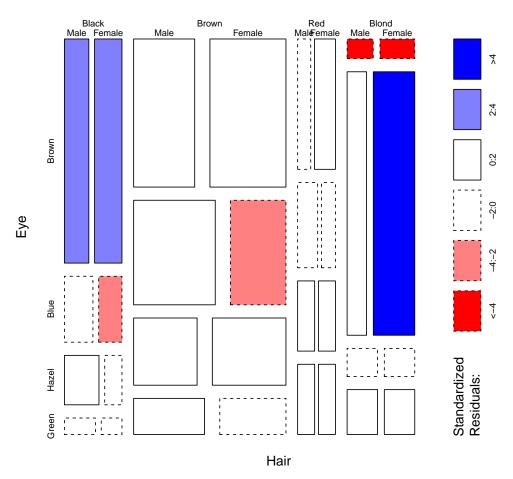
Formula interface for tabulated data :

mosaicplot(~ Sex + Age + Survived, data = Titanic, color = TRUE)



mosaicplot(HairEyeColor, shade = TRUE)

HairEyeColor

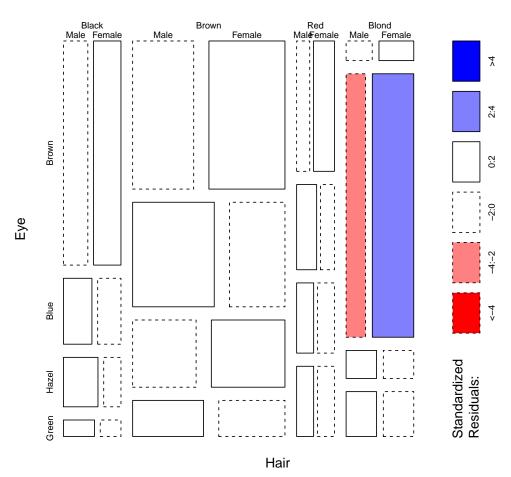


Independence model of hair and eye color and sex. Indicates that ## there are more blue eyed blonde females than expected in the case ## of independence and too few brown eyed blonde females. ## The corresponding model is :

```
fm <- loglin(HairEyeColor, list(1, 2, 3))
pchisq(fm$pearson, fm$df, lower.tail = FALSE)</pre>
```

```
mosaicplot(HairEyeColor, shade = TRUE, margin = list(1:2, 3))
```

HairEyeColor

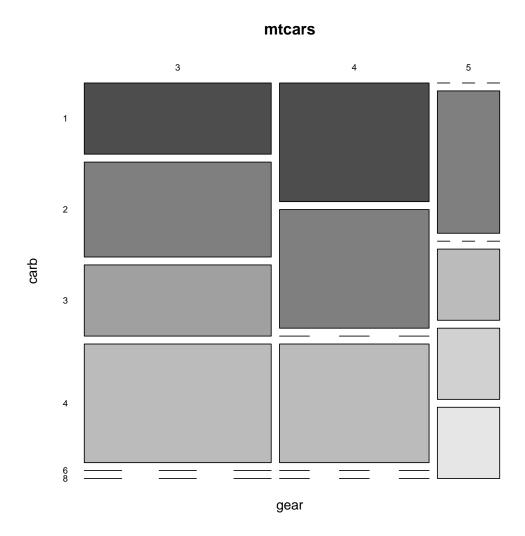


Model of joint independence of sex from hair and eye color. Males ## are underrepresented among people with brown hair and eyes, and are ## overrepresented among people with brown hair and blue eyes. ## The corresponding model is :

```
fm <- loglin(HairEyeColor, list(1:2, 3))
pchisq(fm$pearson, fm$df, lower.tail = FALSE)</pre>
```

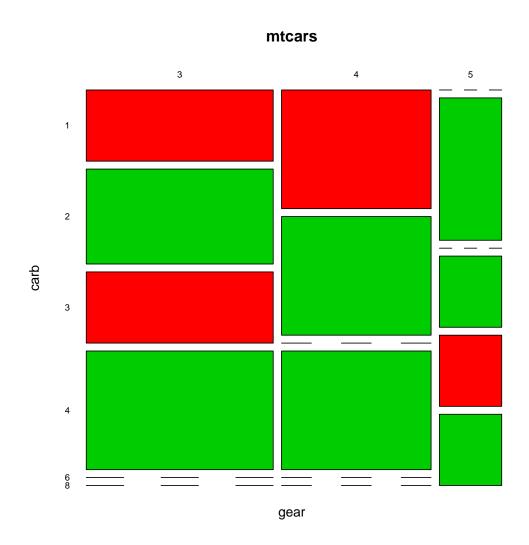
Formula interface for raw data : visualize cross-tabulation of numbers ## of gears and carburettors in Motor Trend car data.

```
mosaicplot(~ gear + carb, data = mtcars, color = TRUE, las = 1)
```



color recycling

```
mosaicplot(~ gear + carb, data = mtcars, color = 2:3, las = 1)
```

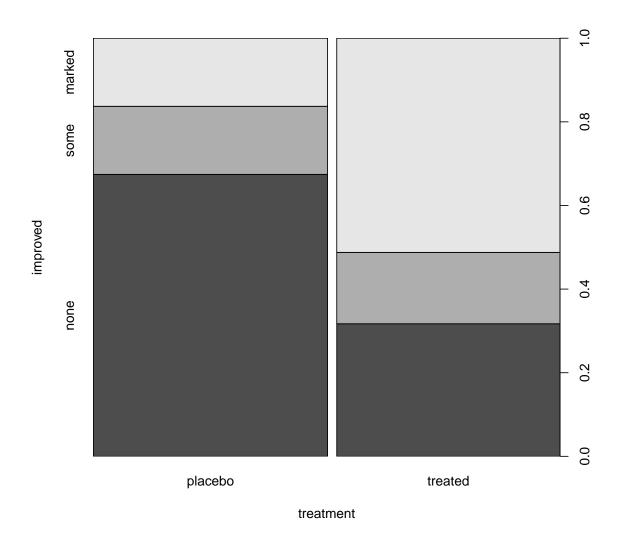


7.4 splineplot

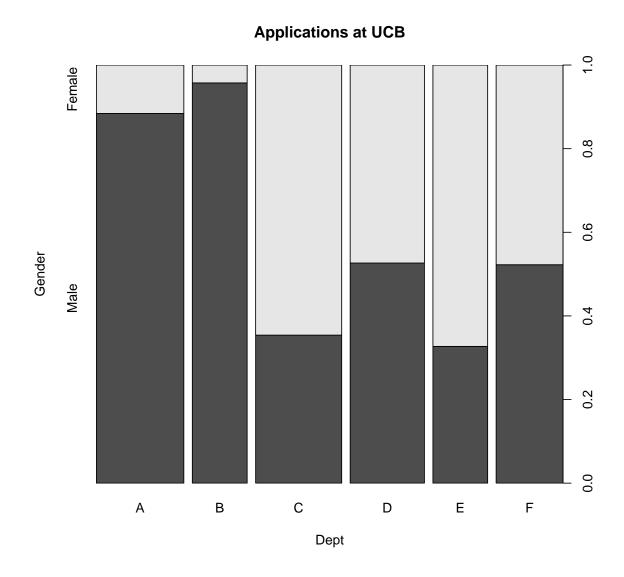
treatment and improvement of patients with rheumatoid arthritis

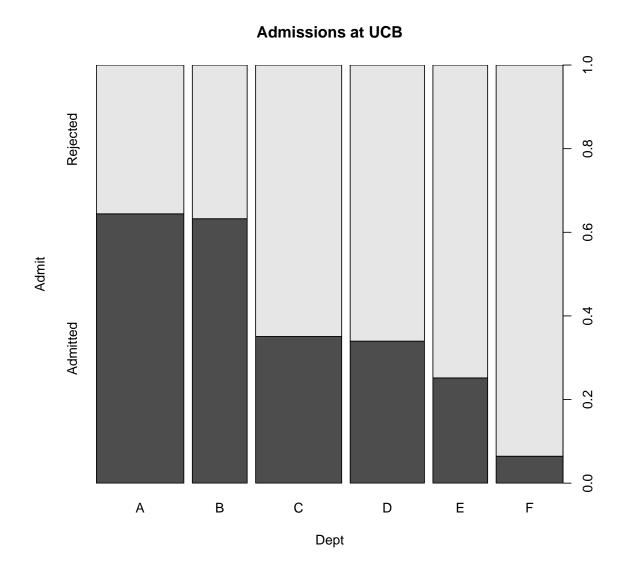
(dependence on a categorical variable)

```
(spineplot(improved ~ treatment))
```



applications and admissions by department at UC Berkeley ## (two-way tables)

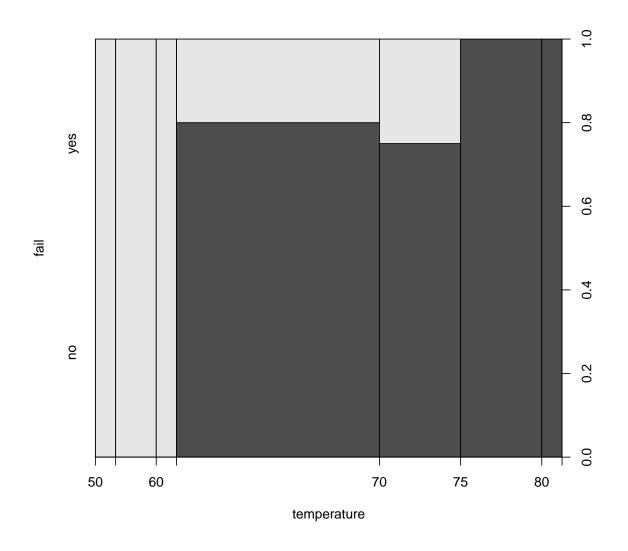




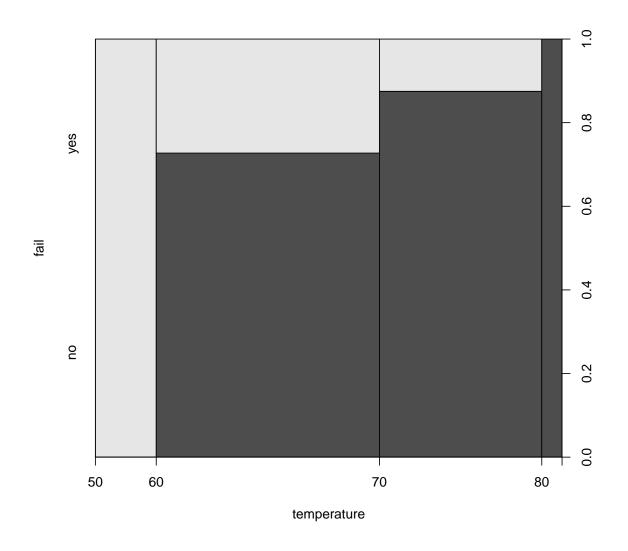
NASA space shuttle o-ring failures

(dependence on a numerical variable)

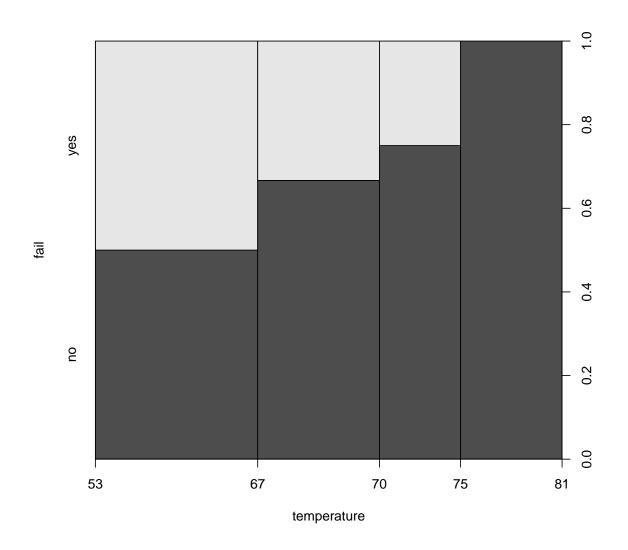
```
(spineplot(fail ~ temperature))
```



(spineplot(fail ~ temperature, breaks = 3))

