

# 3. More Advanced Graphics

# Thomas Lumley Ken Rice

Universities of Washington and Auckland

Seattle, July 2014

### **Outline**

- Colour and pre-attentive perception: facts about graphics
- Too many variables: parallel coordinates, transparency
- Too many dimensions: hexagonal binning, transparency

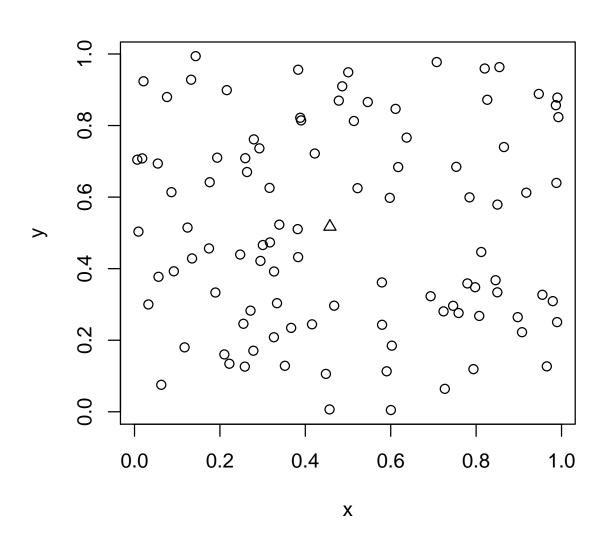
### Colour coding

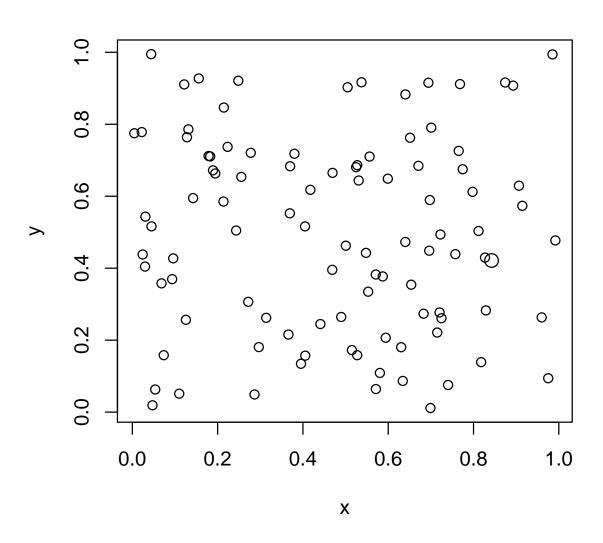
'Simple' plots involve two-dimensional data, which we measure on the x and y axes.

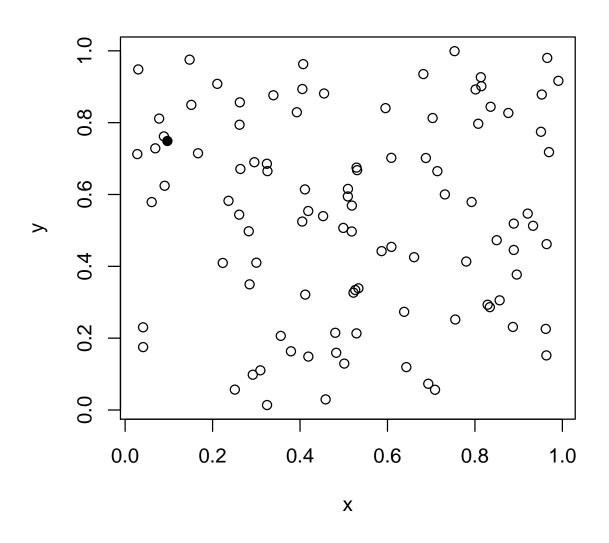
For higher-dimensions, some traditional approaches are;

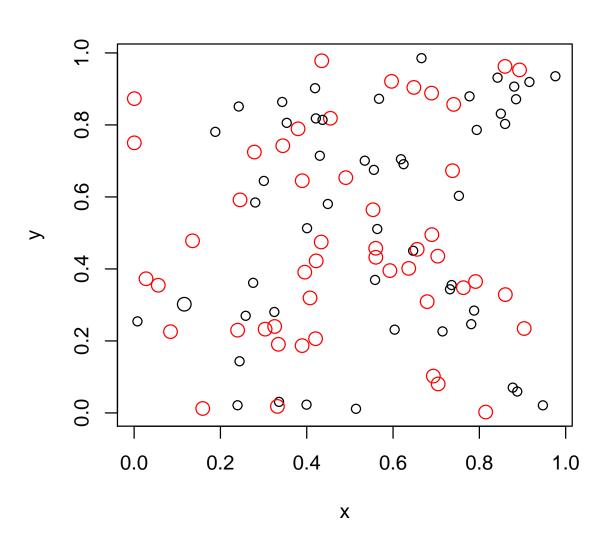
- Different colors for e.g. men, women (col)
- Different-shaped symbols (pch), or different sizes (cex)

