

3. More Advanced Graphics

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Outline

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

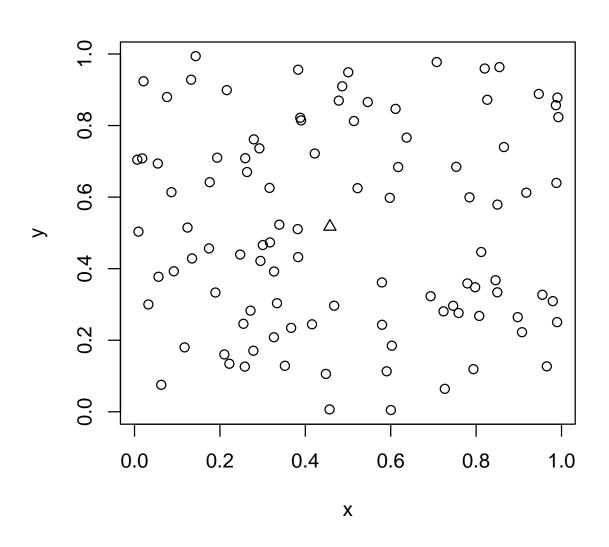
Colour coding

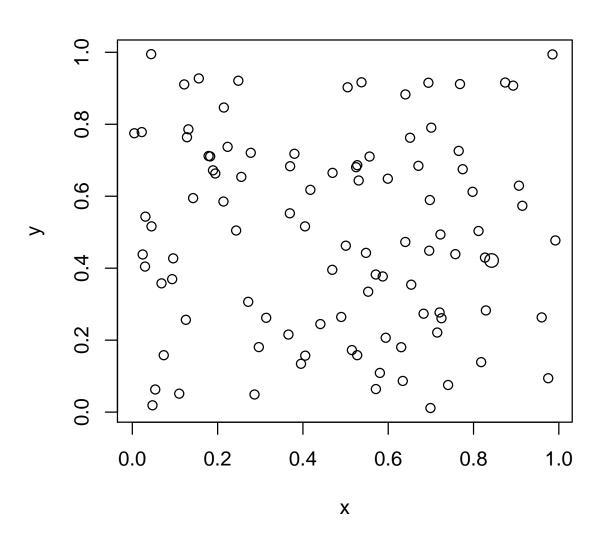
'Simple' plots involve two-dimensional data, which we measure on the x and y axes.

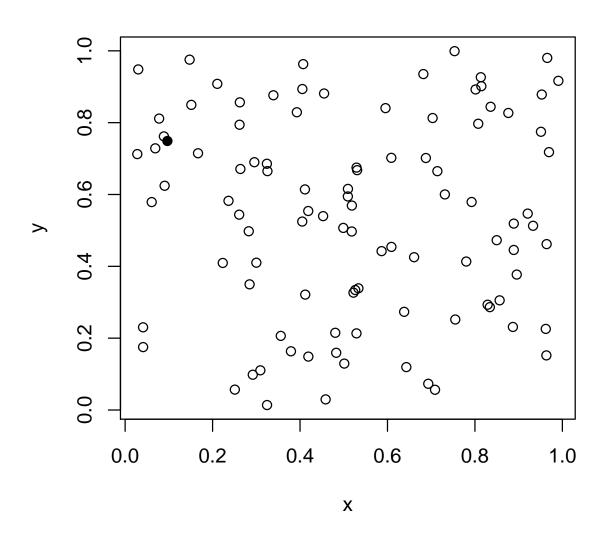
For higher-dimensions, some traditional approaches are;