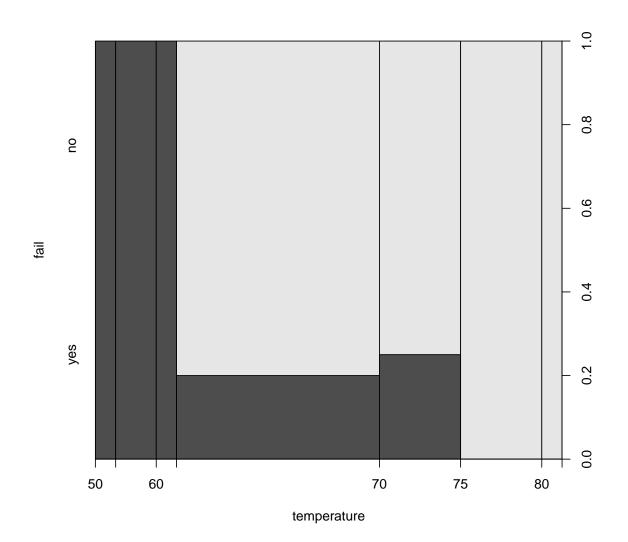


(spineplot(fail ~ temperature, breaks = quantile(temperature)))



highlighting for failures

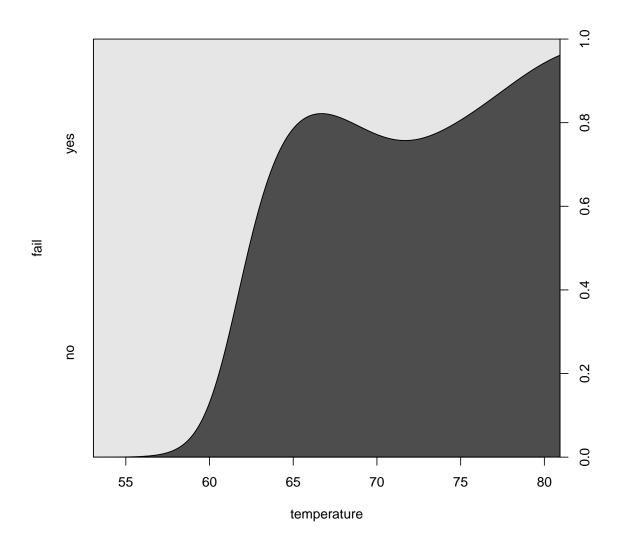
```
spineplot(fail ~ temperature, ylevels = 2:1)
```



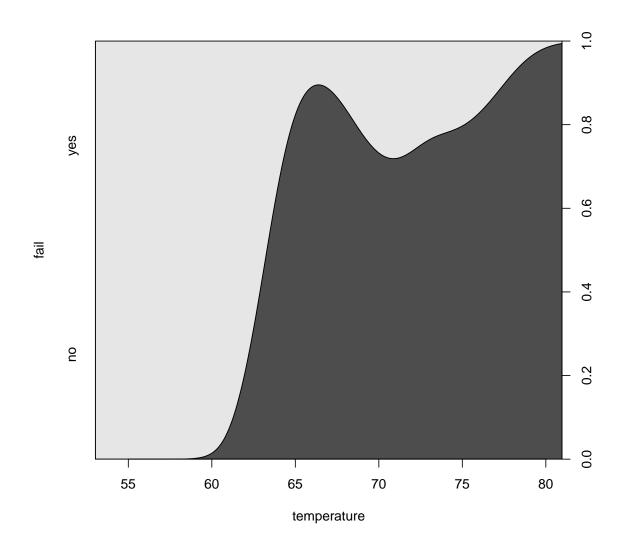
7.5 cdplot (Conditional Density Plots)

NASA space shuttle o-ring failures

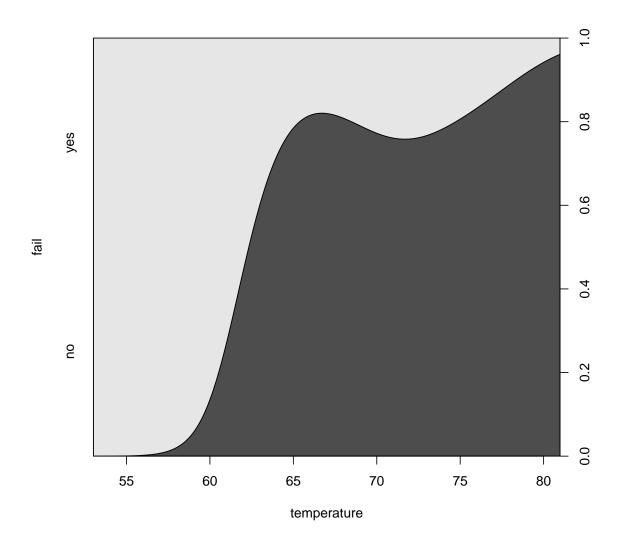
```
cdplot(fail ~ temperature)
```



cdplot(fail ~ temperature, bw = 2)

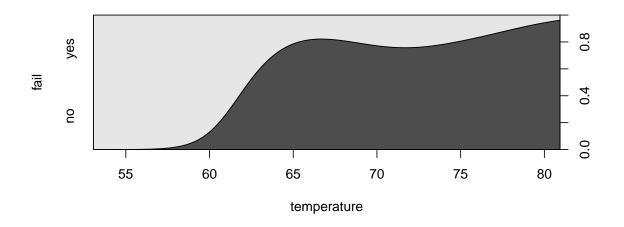


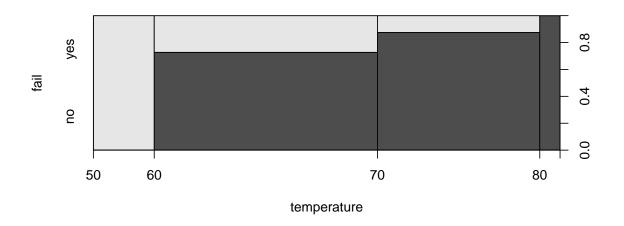
cdplot(fail ~ temperature, bw = "SJ")



compare with spinogram on the same graph

```
layout(1:2)
cdplot(fail ~ temperature)
(spineplot(fail ~ temperature, breaks = 3))
```

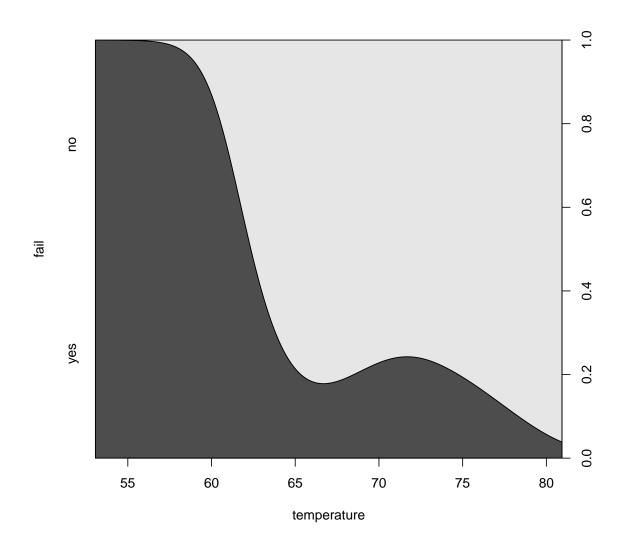




layout(1)

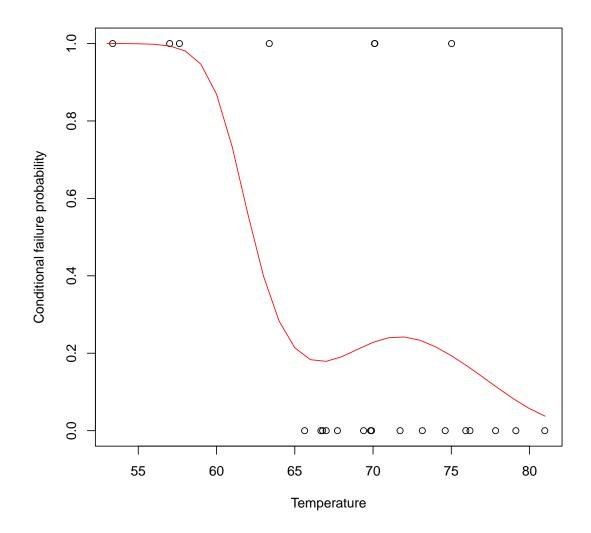
highlighting for failures

cdplot(fail ~ temperature, ylevels = 2:1)



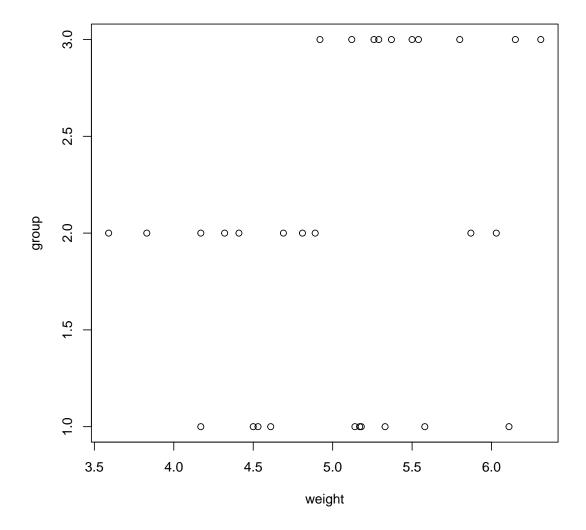
scatter plot with conditional density

```
cdens <- cdplot(fail ~ temperature, plot = FALSE)</pre>
```

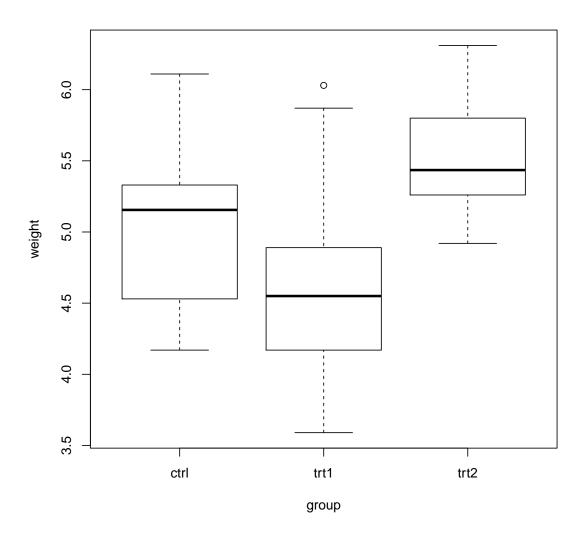


8 Plot factor variables

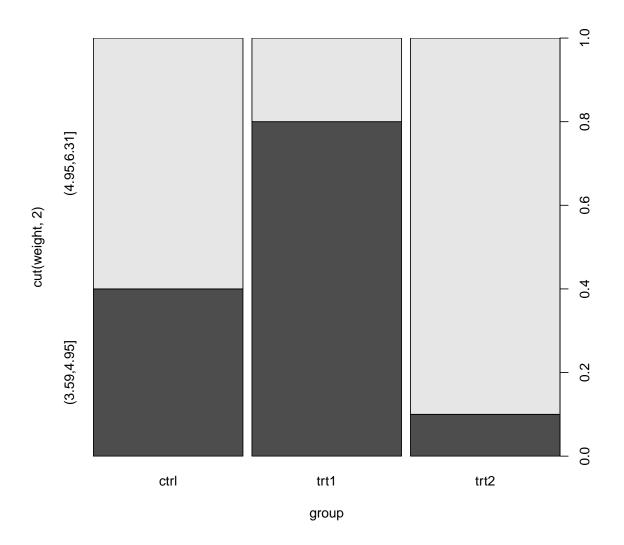
```
require(grDevices)
plot(PlantGrowth) # -> plot.data.frame
```



