Visualising big data in R

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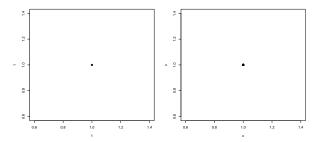
The challenge of visualising big data

- Only a few million pixels on a screen, but many more data points
- Therefore need to generate a suitable summary to plot instead
- Directly visualising raw big data is probably pointless, at least for static graphics (real-time manipulation of big data, e.g. fly throughs is another matter...)
- \rightarrow A typical 1D/2D plot of big data will have lots of overlapping & therefore **obscured** points: these different values will be visually indistinguishable

Hidden plot points

• Single vs. multiple overlapping points

```
pdf(file = "N1.pdf", compress = FALSE); plot(1, 1, pch = 19)
bmp(file = "N1.bmp", antialias = "none"); plot(1, 1, pch = 19)
x <- rep(1, 1e3) # 1000 identical points
pdf(file = "N1000.pdf", compress = FALSE); plot(x, x, pch = 19)
bmp(file = "N1000.bmp", antialias = "none"); plot(x, x, pch = 19)
graphics.off()</pre>
```



The 2 plots look identical (apart from subtle anti-aliasing effects)

Image file size comparison

• List the size of each image file (in bytes):

```
sapply(list.files(pattern="N1.*"), function(f) file.info(f)$size)
N1000.bmp N1000.pdf N1.bmp N1.pdf
    231478 75390 231478 12452
```

- ullet The bitmap graphics raster images are identical in size, but the (uncompressed) vector graphics PDFs differ in size by a factor \sim 6
- ullet Raster graphics resolve overlaps: 1000 overlapping points \equiv 1 point, but vector graphics retain each point as a separate entity
- Similar principle required for Big Data, but need more control...

The bigvis concept

Download the paper (vita.had.co.nz/papers/bigvis.html), by Hadley Wickham: Bin-summarise-smooth: a framework for visualising large data

- Bigvis is a scheme for pre-processing big datasets
 - ullet output can then be handled by conventional (R) plot tools
 - processing done in (fast) compiled C++ code, using Rcpp package
- bigvis goal: be able to plot 100 million points in under 5 seconds
- Also provides outlier removal and smoothing:
 - big data means very rare cases can occur ⇒ outliers may be more of a problem
 - smoothing very important to highlight trends & suppress noise

Installing bigvis

Project website: github.com/hadley/bigvis

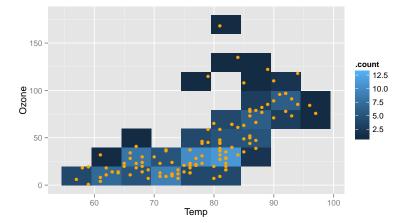
```
install.packages("devtools")
library(devtools)
install_github("bigvis")
```

Recent blog article about bigvis:

blog. revolution analytics. com/2013/04/visualize-large-data-sets-with-the-bigvis-package. html.

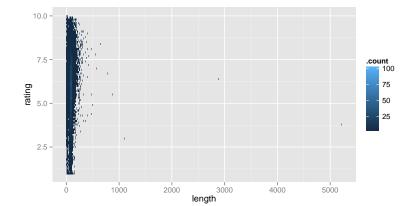
bigvis applied to a small dataset

```
library(bigvis); library(ggplot2)  # load packages
binData <- with(airquality, condense(bin(Ozone, 20), bin(Temp, 5)))
p <- ggplot(data=binData, aes(Temp, Ozone, fill=.count)) + geom_tile()
p + geom_point(data=airquality, aes(fill=NULL), colour="orange")</pre>
```



Movie length vs. IMDB rating: big-ish data, with outliers!

• ~ 130,000 row data frame (bigvis::movies, from imdb.com)



...in case you were wondering

• Which films are longer than...1000 minutes?! (\sim 17 hours!)

```
longest <- subset(movies, length > 1e3)
longest[c("title", "length", "rating")]
```

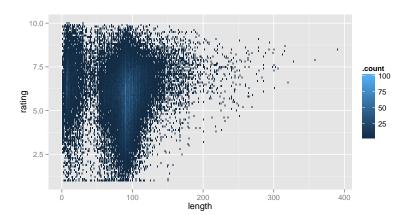
title	length	rating
Cure for Insomnia, The	5220	3.8
Four Stars	1100	3
Longest Most Meaningless Movie in the World, The	2880	6.4

...aptly named!

Bigvis plot with outliers removed

- Outliers a problem with Big Data: extreme events do occur
- Update previous plot to use bigvis peel function to strip off outermost (1%, by default) extreme values:

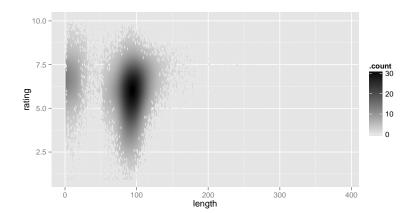
last_plot() %+% peel(binData) # same plot, different dataset



Smoothing in bigvis

 Also use autoplot function from bigvis; peel off outliers first, then smooth with different bandwidths for length & rating

```
smoothBinData <- smooth(peel(binData), h=c(20, 1))
autoplot(smoothBinData)</pre>
```



Live Demo

```
N <- 1e7
raw <- data.frame(x = rt(N, 2), y = rt(N, 2)) # 10 million rows
## ~3s to run:
system.time( binned <- with(raw, condense(bin(x), bin(y)) ) )
## Plot condensed (i.e. pre-processed) data:
ggplot(data=binned, aes(x,y, fill=.count)) + geom_tile()
## Peel outliers & replot:
system.time( peeled <- peel(binned) )
ggplot(data=peeled, aes(x,y, fill=.count)) + geom_tile()</pre>
```

Summary,

- Pre-processing to generate statistical summaries is the key to plotting Big Data
- The R bigvis package is a very powerful tool for plotting large datasets and is still under active development
 - ullet includes features to strip outliers, smooth & summarise data
- v3.0.0 of R (released Apr 2013) represents a solid platform for extending the outstanding data analysis & visualisation capabilities of R to meet the challenge of Big Data, with excellent prospects for future releases

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Session information

These slides were created with (Emacs) org mode.

R version R version 3.0.0 (2013-04-03), i686-pc-linux-gnu

Birmingham R User Meeting (BRUM)



www.birminghamR.org

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