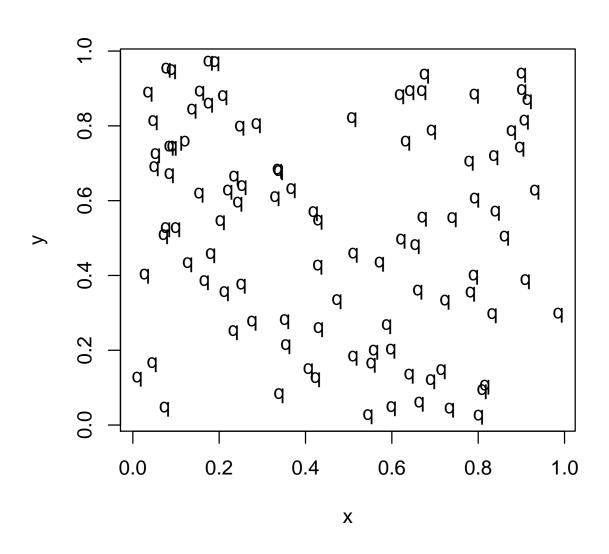
For  $\leq$  100's of data points, modest use of these is fine. But your eye is not good at concentrating e.g. just on the purple points, in a fully Technicolor plot;











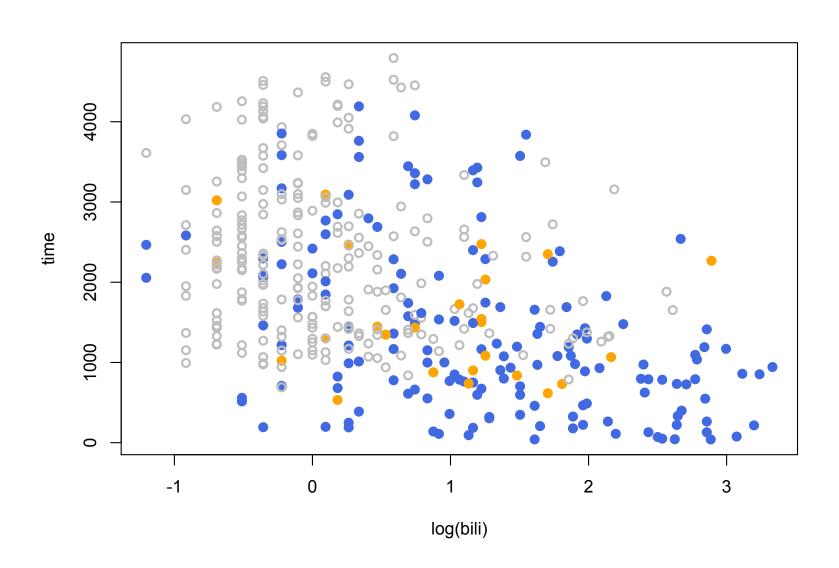
### **Preattentive perception**

Some differences are processed by the brain before you get to see the image: pre-attentive perception.

Important because

- It's easier to see things
- You can look at just one subset of the points and see patterns
- Like colour-blindness, illustrates that there are **facts** about graphics, not just artistic taste

# **Preattentive perception**



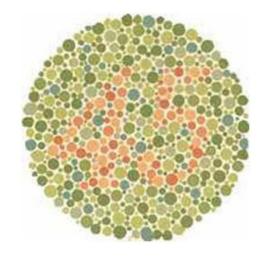
### Color schemes

Color choice is best left to experts, or people with taste.

http://www.colorbrewer.org has color schemes designed for the National Cancer Atlas, also in package RColorBrewer

colorspace package has color schemes based on straight lines in a perceptually-based color space (rather than RGB).