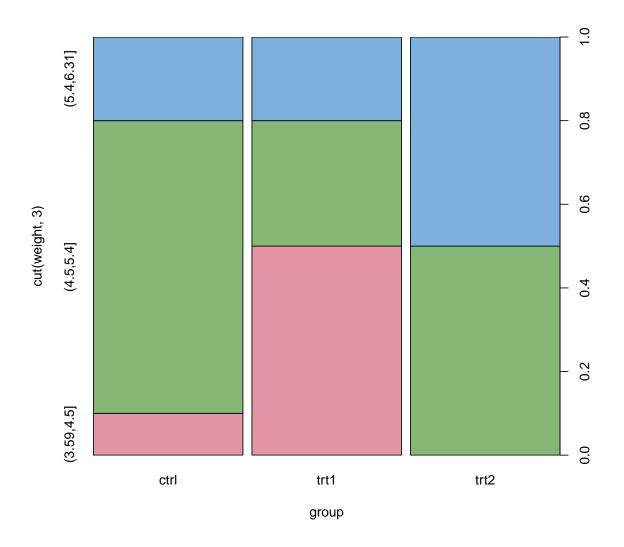
plot(weight ~ group, data = PlantGrowth) # numeric vector ~ factor



plot(cut(weight, 2) ~ group, data = PlantGrowth) # factor ~ factor

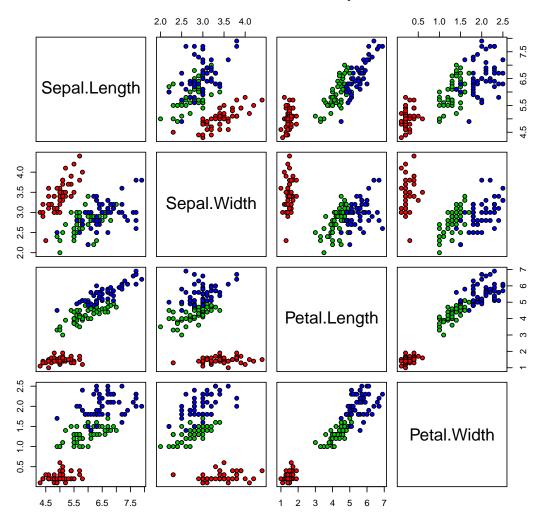


passing "..." to spineplot() eventually :



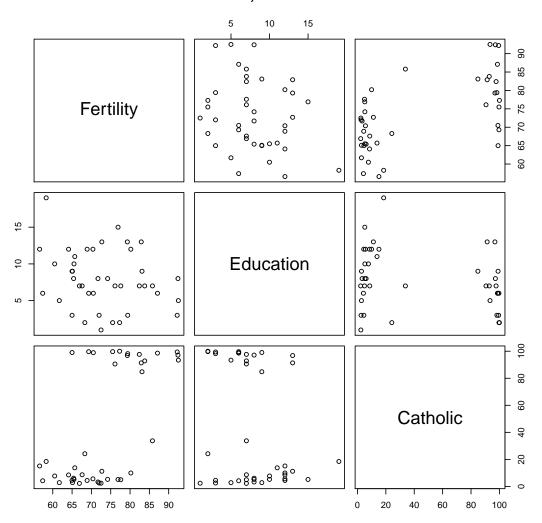
9 Matrix plot

Anderson's Iris Data -- 3 species

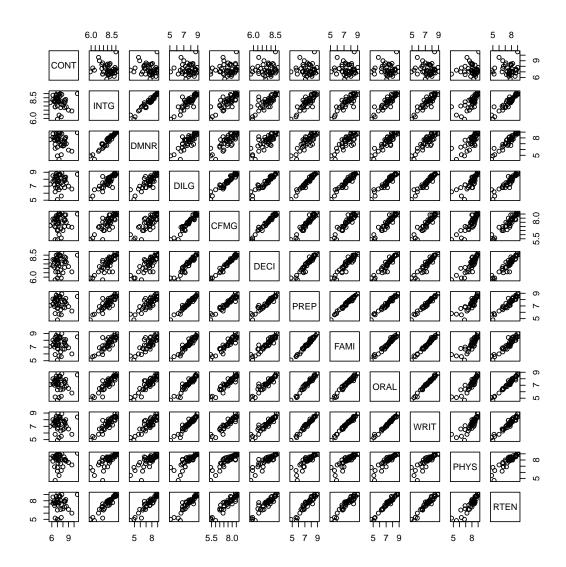


formula method

Swiss data, Education < 20



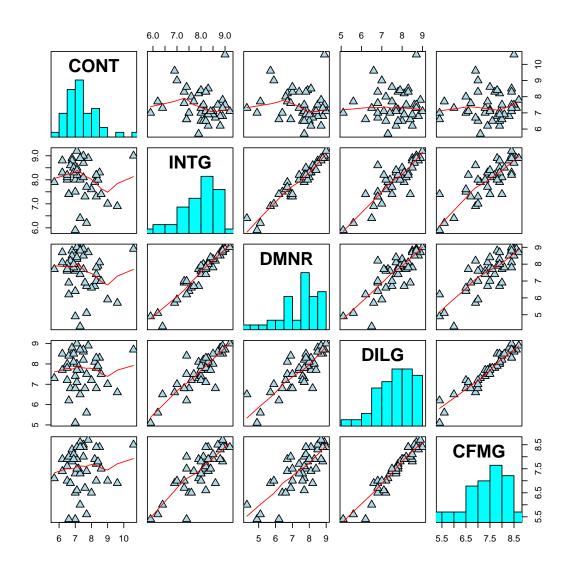
pairs(USJudgeRatings)



put histograms on the diagonal

```
panel.hist <- function(x, ...)
{
    usr <- par("usr"); on.exit(par(usr))
    par(usr = c(usr[1:2], 0, 1.5) )
    h <- hist(x, plot = FALSE)
    breaks <- h$breaks; nB <- length(breaks)
    y <- h$counts; y <- y/max(y)
    rect(breaks[-nB], 0, breaks[-1], y, col="cyan", ...)
}</pre>
```

```
pairs(USJudgeRatings[1:5], panel=panel.smooth,
    cex = 1.5, pch = 24, bg="light blue",
    diag.panel=panel.hist, cex.labels = 2, font.labels=2)
```

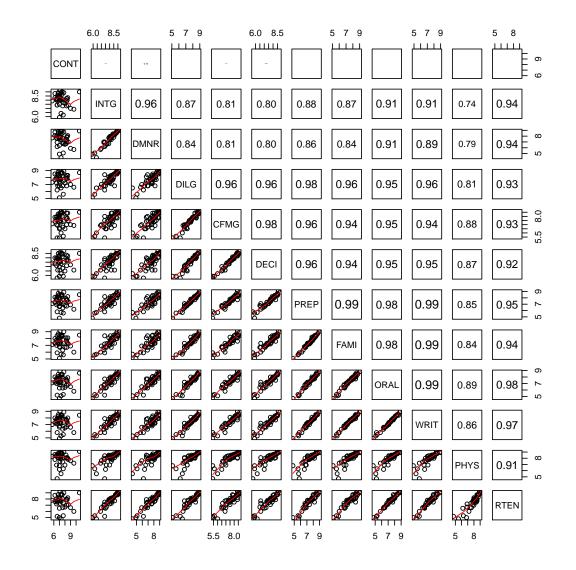


put (absolute) correlations on the upper panels, with size proportional to the correlations.

```
panel.cor <- function(x, y, digits=2, prefix="", cex.cor)
{
    usr <- par("usr"); on.exit(par(usr))
    par(usr = c(0, 1, 0, 1))
    r <- abs(cor(x, y))
    txt <- format(c(r, 0.123456789), digits=digits)[1]
    txt <- paste(prefix, txt, sep="")</pre>
```

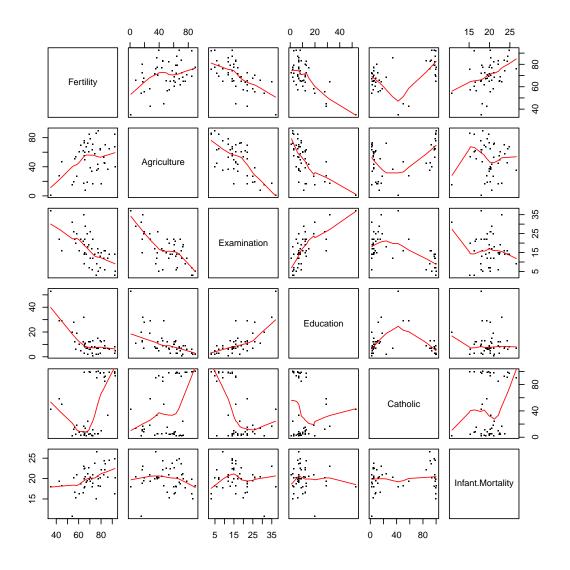
```
if(missing(cex.cor)) cex.cor <- 0.8/strwidth(txt)
text(0.5, 0.5, txt, cex = cex.cor * r)
}</pre>
```

pairs(USJudgeRatings, lower.panel=panel.smooth, upper.panel=panel.cor)

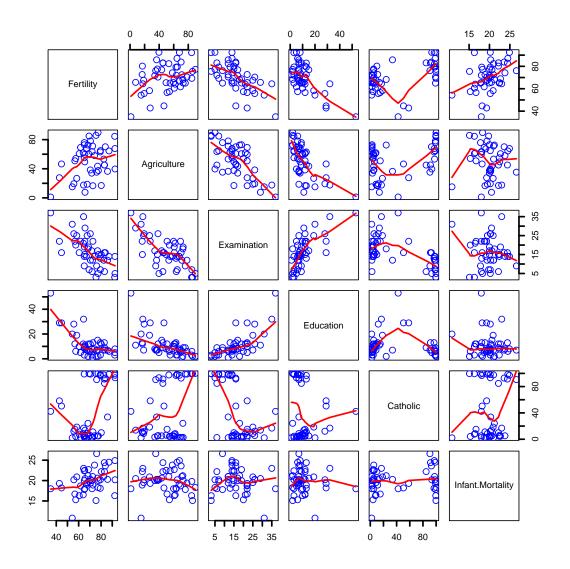


10 Simplepanel plots

pairs(swiss, panel = panel.smooth, pch = ".") # emphasize the smooths



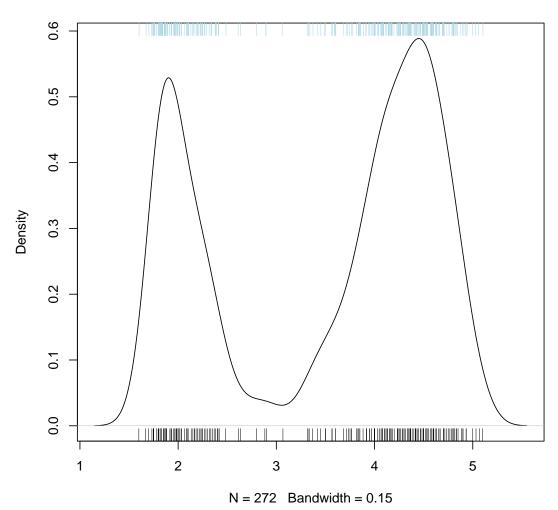
pairs(swiss, panel = panel.smooth, lwd = 2, cex= 1.5, col="blue") # hmm...



11 jitter

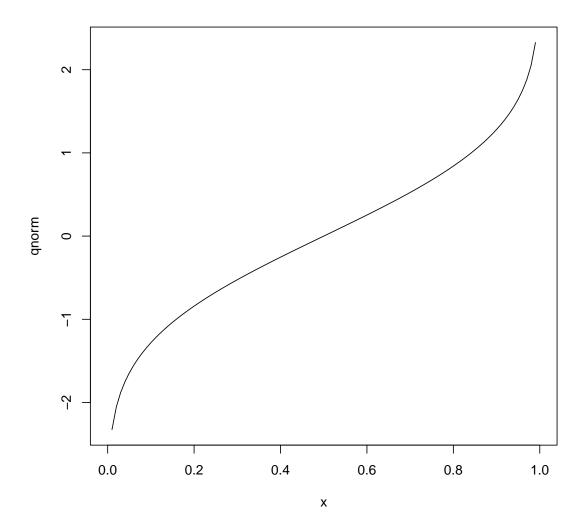
```
require(stats) # both 'density' and its default method
with(faithful, {
    plot(density(eruptions, bw = 0.15))
    rug(eruptions)
    rug(jitter(eruptions, amount = 0.01), side = 3, col = "light blue")
})
```

density.default(x = eruptions, bw = 0.15)



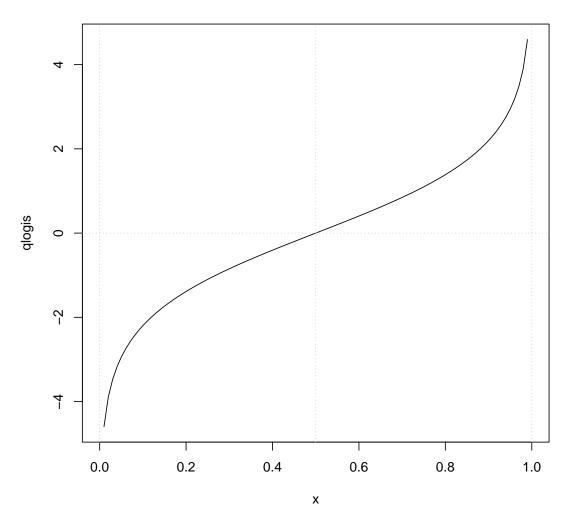
12 curves

plot(qnorm)