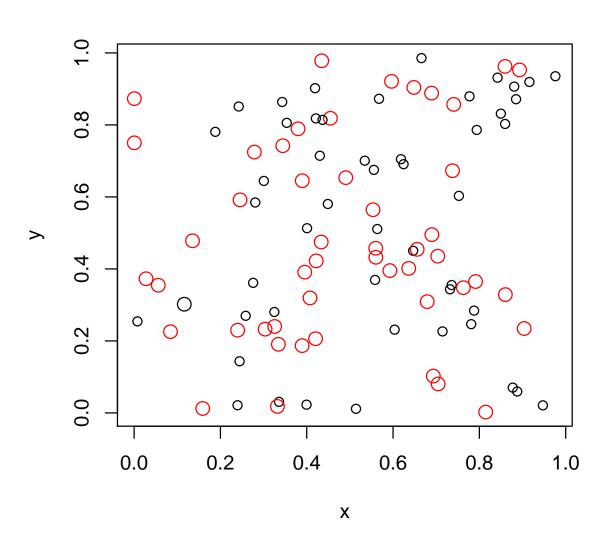
- Different colors for e.g. men, women (col)
- Different-shaped symbols (pch), or different sizes (cex)

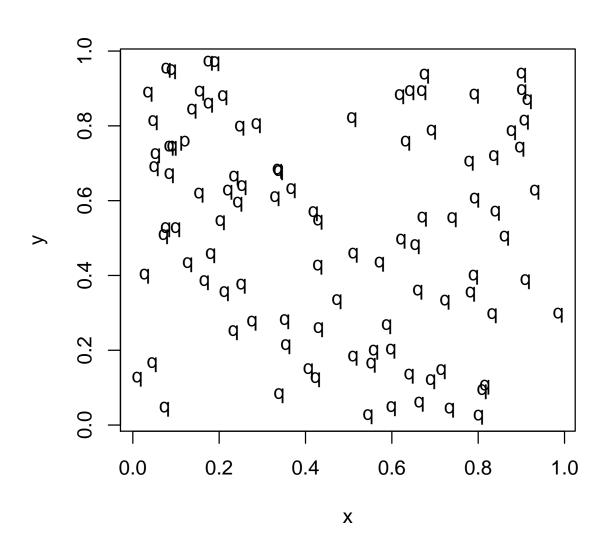
For \leq 100's of data points, modest use of these is fine. But your eye is not good at concentrating e.g. just on the purple points, in a fully Technicolor plot;