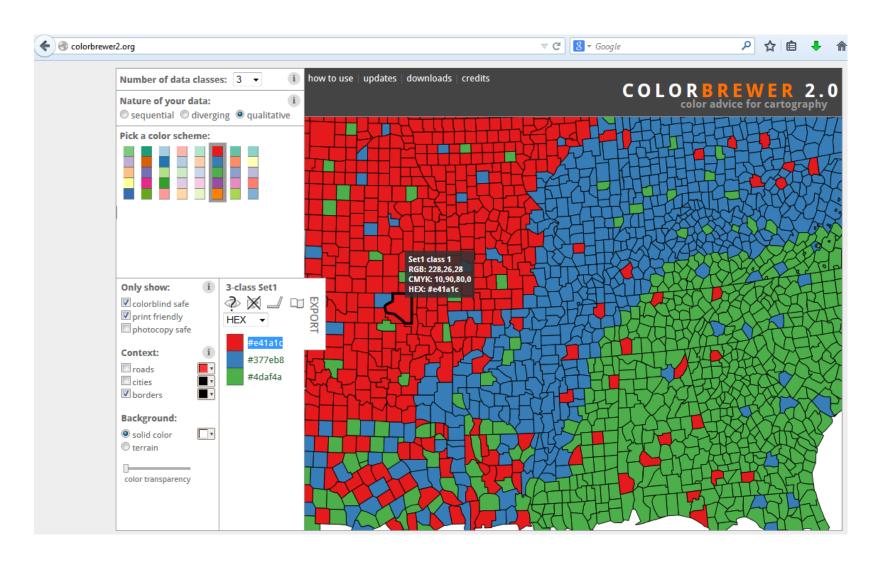
dichromat package attempts to show the impact of red:green color blindness on your R color schemes.



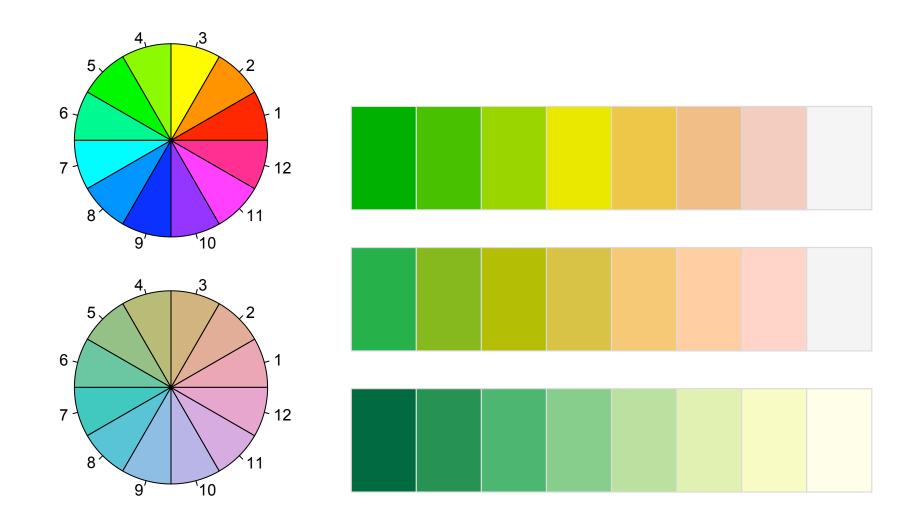
[Code for examples is in file colorpalettes.R on course website]

### Color brewer

R accepts 'hex' colors, e.g. col="#e41a1c" here;

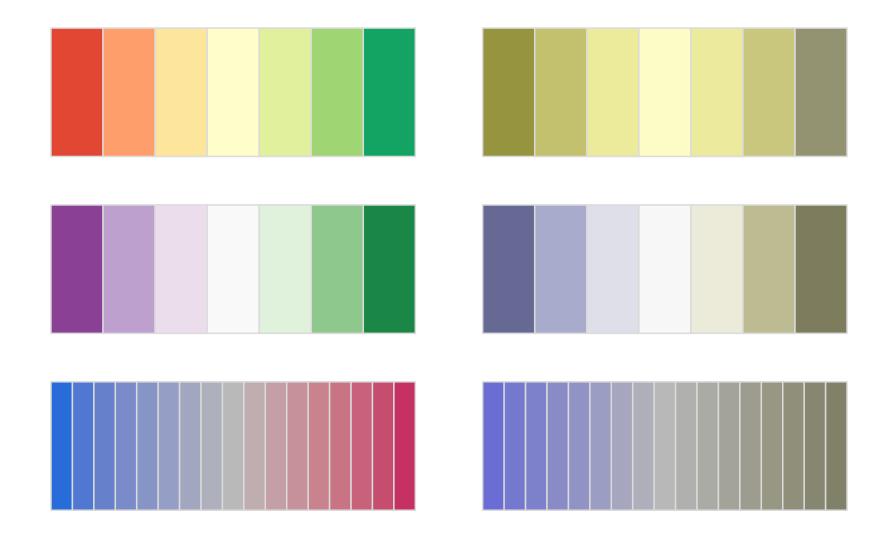


### Color choice



(nb B&W printed copies of this slide may not be helpful!)

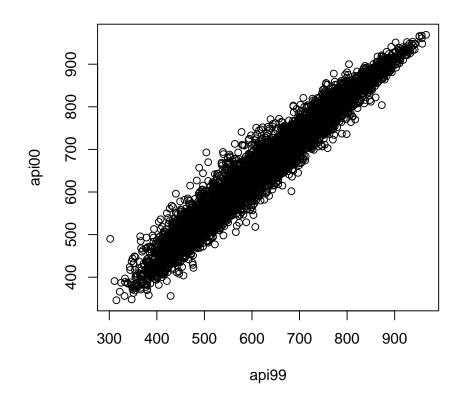
### **Color blindness**



(nb B&W printed copies of this slide may not be helpful!)