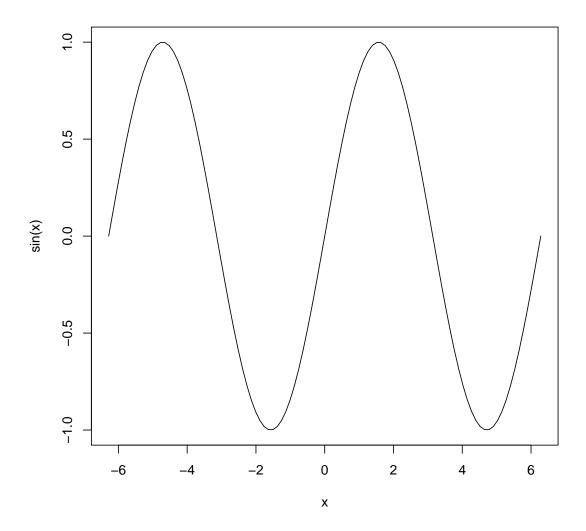


```
plot(qlogis, main = "The Inverse Logit : qlogis()")
abline(h=0, v=0:2/2, lty=3, col="gray")
```

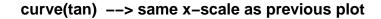
The Inverse Logit : qlogis()

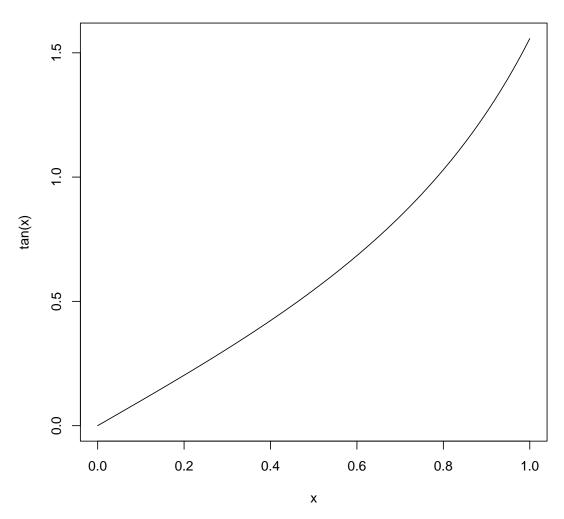


curve(sin, -2*pi, 2*pi)



curve(tan, main = "curve(tan) --> same x-scale as previous plot")

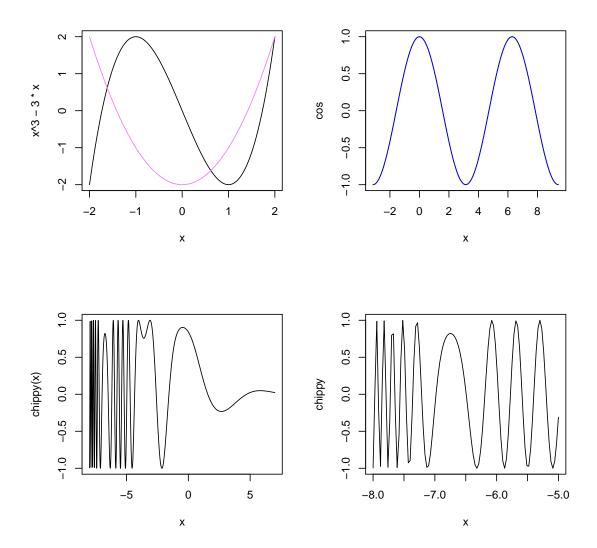




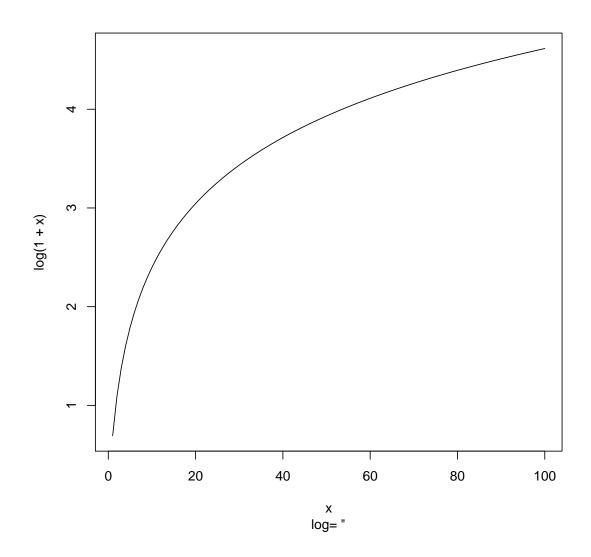
```
op <- par(mfrow=c(2,2))
curve(x^3-3*x, -2, 2)
curve(x^2-2, add = TRUE, col = "violet")

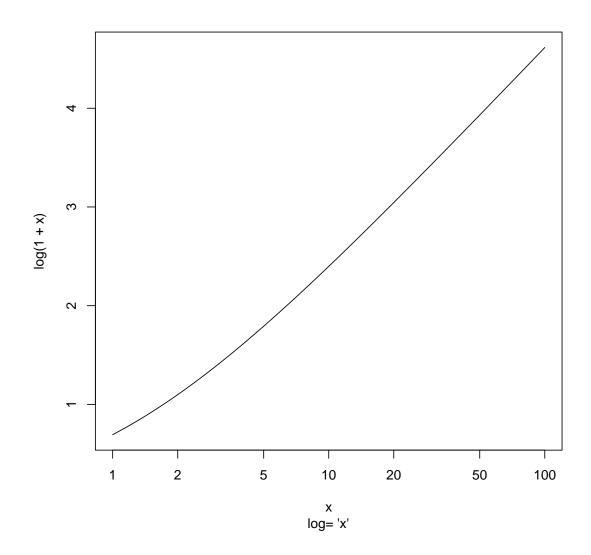
plot(cos, -pi, 3*pi)
plot(cos, xlim = c(-pi,3*pi), n = 1001, col = "blue", add=TRUE)

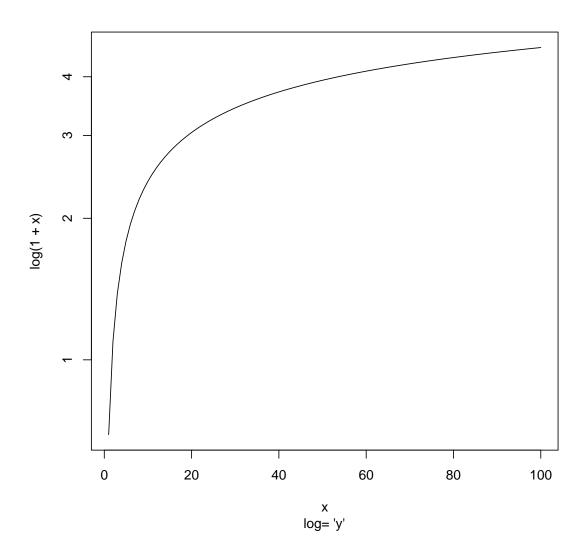
chippy <- function(x) sin(cos(x)*exp(-x/2))
curve(chippy, -8, 7, n=2001)
plot (chippy, -8, -5)</pre>
```

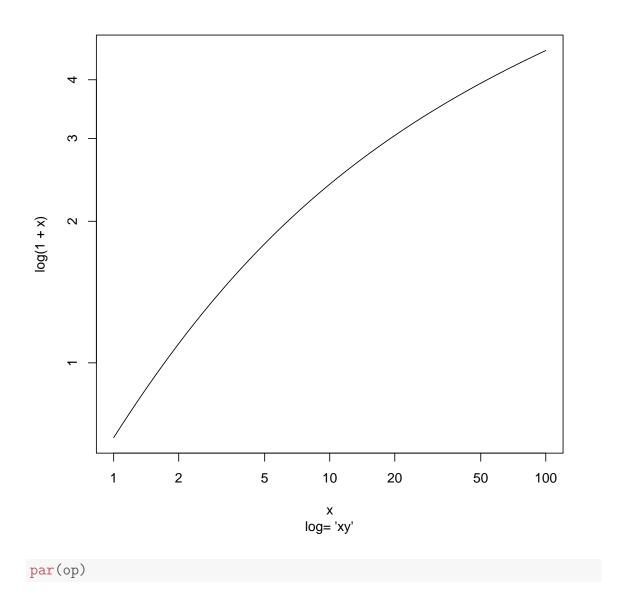


```
for(ll in c("","x","y","xy"))
    curve(log(1+x), 1,100, log=ll, sub=paste("log= '",ll,"'",sep=""))
```









13 loess (regression non-paramétrique)

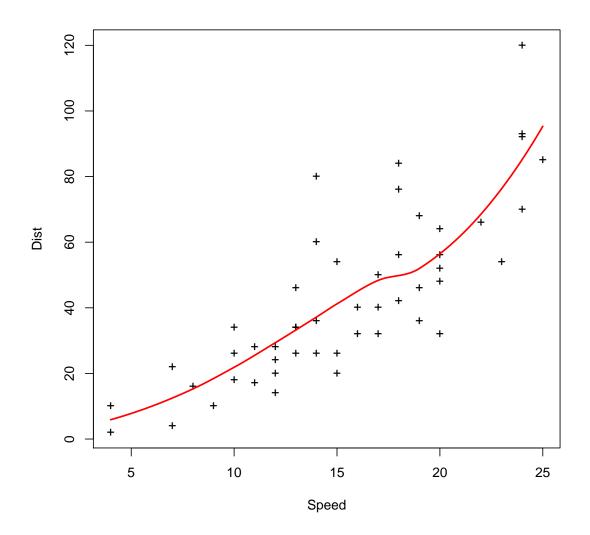
```
## 15 16 17 18 19 20 21
## 51.986702 56.461318 61.959729 68.569313 76.316068 85.212121 95.324047
       22
               23
                      24
                              25
                                      26
##
       NA
               NA
                      NA
                              NA
                                      NA
##
## $se.fit
    1
             2 3 4 5
## 7.568120 5.945831 4.990827 4.545284 4.308639 4.115049 3.789542 3.716231
            10 11 12 13
      9
                                        14
                                               15
## 3.776947 4.091747 4.709568 4.245427 4.035929 3.753410 4.004705 4.043190
## 17 18 19 20 21
                                        22
                                               23
                                                      24
## 4.026105 4.074664 4.570818 5.954217 8.302014
                                               NA
                                                      NA
                                        NA
      25
             26
##
      NA
             NA
##
## $residual.scale
## [1] 15.29496
##
## $df
## [1] 44.6179
```

to allow extrapolation

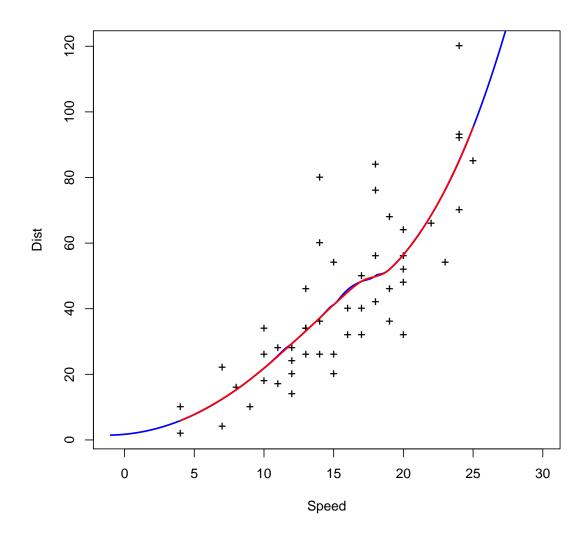
```
cars.lo2 <- loess(dist ~ speed, cars,</pre>
control = loess.control(surface = "direct"))
predict(cars.lo2, data.frame(speed = seq(5, 30, 1)), se = TRUE)
## $fit
##
    1
             2
                    3
                                 4
                                         5
            9.926596 12.442424 15.281082
   7.741006
                                    18.425712 21.865315
            8
                   9 10
##
        7
                                        11
##
  25.713413 29.350386
                   33.230660 37.167935 41.205226 45.781544
                    15 16
              14
                                     17
##
  48.355889 50.067148 51.986702 56.445263 62.025404 68.569313
                20
                   21 22
                                         23
##
       19
##
  76.193111 85.053364 95.300523 106.974661 120.092581 134.665851
   25
##
## 150.698545 168.190283
##
## $se.fit
                   3
                          4 5
               2
##
        1
## 7.565991 5.959097 5.012013 4.550013 4.321596 4.119331 3.939804
               9
                  10 11 12
                                         13
                                                     14
  3.720098 3.780877 4.096004 4.714469 4.398936 4.040129
##
                                                4.184257
## 15 16 17 18 19
```

```
## 4.008873 4.061865 4.033998 4.078904 4.584606 5.952480 8.306901
## 22 23 24 25 26
## 11.601911 15.792480 20.864660 26.823827 33.683999
##
## $residual.scale
## [1] 15.31087
##
## $df
## [1] 44.55085
```

```
plot(cars.lo, xlab="Speed", ylab="Dist", pch="+")
lines(seq(min(cars$speed), max(cars$speed), 0.1),predict(cars.lo, data.frame(speed))
```