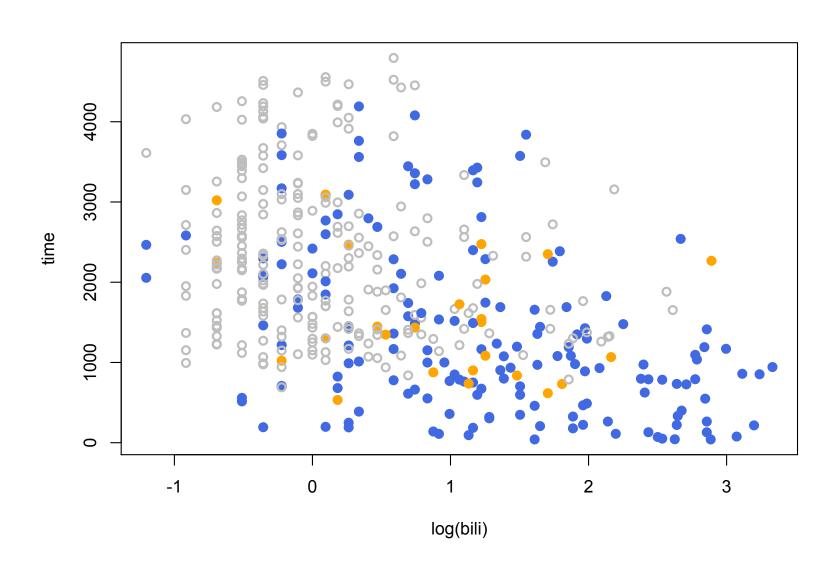
Preattentive perception

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



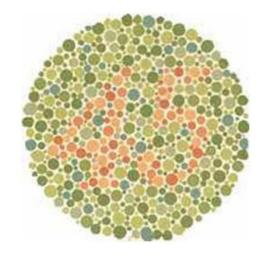
Color schemes

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

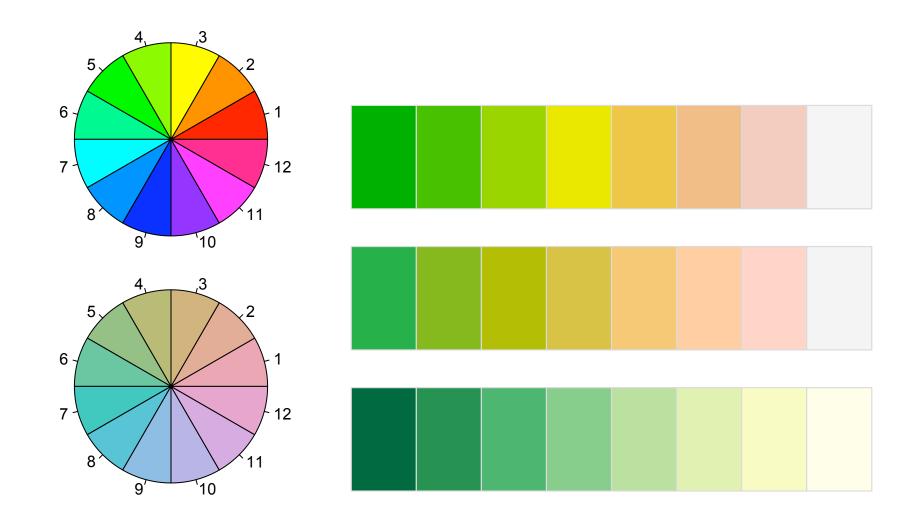
colorspace package has color schemes based on straight lines in a perceptually-based color space (rather than RGB).

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



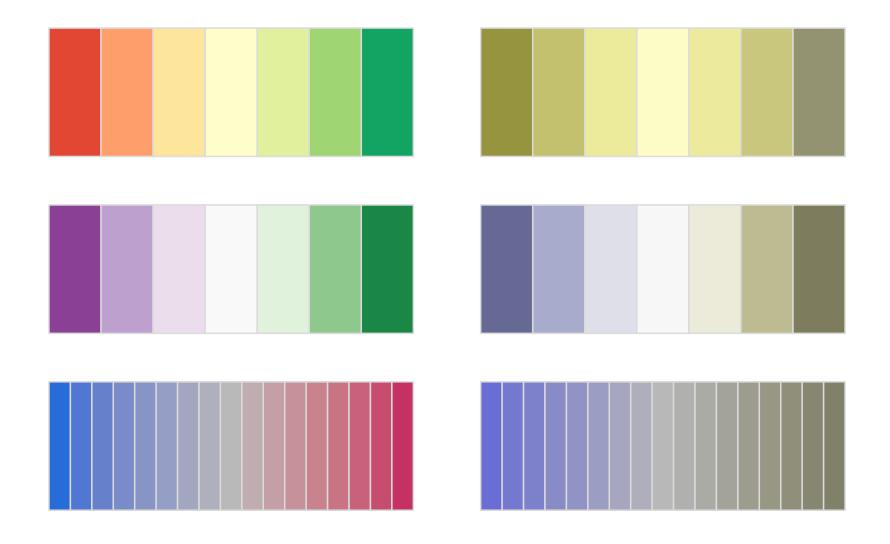
[Code for examples is in file colorpalettes.R on course website]