### Larger data

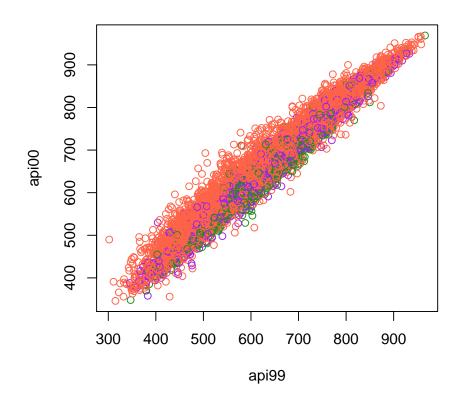
For large(ish) data, 'overlap' is a fundamental problem...



(California Academic Performance Index on 6194 schools)

### Larger data

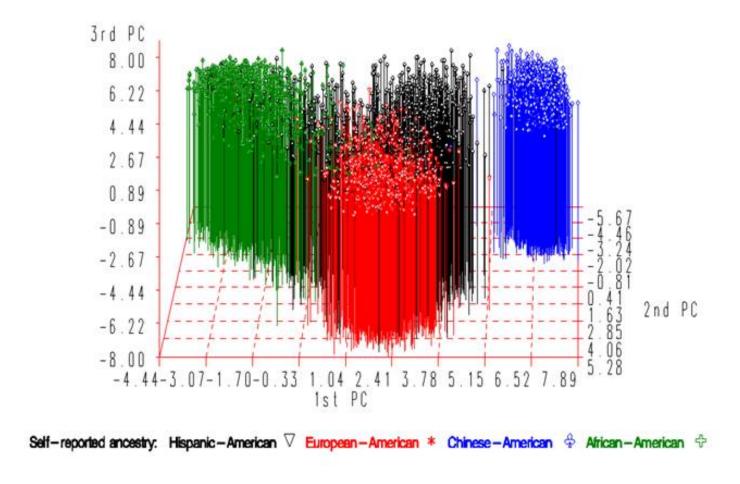
... which remains, when we color-code.



Colors denote Elementary, Middle & High Schools

### Larger data

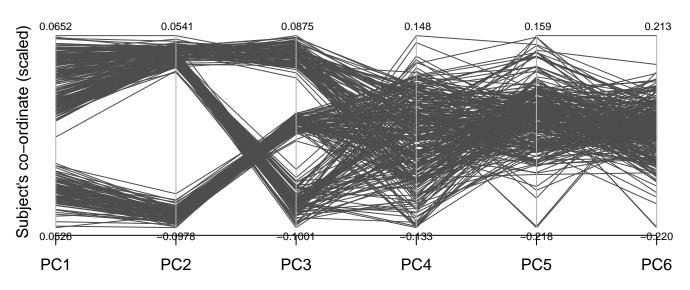
With three dimensions + color-codes, this can happen;



(R does have persp(), for occasional use)

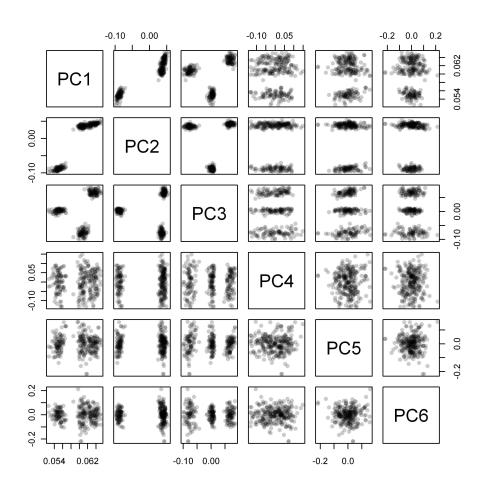
For even higher-dimensional data, scatterplots can not provide adequate summaries. For data where the dimensions can be ordered, the parallel co-ordinates plot is useful;