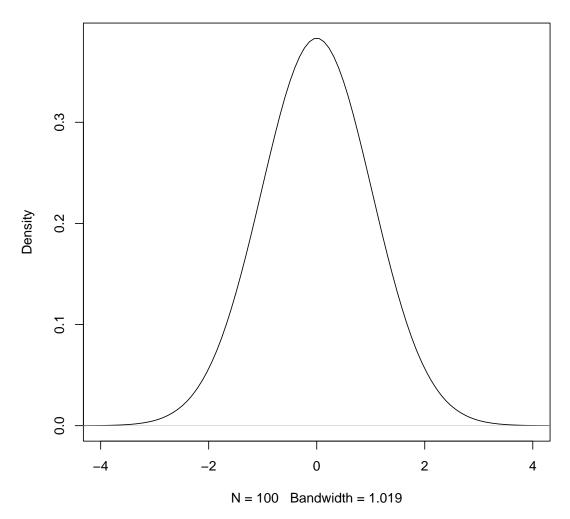
```
plot(cars.lo2, xlab="Speed", ylab="Dist", pch="+", xlim=c(min(cars$speed)-5,max(cars
lines(seq(min(cars$speed)-5, max(cars$speed)+5, 0.1),predict(cars.lo2, data.frame(speed))
lines(seq(min(cars$speed)-5, max(cars$speed)+5, 0.1),predict(cars.lo, data.frame(speed))
```



14 density estimation

```
require(graphics)
plot(density(c(-20,rep(0,98),20)), xlim = c(-4,4))# IQR = 0
```

density.default(x = c(-20, rep(0, 98), 20))

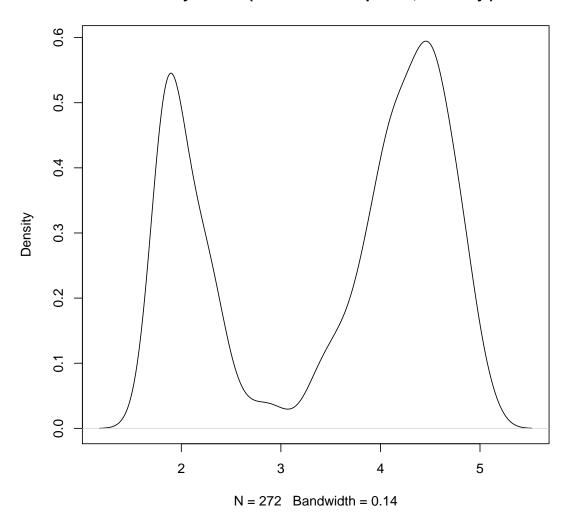


```
# The Old Faithful geyser data
d <- density(faithful$eruptions, bw = "sj")</pre>
d
##
## Call:
   density.default(x = faithful\$eruptions, bw = "sj")
## Data: faithful$eruptions (272 obs.); Bandwidth 'bw' = 0.14
##
##
          X
## Min.
         :1.180
                   Min.
                          :0.0001834
   1st Qu.:2.265 1st Qu.:0.0422638
##
## Median :3.350 Median :0.1709243
```

```
## Mean :3.350 Mean :0.2301726
## 3rd Qu.:4.435 3rd Qu.:0.4134348
## Max. :5.520 Max. :0.5945634
```

plot(d)

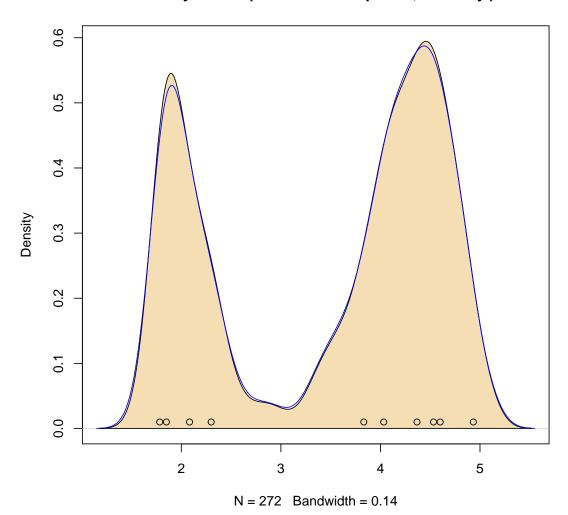
density.default(x = faithful\$eruptions, bw = "sj")



```
plot(d, type = "n")
polygon(d, col = "wheat")
## Missing values:
x <- xx <- faithful$eruptions
x[i.out <- sample(length(x), 10)] <- NA</pre>
```

```
doR <- density(x, bw = 0.15, na.rm = TRUE)
lines(doR, col = "blue")
points(xx[i.out], rep(0.01, 10))</pre>
```

density.default(x = faithful\$eruptions, bw = "sj")



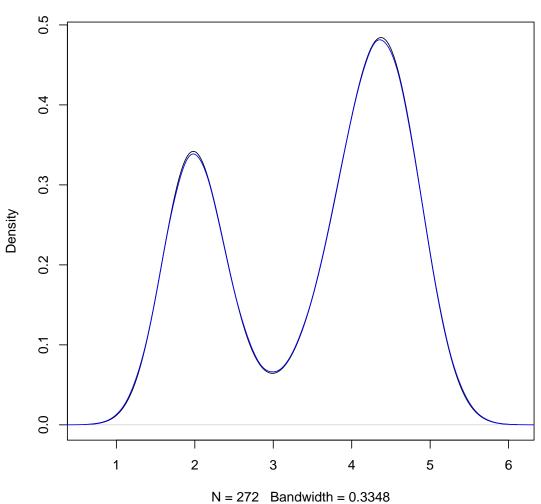
```
## Weighted observations:
fe <- sort(faithful$eruptions) # has quite a few non-unique values
## use 'counts / n' as weights:
dw <- density(unique(fe), weights = table(fe)/length(fe), bw = d$bw)
utils::str(dw) ## smaller n: only 126, but identical estimate:
## List of 7
## $ x : num [1:512] 1.18 1.19 1.2 1.21 1.21 ...
## $ y : num [1:512] 0.000183 0.000223 0.00027 0.000328 0.000397 ...</pre>
```

```
## $ bw : num 0.14
## $ n : int 126
## $ call : language density.default(x = unique(fe), bw = d$bw, weights = tabl
## $ data.name: chr "unique(fe)"
## $ has.na : logi FALSE
## - attr(*, "class") = chr "density"

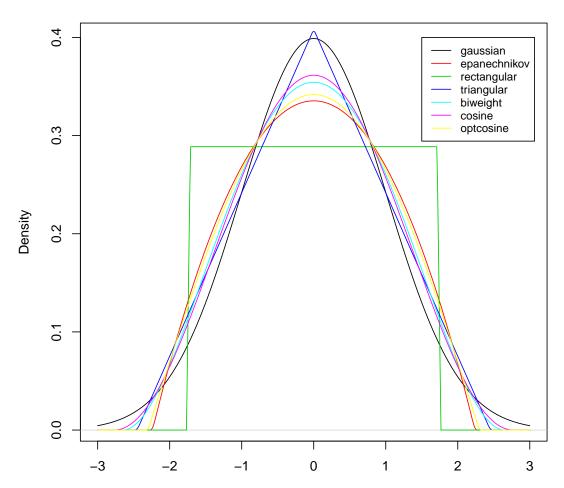
stopifnot(all.equal(d[1:3], dw[1:3]))
```

```
## simulation from a density() fit:
# a kernel density fit is an equally-weighted mixture.
fit <- density(xx)
N <- 1e6
x.new <- rnorm(N, sample(xx, size = N, replace = TRUE), fit$bw)
plot(fit)
lines(density(x.new), col="blue")</pre>
```



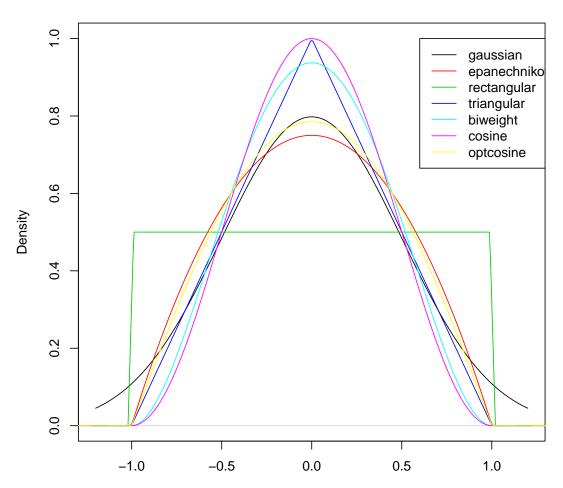






```
## show the kernels in the S parametrization
plot(density(0, from=-1.2, to=1.2, width=2, kernel="gaussian"), type="l",
        ylim = c(0, 1), xlab="", main="R's density() kernels with width = 1")
for(i in 2:length(kernels))
    lines(density(0, width = 2, kernel = kernels[i]), col = i)
legend(0.6, 1.0, legend = kernels, col = seq(kernels), lty = 1)
```

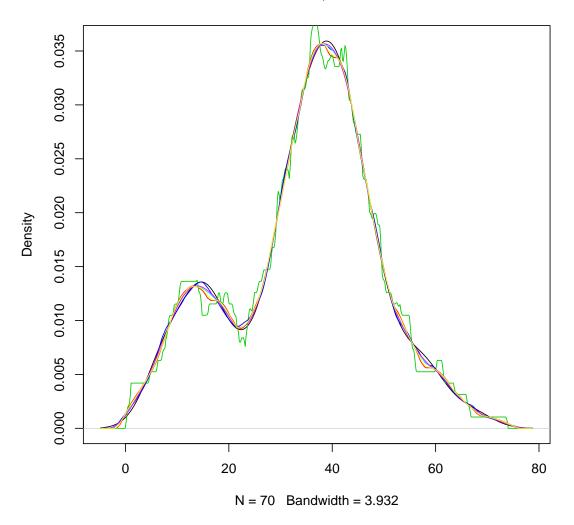




```
##----- Semi-advanced theoretic from here on ----
(RKs <- cbind(sapply(kernels,
                     function(k) density(kernel = k, give.Rkern = TRUE))))
##
                     [,1]
## gaussian
                0.2820948
## epanechnikov 0.2683282
## rectangular 0.2886751
## triangular
                0.2721655
## biweight
                0.2699746
## cosine
                0.2711340
## optcosine
                0.2684756
100*round(RKs["epanechnikov",]/RKs, 4) ## Efficiencies
```

```
## gaussian 95.12
## epanechnikov 100.00
## rectangular 92.95
## triangular 98.59
## biweight 99.39
## cosine 98.97
## optcosine 99.95
```

same sd bandwidths, 7 different kernels

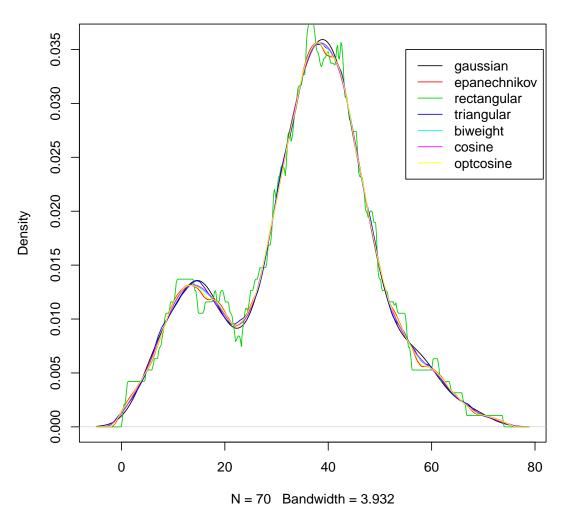


```
## Bandwidth Adjustment for "Exactly Equivalent Kernels"
h.f <- sapply(kernels, function(k)density(kernel = k, give.Rkern = TRUE))
(h.f <- (h.f["gaussian"] / h.f)^ .2)

## gaussian epanechnikov rectangular triangular biweight
## 1.0000000 1.0100567 0.9953989 1.0071923 1.0088217
## cosine optcosine
## 1.0079575 1.0099458

## -> 1, 1.01, .995, 1.007,... close to 1 => adjustment barely visible..
```

equivalent bandwidths, 7 different kernels



15 Radar plots

```
library(ggplot2)
if(!require("ggradar")){
install.packages("extrafont")
download.file("https://dl.dropboxusercontent.com/u/2364714/airbnb_ttf_fonts/
Circular Air-Light 3.46.45 PM.ttf",
   "/Library/Fonts/Circular Air-Light 3.46.45 PM.ttf", method="curl")
extrafont::font_import(pattern = 'Circular', prompt=FALSE)
devtools::install_github("ricardo-bion/ggradar", dependencies=TRUE)
library(ggradar)
}
library(ggradar)
```