Color choice



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

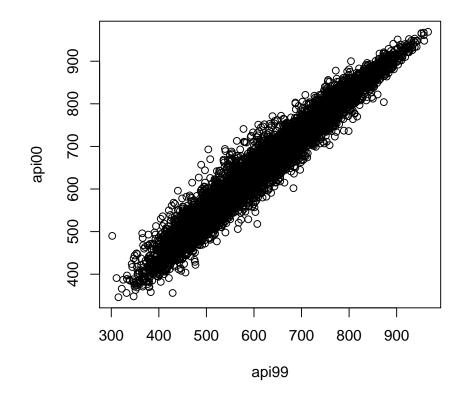
Color blindness



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

Larger data

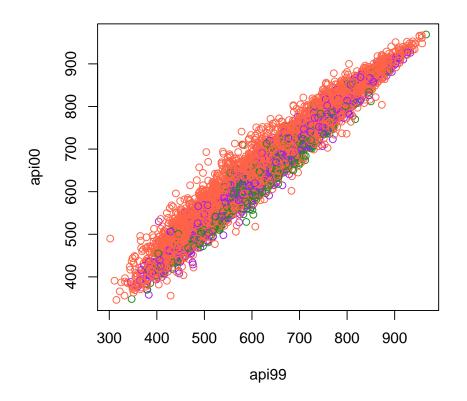
For large(ish) data, 'overlap' is a fundamental problem...



(California Academic Performance Index on 6194 schools)

Larger data

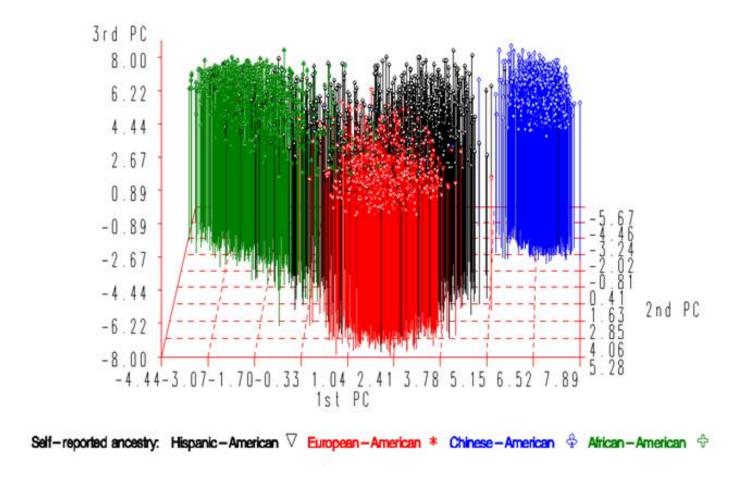
... which remains, when we color-code.



Colors denote Elementary, Middle & High Schools

Larger data

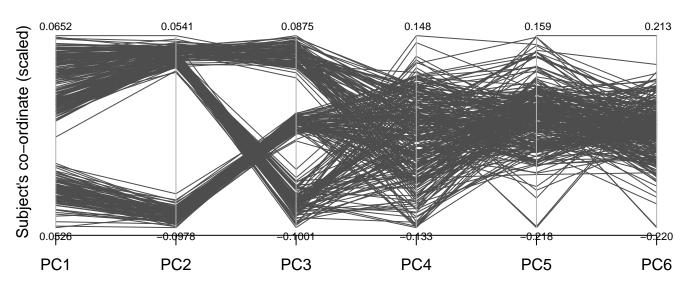
With three dimensions + color-codes, this can happen;



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

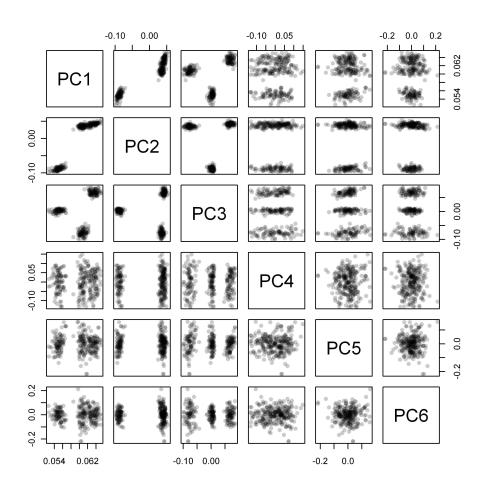
For even higher-dimensional data, scatterplots can not provide adequate summaries. For data where the dimensions can be ordered, the parallel co-ordinates plot is useful;

