## Patience: Base package graphics

advanced plots in the base package

## **Graphical Functions**

Two types of plotting function in the base package:

```
√ high-level, e.g., plot(), boxplot(), image()
√ low-level, e.g., title(), points(), abline()
```

- ➤ In general, **modularity is good**, meaning use low level function to decorate plots.
- Let's start with something familiar the "mtcars" dataset.

#### **An Old Favourite**

```
Do it yourself: Get it into the memory:
> attach (mtcars)
Refresh yours!
  mtcars 4
  ?mtcars
  View (mtcars) ↔
  str(mtcars) ↔
> plot(mtcars) ↔
```

Notice how insane the plot() function can be = all against all. Let's make things behave properly.

## **Bar Charts and Aggregating Counts**

We can use barplot for plotting counts (in which both the x and y variable are categorical).

```
Do it yourself: Construct a barplot for number of gears in mtcars data set:
```

- > gear ↔
- > counts <- table(gear) 4

Yes, `table()` tabulates count data = cool...

- > counts4
- > barplot(counts) 4

That looks alright, but a bit simple...

## **Elaborating Plots**

#### **Do it yourself:** plot the density of the counts:

```
> barplot(counts/length(gear), names.arg=c("three","four","five"))
```

#### and now label our plot:

```
> title(main="Car Distribution",
xlab="Number of Gears", ylab="Relative
Frequency")
```

We wrote decorative text using a low level function.

#### **Bar Charts with Point Estimates**

We can use barplot for point estimates of a continuous y variable; we need an \*apply function.

**Do it yourself:** calculate the mean mpg rating for cars with different numbers of gears:

- > heights <- tapply(mpg, gear, mean) <- and now plot and label these:
- > barplot(heights)←
- > title(main="Mean Efficiency",
  xlab="Number of Gears", ylab="Miles
  per Gallon")

#### **Confidence Intervals**

We can also plot error bars. For this we need an extra library\*.

Do it yourself: install and then load the package gplots:

- > install.packages('gplots')↔
- > library(gplots) 4

This package contains an enhanced version of the barplot() function called barplot2().

<sup>\*</sup> it is also possible to plot error bars with the base package + the low-level function arrows().

#### **Confidence Intervals**

```
Do it yourself: here we exploit summary stats produced
by the t.test() function*:
> lower <- tapply(mpg, gear, function(i)</pre>
t.test(i)$conf.int[1]) ←
> upper <- tapply(mpg, gear, function(i)</pre>
t.test(i)$conf.int[2]) ←
Plot the graph and decorate:
> barplot2 (heights, plot.ci=TRUE, ci.l=lower,
ci.u=upper, ci.width=0.2)
> title(main="Car Efficiency", xlab="Number of
Gears", ylab="Miles per Gallon") ←
```

## **Multiple Predictors**

**Do it yourself: Construct a** barplot **for number of** gears **in** mtcars **data set grouped by** vs:

```
> counts <- table(vs,gear) <-
> barplot(counts,
col=c("lightblue","mistyrose"), legend=T,
args.legend=list(title="vs")) <-
> title(main="Car Distribution by Gears
and VS", xlab="Number of Gears") <-
```

Try adding beside=TRUE as an argument to the barplot() call. What happens?

## **Distribution Plots: Boxplot**

You can use boxplot() to get an idea of the distribution of y ~ (as affected by) x with an intuitive model formula.

Something like tapply() is happening behind the scenes.

## Warning!!!



**Do it yourself** This is what happens when you pass variables without the model notation:

```
> boxplot(am, mpg, names=c("all am value", "all mpg values")) ←
```

So there's nothing wrong with passing vectors to boxplot(), but don't expect tapply() behaviour unless you use ~ notation.