#### **Leading Principal Components, n=279, 10000 SNPs**



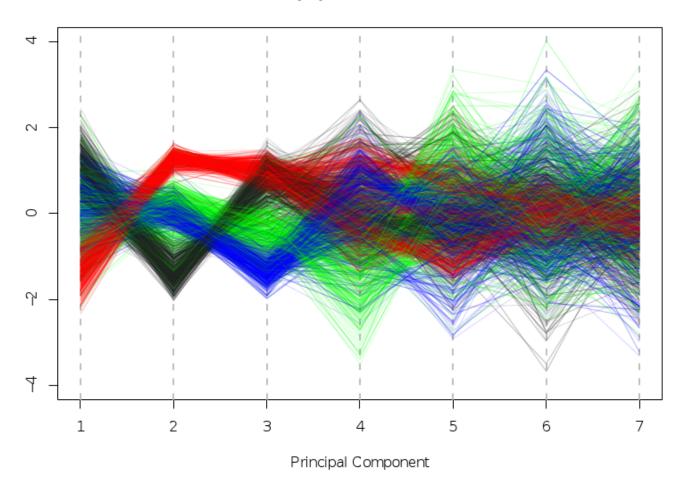
- Each multi-dimensional data point (i.e. each person) is represented by a line – not a point
- parcoord() in the MASS package is one simple implementation
   writing your own version is not a big job
- Coloring the lines also helps (example later)
- Scaling of axes, and their vertical positions are arbitrary
- Doing 'Principal Components Analysis' is just choosing axes for your data so that their variance is maximized on axis 1, then axis 2, ...

A pairs() plot of the same thing; (nasty!)



The pin cushion data++: colors indicate self-report ancestry

#### Whole MESA population - normalized PCs



### **Transparency**

The colors in the last examples were transparent. As well as specifying e.g. col=2 or col="red", you can also specify

col="#FF000033"

– coded as RRGGBB in hexadecimal, with transparency 33 (also hexadecimal). This is a 'pale' red – 33/FF  $\approx$  20%.

Get from color names to RGB with col2rgb(), and from base 10 to base 16 using format(as.hexmode(11), width=2)

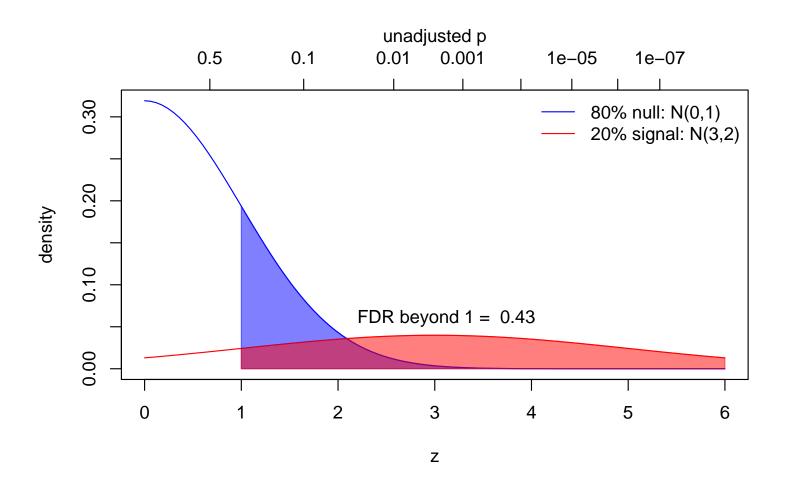
(Or, go to colorbrewer or a similar site and take the hex from there!)

### **Transparency**

An example; (also shows other graphics commands) curve(0.8\*dnorm(x), 0, 6, col="blue", ylab="density", xlab="z") curve(0.2\*dnorm(x,3,2), 0, 6, col="red", add=T) $xvals \leftarrow seq(1, 6, l=101)$ polygon( c(xvals, 6, 1), c(0.8\*dnorm(xvals), 0, 0),density=NA, col="#0000FF80") # tranparent blue polygon( c(xvals, 6, 1), c(0.2\*dnorm(xvals, 3, 2), 0, 0),density=NA, col="#FF000080") # tranparent red legend("topright", bty="n", lty=1, col=c("blue", "red"), c("80% null: N(0,1)", "20% signal: N(3,2)"))axis(3, at=qnorm(c(0.25, 0.5\*10^(-1:-7)), lower=F), c(0.5,  $10^{-1:-7}$ )) mtext(side=3, line=2, "unadjusted p") text(2.2, 0.07, adj=c(0,1), paste("FDR beyond 1 = ",round(0.8\*pnorm(1,lower=F)/(0.8\*pnorm(1,lower=F) + 0.2\*pnorm(1,3,2,lower=F)),3)))