Leading Principal Components, n=279, 10000 SNPs



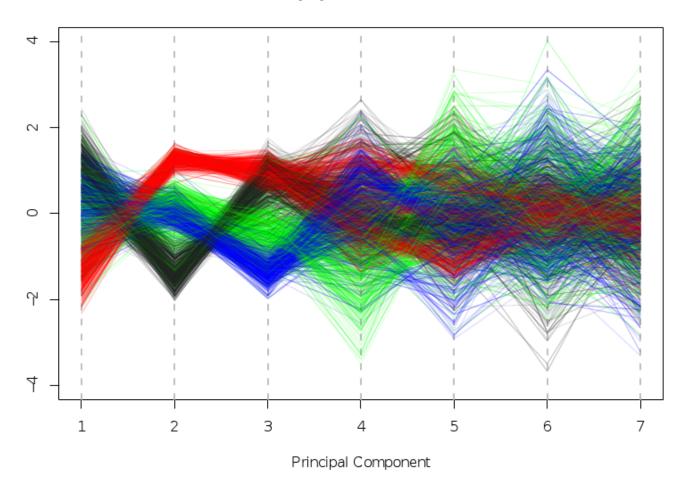
- Each multi-dimensional data point (i.e. each person) is represented by a line not a point
- parcoord() in the MASS package is one simple implementation
 writing your own version is not a big job
- Coloring the lines also helps (example later)
- Scaling of axes, and their vertical positions are arbitrary
- 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%.

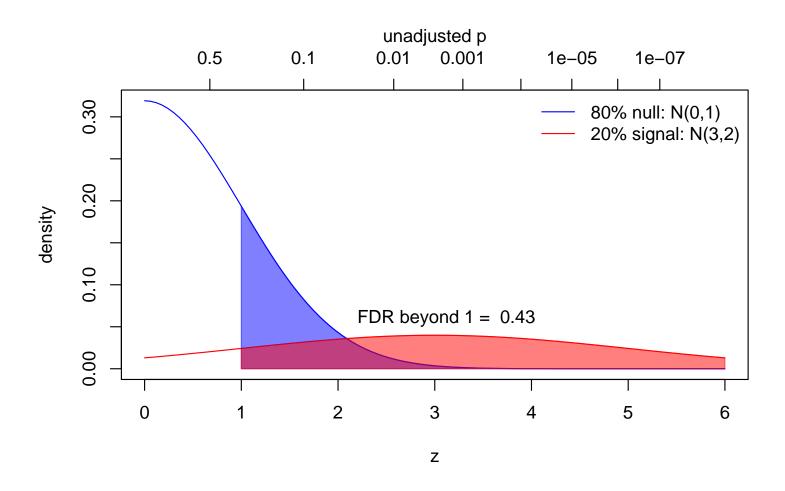
Get from color names to RGB with col2rgb(), and from base 10 to base 16 using format(as.hexmode(11), width=2)

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) $xvals \leftarrow seq(1, 6, 1=101)$ 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)))