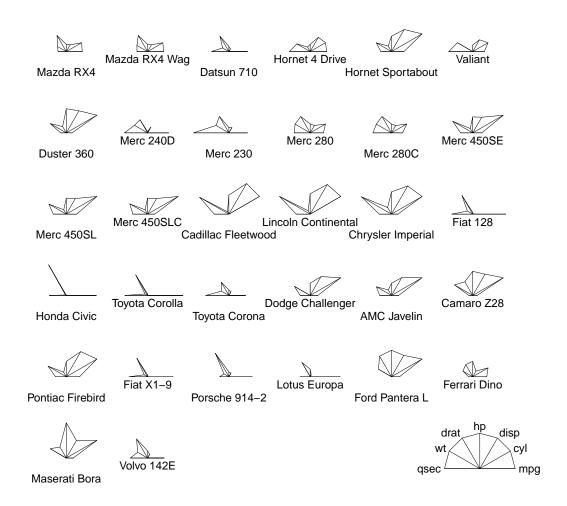
```
suppressPackageStartupMessages(library(dplyr))
library(scales)

mtcars %>%
    add_rownames( var = "group" ) %>%
    mutate_each(funs(rescale), -group) %>%
    tail(4) %>% select(1:10) -> mtcars_radar

print(ggradar(mtcars_radar))
```

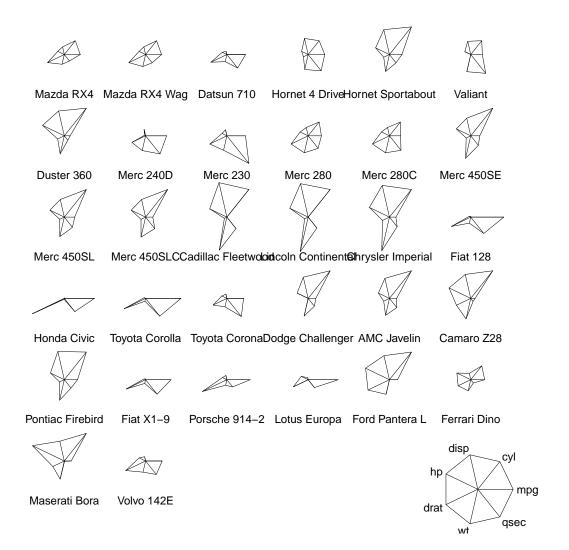
```
require(grDevices)
stars(mtcars[, 1:7], key.loc = c(14, 2),
    main = "Motor Trend Cars : stars(*, full = F)", full = FALSE)
```

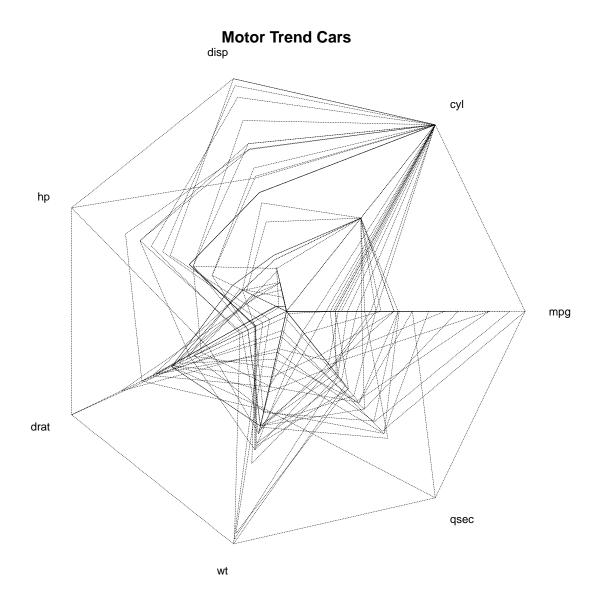
Motor Trend Cars : stars(*, full = F)



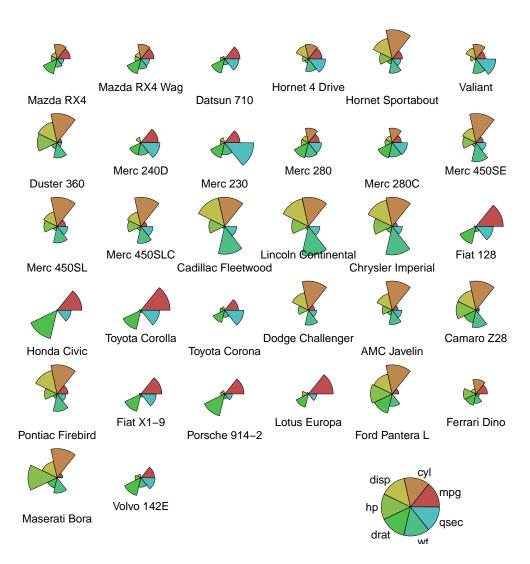
```
stars(mtcars[, 1:7], key.loc = c(14, 1.5),
    main = "Motor Trend Cars : full stars()",flip.labels=FALSE)
```

Motor Trend Cars : full stars()



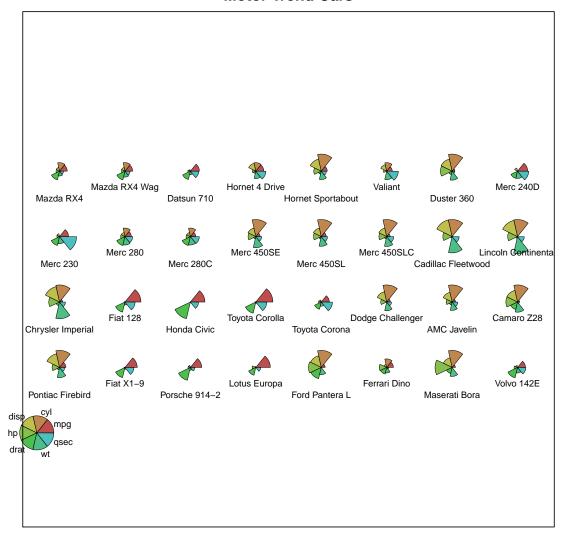


Motor Trend Cars

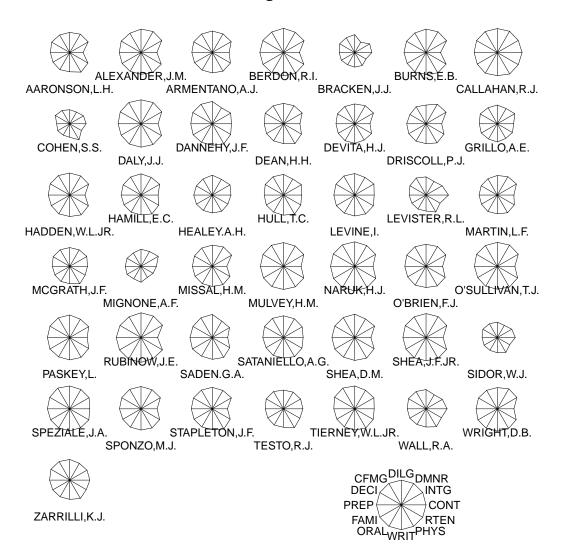


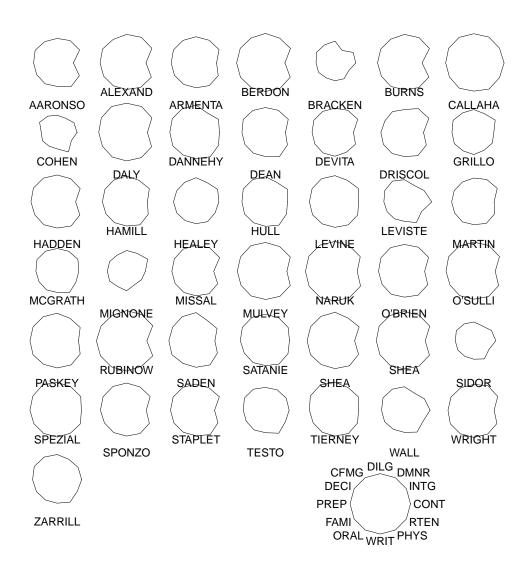
```
stars(mtcars[, 1:7], len = 0.6, key.loc = c(1.5, 0),
    main = "Motor Trend Cars", draw.segments = TRUE,
    frame.plot=TRUE, nrow = 4, cex = .7)
```

Motor Trend Cars

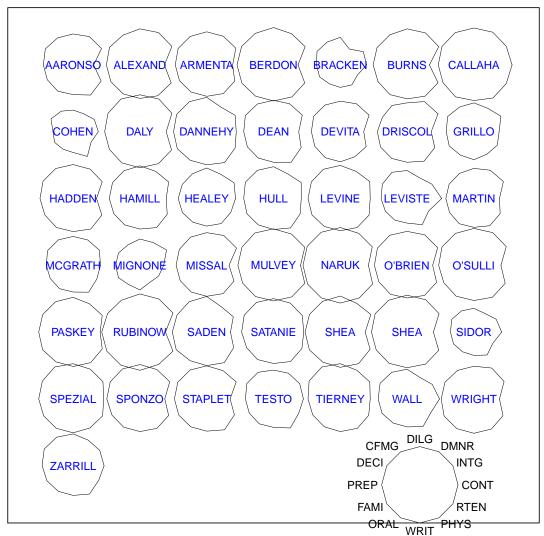


Judge not ...

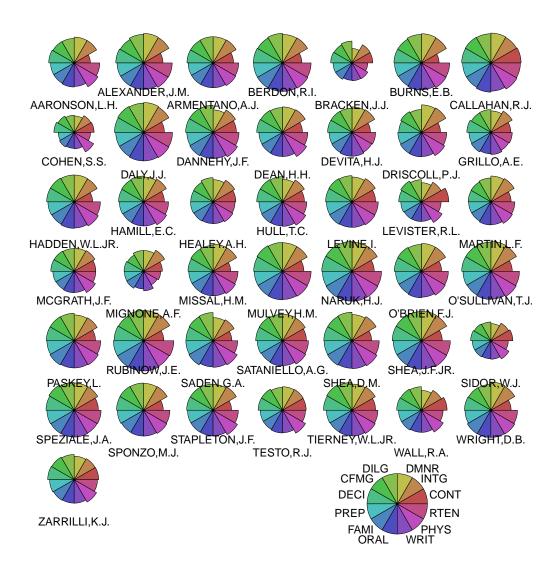




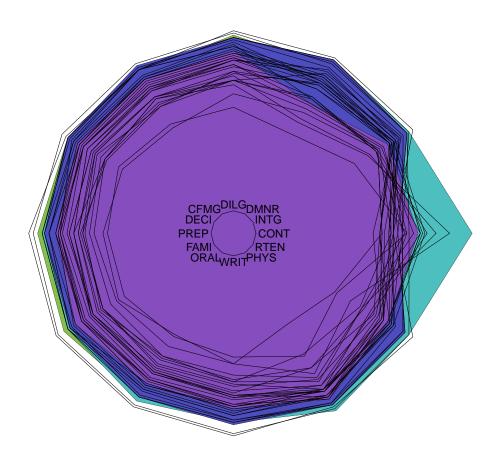
Judge not ...