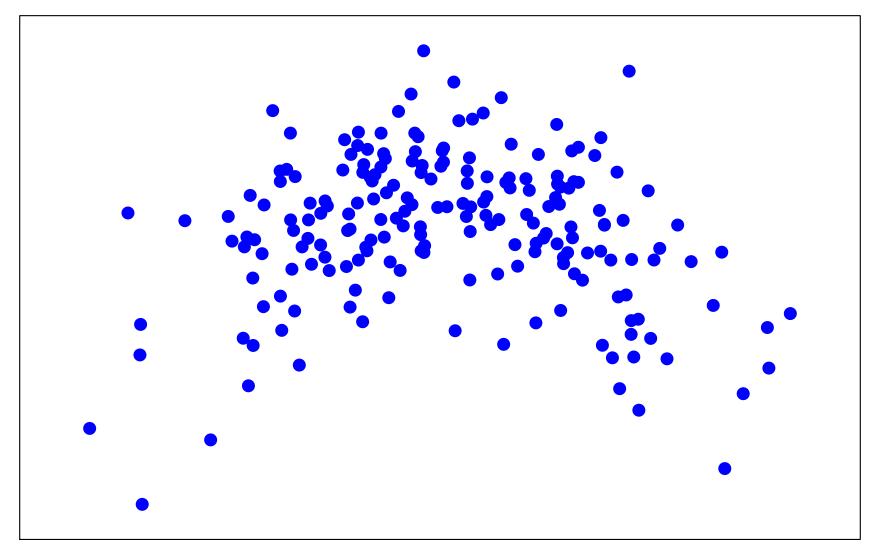
### **Transparency**

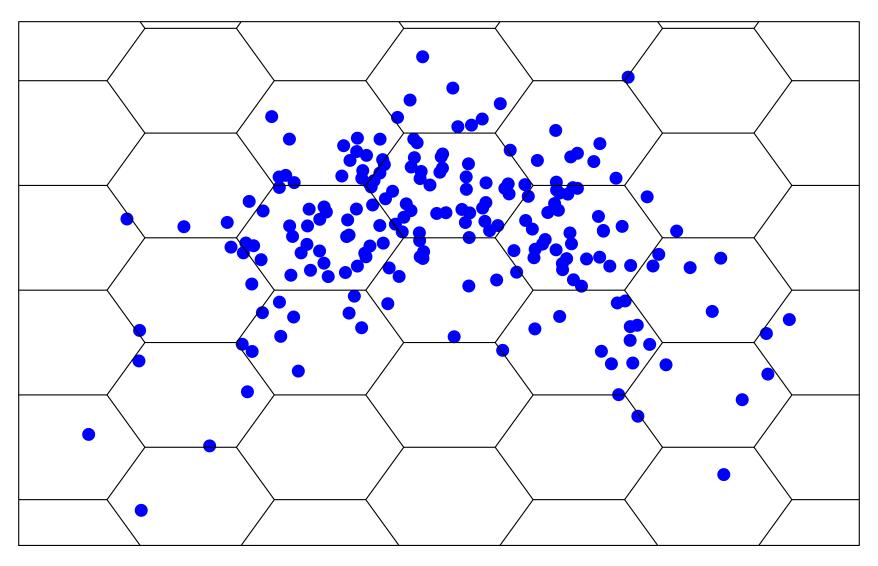
Here's the output;