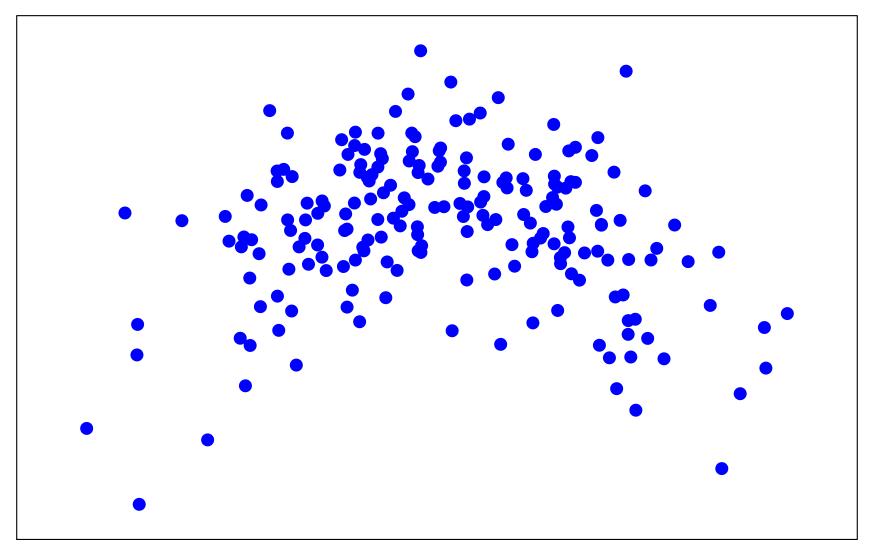
Transparency

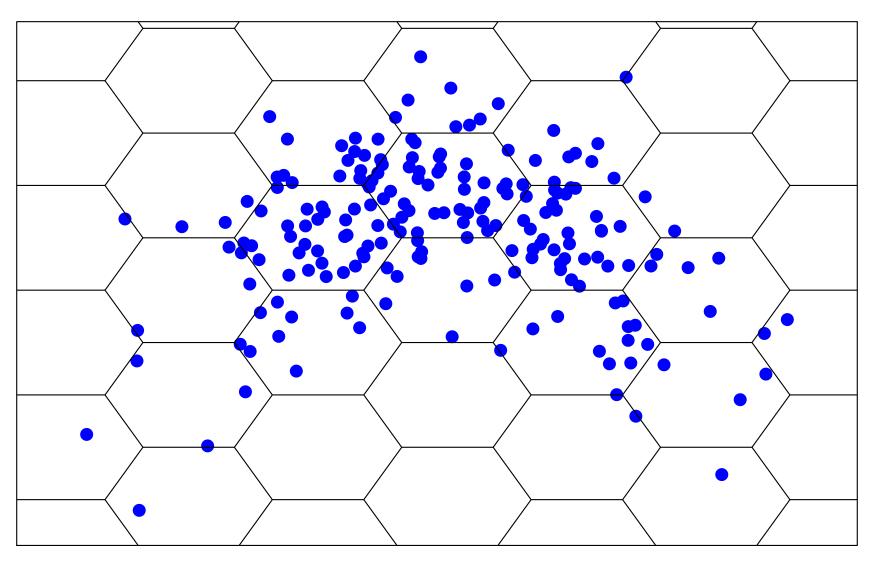
Here's the output;