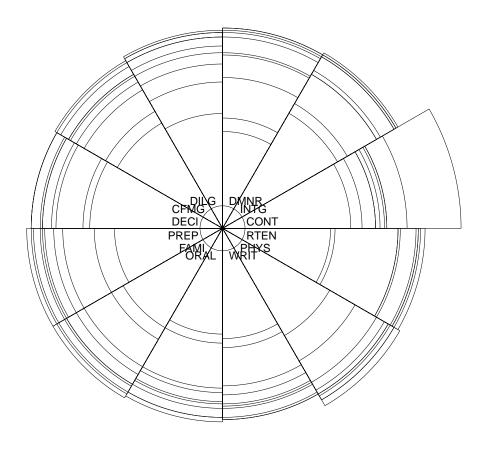
```
## 'Segments':
stars(USJudge, draw.segments = TRUE, scale = FALSE, key.loc = c(13,1.5))
```



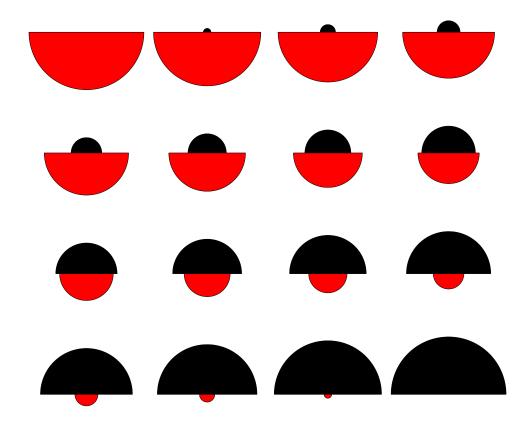
US Judges rated



US Judges 1-10







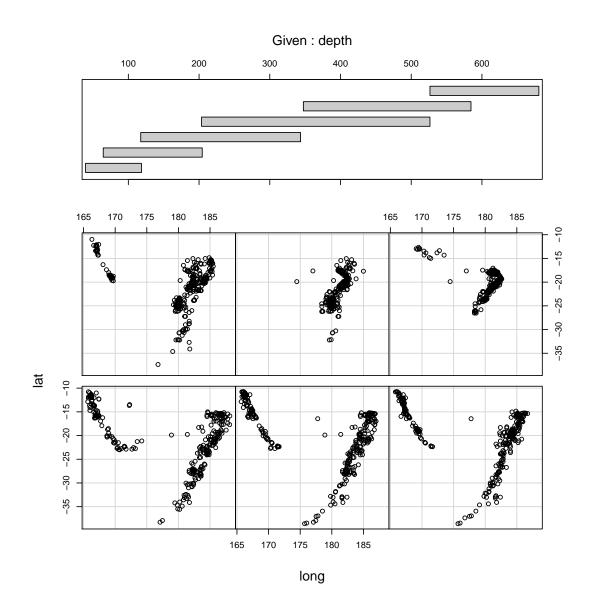
16 Steam and leaf

```
stem(islands)
##
    The decimal point is 3 digit(s) to the right of the |
##
##
     0 | 0000000000000000000000000000111111222338
##
     2 | 07
##
     4 | 5
##
      6 | 8
##
##
     8 | 4
## 10 | 5
```

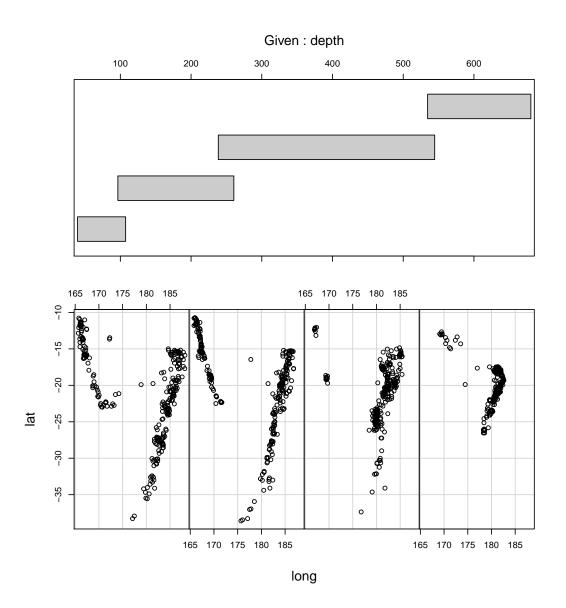
```
12 |
##
##
   14 |
## 16 | 0
stem(log10(islands))
##
##
    The decimal point is at the |
##
##
    1 | 1111112222233444
##
    1 | 5555556666667899999
## 2 | 3344
## 2 | 59
## 3 |
## 3 | 5678
## 4 | 012
```

17 Conditioning plots

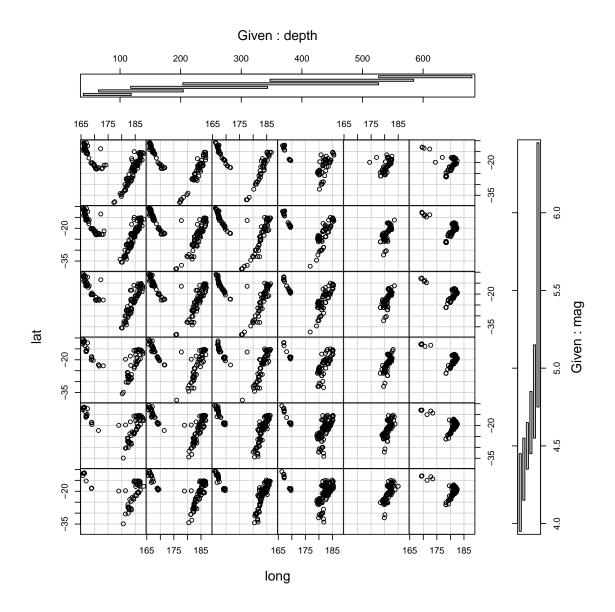
```
## Tonga Trench Earthquakes
coplot(lat ~ long | depth, data = quakes)
```



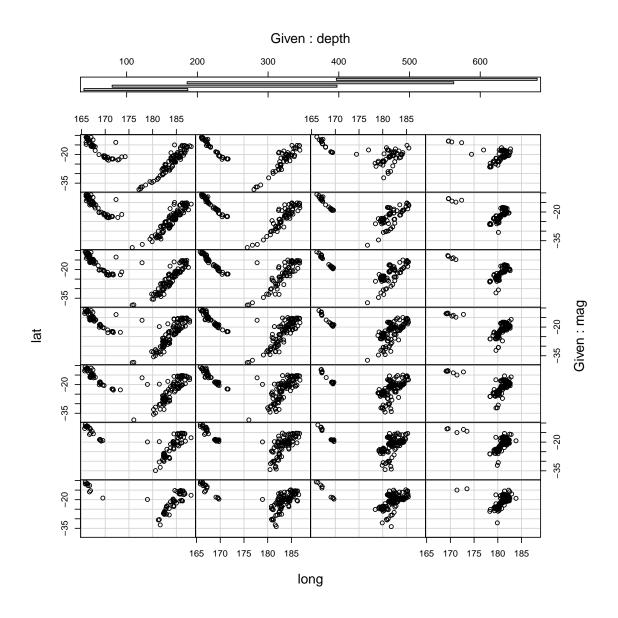
given.depth <- co.intervals(quakes\$depth, number=4, overlap=.1)
coplot(lat ~ long | depth, data = quakes, given.v=given.depth, rows=1)</pre>

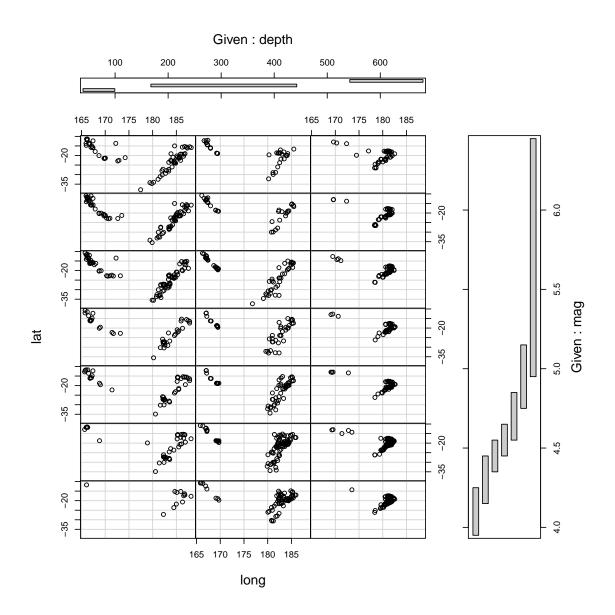


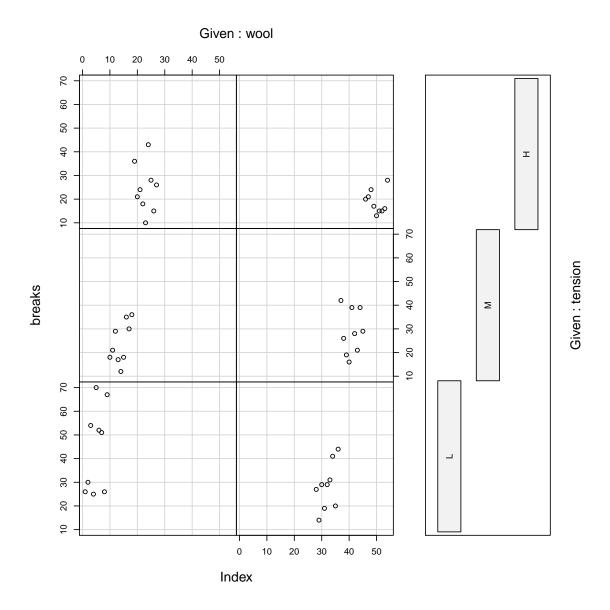
```
## Conditioning on 2 variables:
11.dm <- lat ~ long | depth * mag
coplot(11.dm, data = quakes)</pre>
```



coplot(ll.dm, data = quakes, number=c(4,7), show.given=c(TRUE,FALSE))

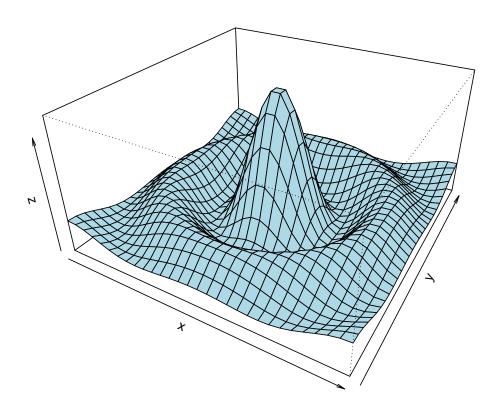




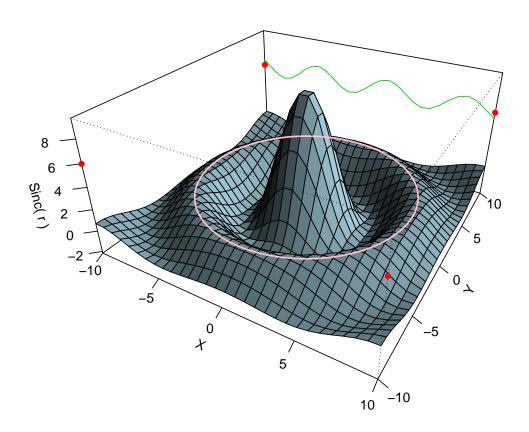


18 persp

```
persp(x, y, z, theta = 30, phi = 30, expand = 0.5, col = "lightblue")
```

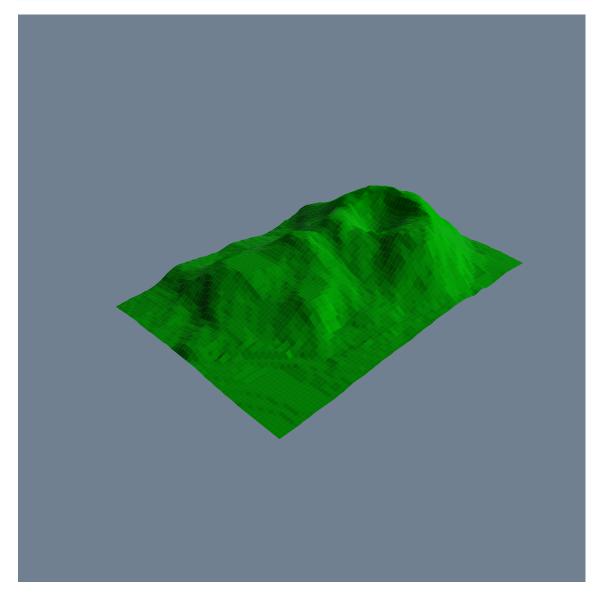


```
points(trans3d(xy[,1], xy[,2], 6, pmat = res), col = 2, pch =16)
lines (trans3d(x, y=10, z= 6 + sin(x), pmat = res), col = 3)
phi <- seq(0, 2*pi, len = 201)
r1 <- 7.725 # radius of 2nd maximum
xr <- r1 * cos(phi)
yr <- r1 * sin(phi)
lines(trans3d(xr,yr, f(xr,yr), res), col = "pink", lwd = 2)</pre>
```



```
## (no hidden lines)
```