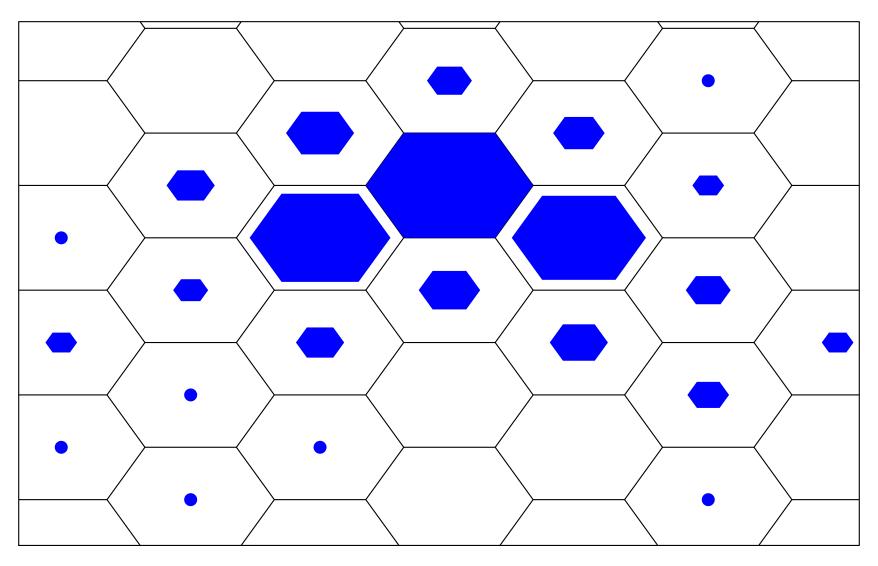


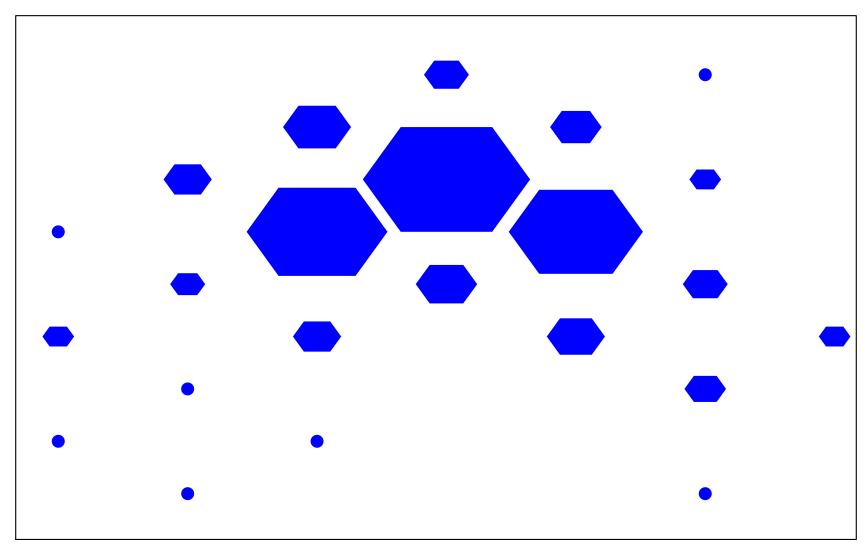
Using transparent plotting symbols is a quick-and-dirty way to adapt scatterplots for use with large datasets.

A better method is 'hexagonal binning'; this is a 2D analog of a histogram — where you would count the number of data in one area, and then draw a bar with height proportional to that count.



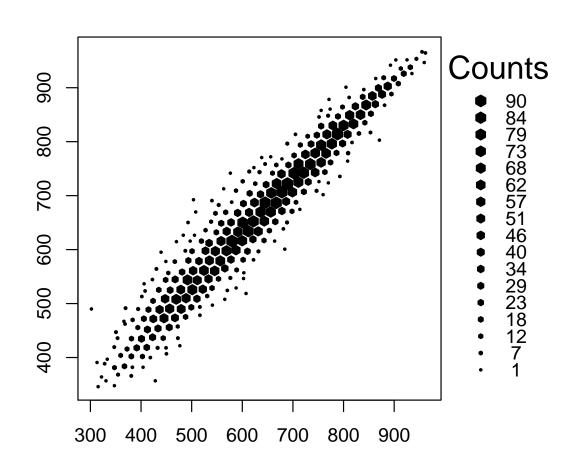






The hexbin() package does all the bin construction, and counting. It has a plot method for its hexbin objects;

```
install.packages(c("hexbin","survey"))
library("hexbin")
library("survey")# for apipop data frame
with(apipop, plot(hexbin(api99,api00), style="centroids"))
```



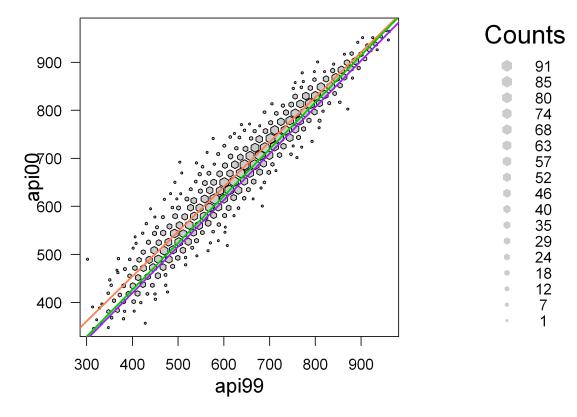
Hexbin is used when you don't *really* care about the exact location of every single point

- Singleton points are plotted 'as usual'; you do (perhaps) care about them
- hexbin centers the 'ink' at the cell data's 'center of gravity'
- style="centroids" gives the center-of-gravity version; the default style is colorscale — usually grayscale. See ?gplot.hexagons for more options

For keen people: the hexbin package doesn't use the standard R graphics plotting devices; instead, it operates through the Grid system (in the grid package) which defines rectangular regions on a graphics device; these viewport regions can have a number of coordinate systems. To add lines to a hexbin plot, the options are;

- Use hexVP.abline() to add these directly
- Move everything into 'standard' graphics not Grid graphics (see ?Grid). The Grid system lets you alter graphics after plotting them
- Write your own plot method for hexbin objects, with standard R graphics commands
- Make do with hexBinning() in the fMultivar package

An example; color-coded lines of best fit, by school type;