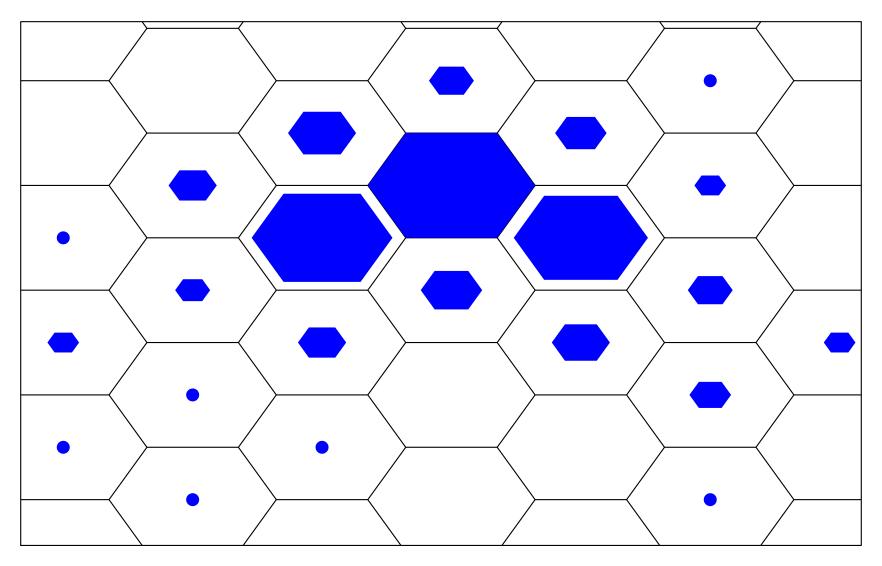


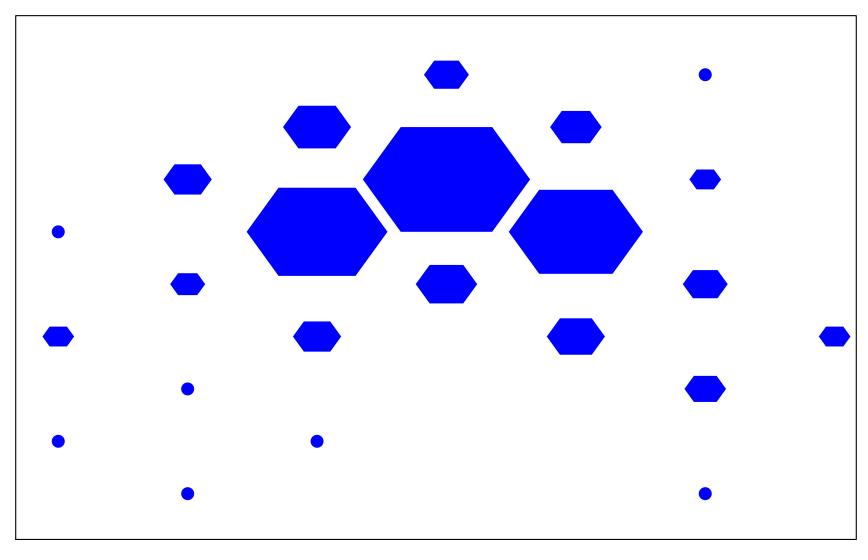
Using transparent plotting symbols is a quick-and-dirty way to adapt scatterplots for use with large datasets.

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.



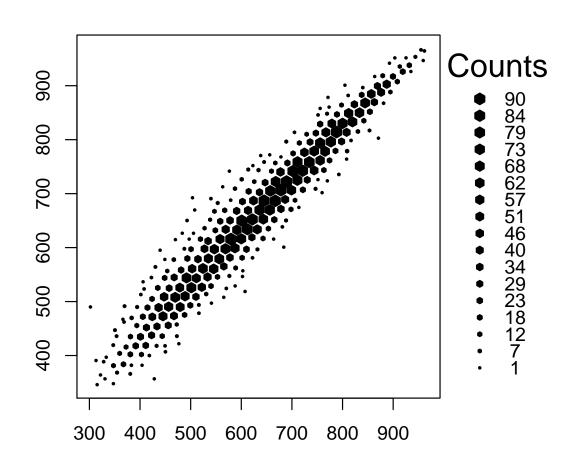






The hexbin() package does all the bin construction, and counting. It has a plot method for its hexbin objects;

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



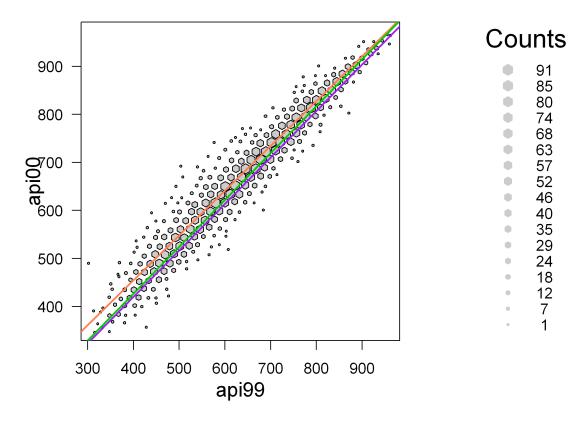
Hexbin is used when you don't *really* care about the exact location of every single point

- 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;