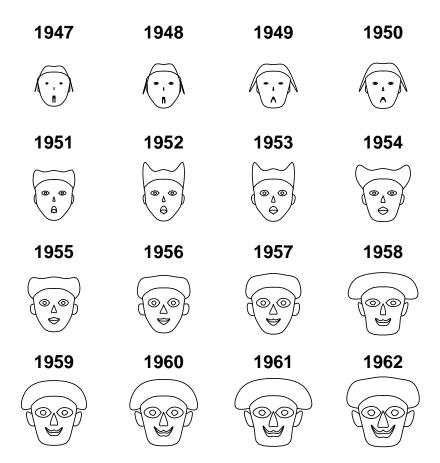
```
# (3) Visualizing a simple DEM model
z <- 2 * volcano  # Exaggerate the relief</pre>
```



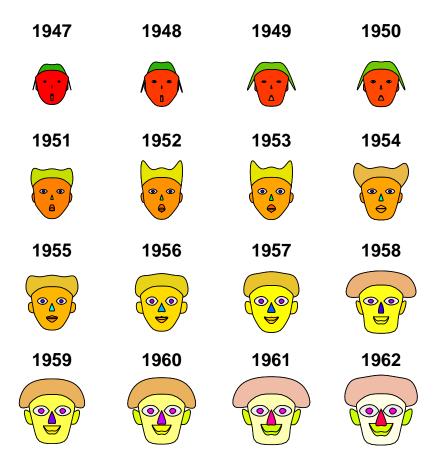
```
par(op)
```

19 Chernof's faces

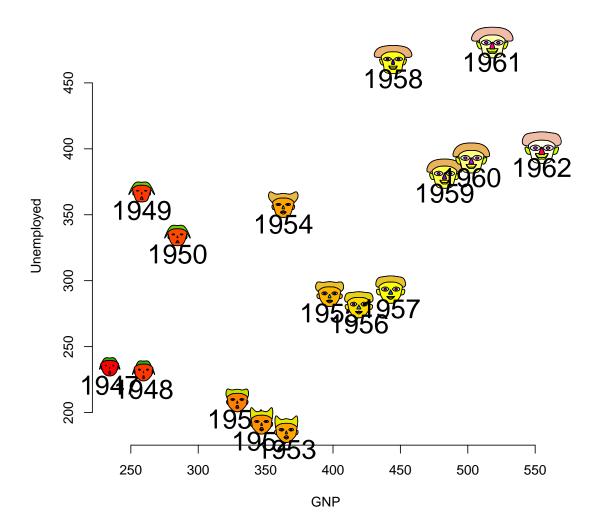
```
library(aplpack)
data(longley)
faces(longley[1:16,],face.type=0)
```



faces(longley[1:16,],face.type=1)

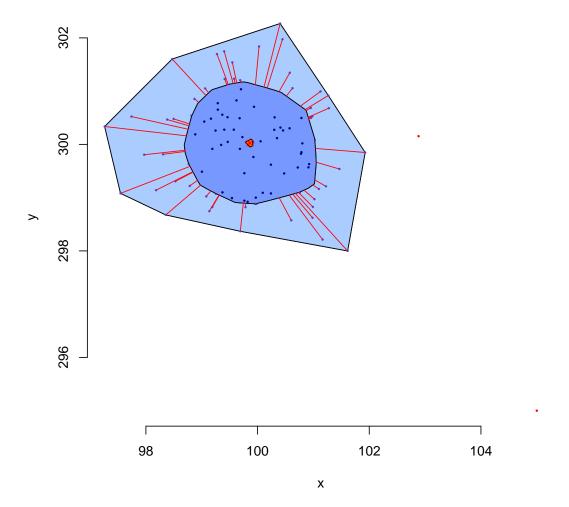


```
plot(longley[1:16,2:3],bty="n")
a<-faces(longley[1:16,],plot=FALSE)
plot.faces(a,longley[1:16,2],longley[1:16,3],width=35,height=30)</pre>
```



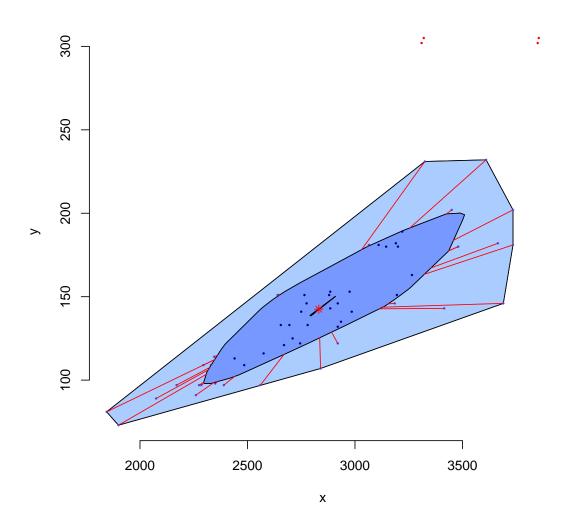
20 Bagplot

20.1 Individual bagplots

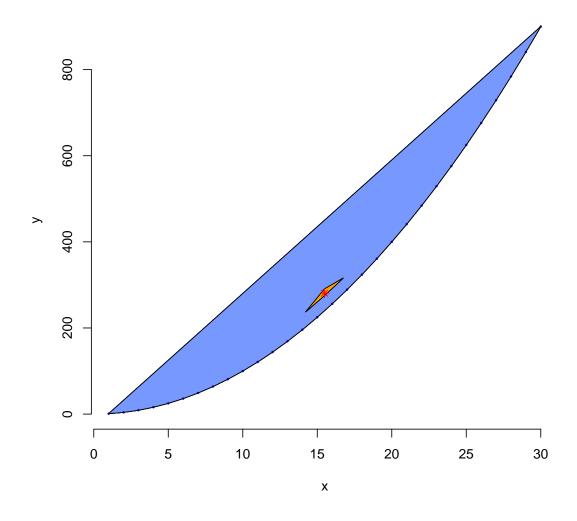


```
bagplot(cardata,factor=3,show.baghull=TRUE,
    show.loophull=TRUE,precision=1,dkmethod=2)
title("car data Chambers/Hastie 1992")
```

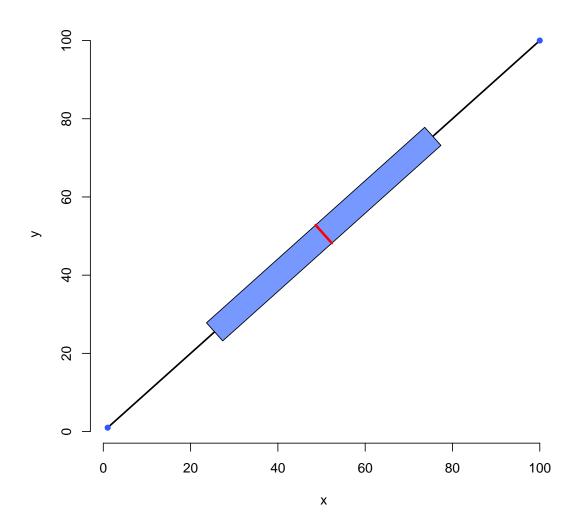
car data Chambers/Hastie 1992



```
# points of y=x*x
bagplot(x=1:30,y=(1:30)^2,verbose=FALSE,dkmethod=2)
```

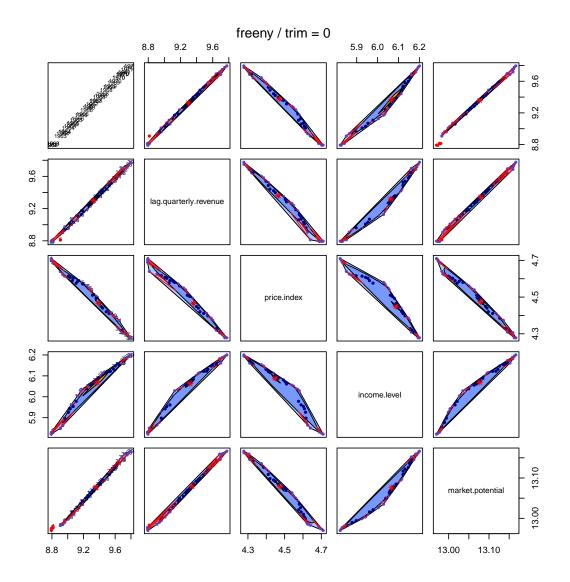


one dimensional subspace
bagplot(x=1:100,y=1:100)

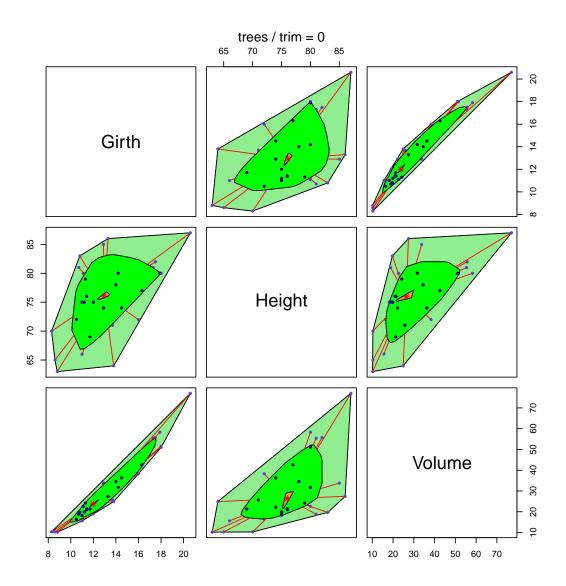


20.2 Pairs of bagplots

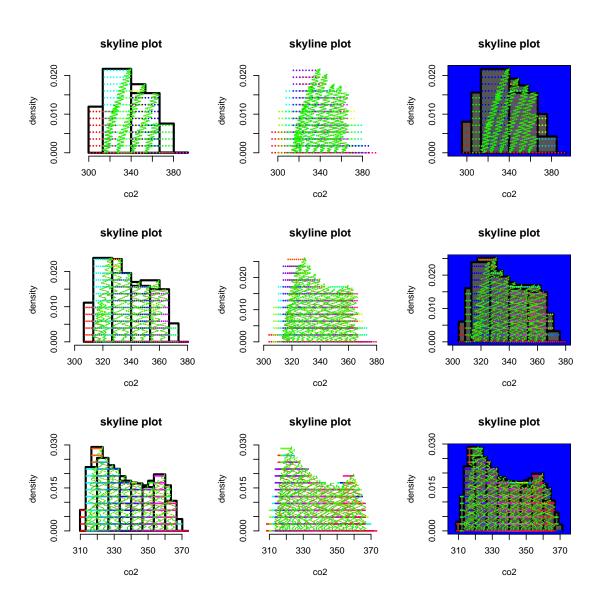
bagplot.pairs(freeny)



bagplot.pairs(trees,col.baghull="green", col.loophull="lightgreen")



21 Skyline plots



```
night="orange" , ylab="density", col.bars = "white", col.border=1 )
skyline.hist(x=rivers, n.class=6, n.hist=1, n.shading=0, main="rivers",
           cex.data=0.1, lwd.data = 2, col.data = "red", pcol.data = "green",
           col.border=NA, night=FALSE , ylab="density", col.bars = "lightblue")
skyline.hist(x=rivers, n.class=6, n.hist=1, n.shading=5, col.shading = "blue",
           main="rivers",
           cex.data=0.1, lwd.data = 1, col.data = "black", pcol.data = "green",
           col.border=NA, night=FALSE , ylab="density", col.bars = "green")
skyline.hist(x=rivers, n.class=6, n.hist=3, n.shading=5, col.shading = "blue",
           main="rivers", col.bars = "green",
           cex.data=0.1, lwd.data = 1, col.data = "black", pcol.data = "green",
           col.border="white", night="magenta" , ylab="density")
skyline.hist(x=rivers, n.class=6, n.hist=4, n.shading=5, col.shading = "blue",
           main="rivers",
           cex.data=0.8, lwd.data = 1, col.data = "blue", pcol.data = "red",
           col.border=NA, night=FALSE , ylab="density", col.bars = "green")
```