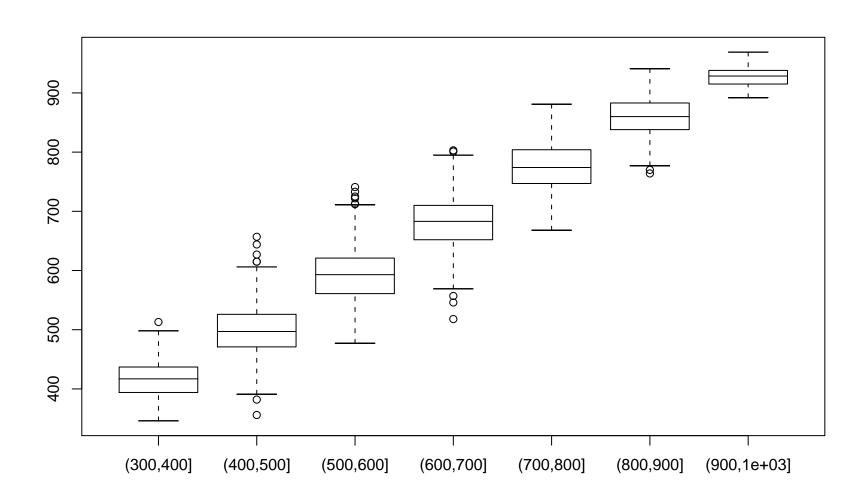
```
lm.e <- coef(lm(api00~api99, data=apipop, subset=stype=="E"))
lm.m <- coef(lm(api00~api99, data=apipop, subset=stype=="M"))
lm.h <- coef(lm(api00~api99, data=apipop, subset=stype=="H"))
hexVP.abline(vp1$plot.vp, lm.e[1], lm.e[2], col="coral")</pre>
```

### Large data: multiple groups

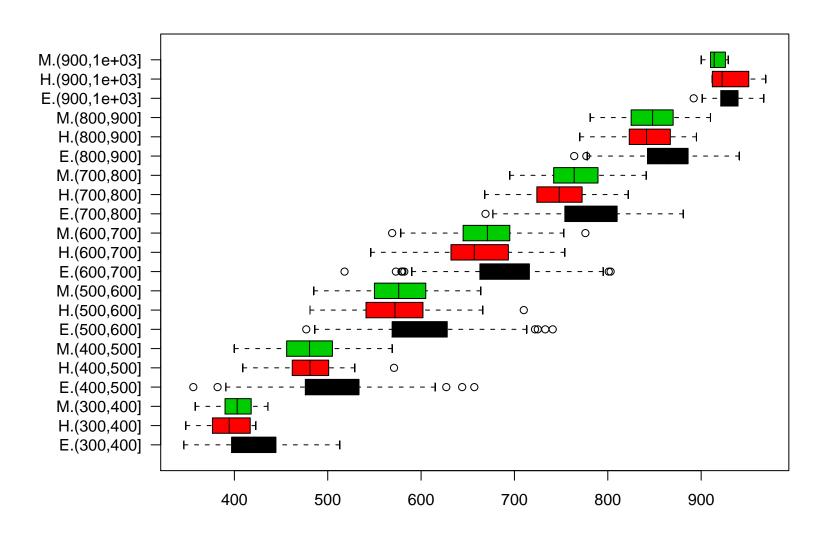
For showing multiple groups, a scatterplot smoother or perhaps boxplots or conditioning may be better.

- cut() turns a variable into a factor by cutting it at the specified points.
- par(mar=) sets the margins around the plot. We need a large left margin for the labels.

# Large data: multiple groups



### Large data: multiple groups

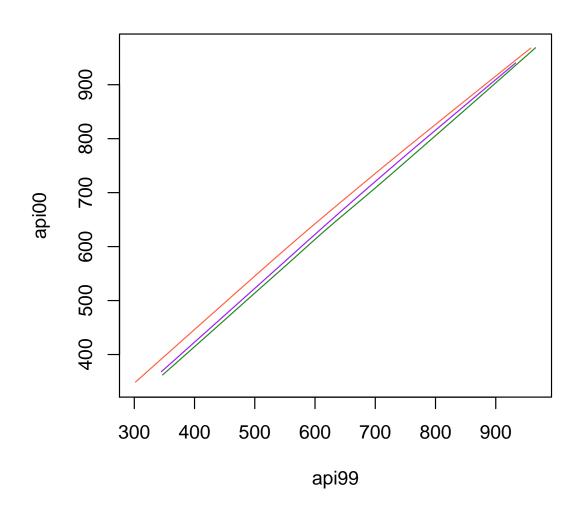


### **S**moothers

We don't plot the data at all, just means (could also plot quantiles with quantreg package)

```
plot(api00~api99,data=apipop,type="n")
 with(subset(apipop, stype=="E"),
      lines(lowess(api99, api00), col="tomato"))
 with(subset(apipop, stype=="H"),
      lines(lowess(api99, api00), col="forestgreen"))
 with(subset(apipop, stype=="M"),
      lines(lowess(api99, api00), col="purple"))
Note the use of type="n"
subset() returns a subset of a data frame.
```

### **Smoothers**



# **Conditioning**

hexbinplot(api00~api99|stype,data=apipop,style="centroid")