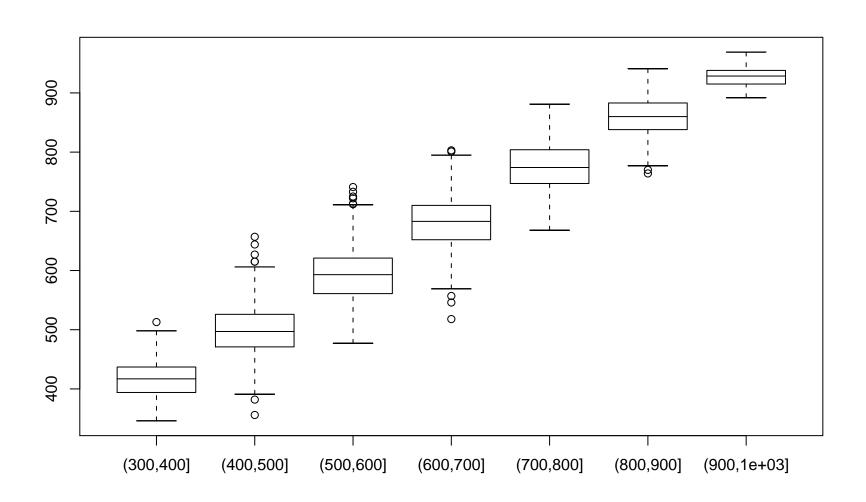
```
lm.e <- coef(lm(api00~api99, data=apipop, subset=stype=="E"))
lm.m <- coef(lm(api00~api99, data=apipop, subset=stype=="M"))
lm.h <- coef(lm(api00~api99, data=apipop, subset=stype=="H"))
hexVP.abline(vp1$plot.vp, lm.e[1], lm.e[2], col="coral")</pre>
```

Large data: multiple groups

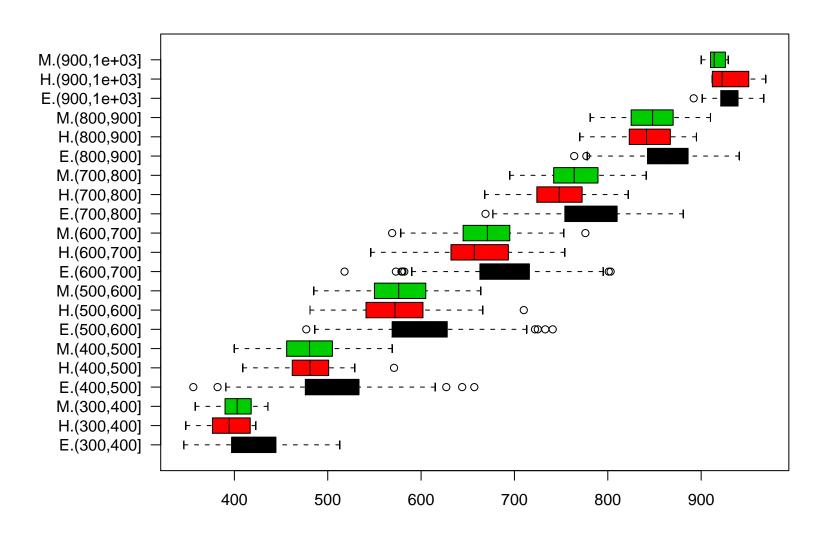
For showing multiple groups, a scatterplot smoother or perhaps boxplots or conditioning may be better.

- 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

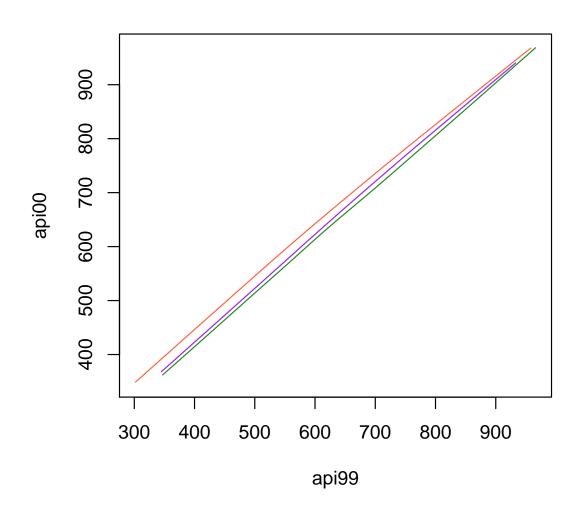


Smoothers

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")