#### **Distribution Plots: Violin Plots**

A nice alternative is vioplot():

#### Do it yourself

Start by installing and loading the required library:

- > install.packages('vioplot') 4
- > library(vioplot) ←

## **Subsetting Data**

 The `subset()` function helps to split a variable/vector using a factor:

```
Do it yourself: Get subsets of data:
> auto <- subset(mpg, am==0,
data=mtcars) ←
> man <- subset(mpg, am==1,
data=mtcars) ←</pre>
```

See my notes for the tidyverse approach to this (which I prefer).

#### **Violin Plots**

Now we can try a violin plot:

```
Do it yourself: Now we can plot:

> vioplot(auto, man,
names=c("Automatic", "Manual")) ←

> title(main="Violin Plot",
xlab="Transmission", ylab="Miles per
US Gallon") ←
```

 But be careful – this does not support the formula notation hence the need for vector splitting...

## **Distribution Plots: Histograms**

```
Do it yourself: Let's compare distributions:
> hist(auto, col=rgb(1,0,0,0.5),
breaks=seq(10,36,2), xlim=c(10,35),
vlim=c(0,5), main="", xlab="")
> hist(man, col=rgb(0,0,1,0.5),
breaks=seq(10,36,2), add=T) \leftarrow
Decorate:
> title(main="Double Histogram", xlab="Miles
per Gallon") ←
> legend("topright", c("automatic", "manual"),
fill=c(rgb(1,0,0,0.5),rgb(0,0,1,0.5))
```

## **Advanced Boxplots (base)**

The formula notation in boxplot supports interrogation of multiple predictors.

**Do it yourself:** We can plot by two categorical predictors using boxplot too.

```
> boxplot(mpg~vs*am, data=mtcars,
col=(c("mistyrose","lightblue")))
> title(main="Car Engines", xlab="Config *
Transmission", ylab="Miles per gallon")
```

Try adding a notch=TRUE argument to the plot call.

I'll leave further decoration to you...

## **Cleaning Up**

```
Do it yourself Detach from the dataset
```

```
> detach(mtcars) ←
```

You can also clean up variables:

```
> rm(auto, man) ↔
```

> rm(list=ls())

Using rm() is playing with fire!

## **Controlling Graphical Parameters**

#### A microarray with minimal data...

#### Do it yourself: Here we load some data from a file:

```
> setwd('~/Desktop/render-master') 
> array <- read.csv('heatmaps_in_r.csv',
header = TRUE, comment.char="#", row.names =
1)
```

# We reset some graphical parameters and make our heatmap:

```
> op <- par(cex.axis=0.6, las=2) 
> image(x=1:10, y=1:4, z=data.matrix(array),
col=rev(heat.colors(10)), xlab='', ylab='') 
> hist(rnorm(100))
```

## **Controlling Graphics**

#### Do it yourself: Reset graphics back to original:

- > par(op) ↔
- > hist(rnorm(100))  $\leftarrow$

Now you can nerd out with graphical parameters:

- > ?par **√**
- > ?points←
- > ?colours 4
- > ?image

Take home: parameters can be set and reset within each device. Meaning...



## Writing to a Device

**Do it yourself:** Now let's get serious with graphics parameters and plotting devices:

```
> pdf("nanoarray.pdf", 12, 6) 4
> par(cex.axis=0.6, las=2) 4
> image(x=1:10, y=1:4, z=data.matrix(array),
col=rev(heat.colors(10)), xlab='', ylab='',
axes=F) 4
> axis(1, at=1:10, labels=rownames(array)) 4
> axis(2, at=1:4, labels=colnames(array)) 4
> title(xlab="sample", ylab="probe",
cex.main=2) 4
> dev.off() 4
```

## Using source() to run all lines in script

 There is a script called makearray.R in the render-master folder:

#### Do it yourself: run this script:

- > source('makearray.R') ↔
  - What has this script done?
  - Hints:
    - Look in the folder again for new files
    - Open the script using RStudio or a text editor
    - Can you see any new objects in memory?

#### In RStudio's Environment

