

**STAT 545A**

**Class meeting #8**

**Monday, October 1, 2012**

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**Department of Statistics and Michael Smith Laboratories**



# Review of last class

Base R graphics: to get great results, you will have to micro-manage every last detail

`par()` is useful function to query, set, and simply learn about all sorts of graphical parameters

Typical workflow: “null” plot to set up co-ordinate system, then slowly add graphical elements

Important details: proper way to log an axis, keep axis labels in readable orientation

# Review of last class

Use professionally designed color palettes ‘as is’, when possible (e.g. RColorBrewer, dichromat).

`colorRamp()` and `colorRampPalette()` are handy for interpolating colors, i.e. expanding a palette that is too small or preparing a discrete palette for use with continuous data

Store color scheme as a `data.frame`; one character variable for the colors plus one or more factors that mimic the factors in your main `data.frame`.

Use `match()` or `merge()` to map factor variables into colors.

The color scheme `data.frame` is handy for making legends and keeping colors consistent across a series of scripts/figures.

Code you see in this lecture can be found in these files:

`bryan-a01-15-latticeStepByStep.R`

`bryan-a01-16-latticePlotGapminderOneYear.R`

`bryan-a01-17-latticeSoln.R`

`bryan-a01-18-latticeDemos.R`

in this directory:

<http://www.stat.ubc.ca/~jenny/notOcto/STAT545A/examples/gapminder/code/>

## Sources I reference heavily in this class

Paul Murrell's book "R Graphics". I own the 2005 edition but note there is a new August 2011 2nd edition, which is obviously more up-to-date and looks substantially expanded.

Home page for 1st ed and 2nd ed (gives R code for figures)

STATSnetBASE (read book online -- I can't 2nd ed there yet)

Google books search

Deepayan Sarkar's book "Lattice"

webpage w/ all code in book

SpringerLink (read book online)

Google books search

For today, focus especially on Ch 5 Scatter Plots and Extensions.

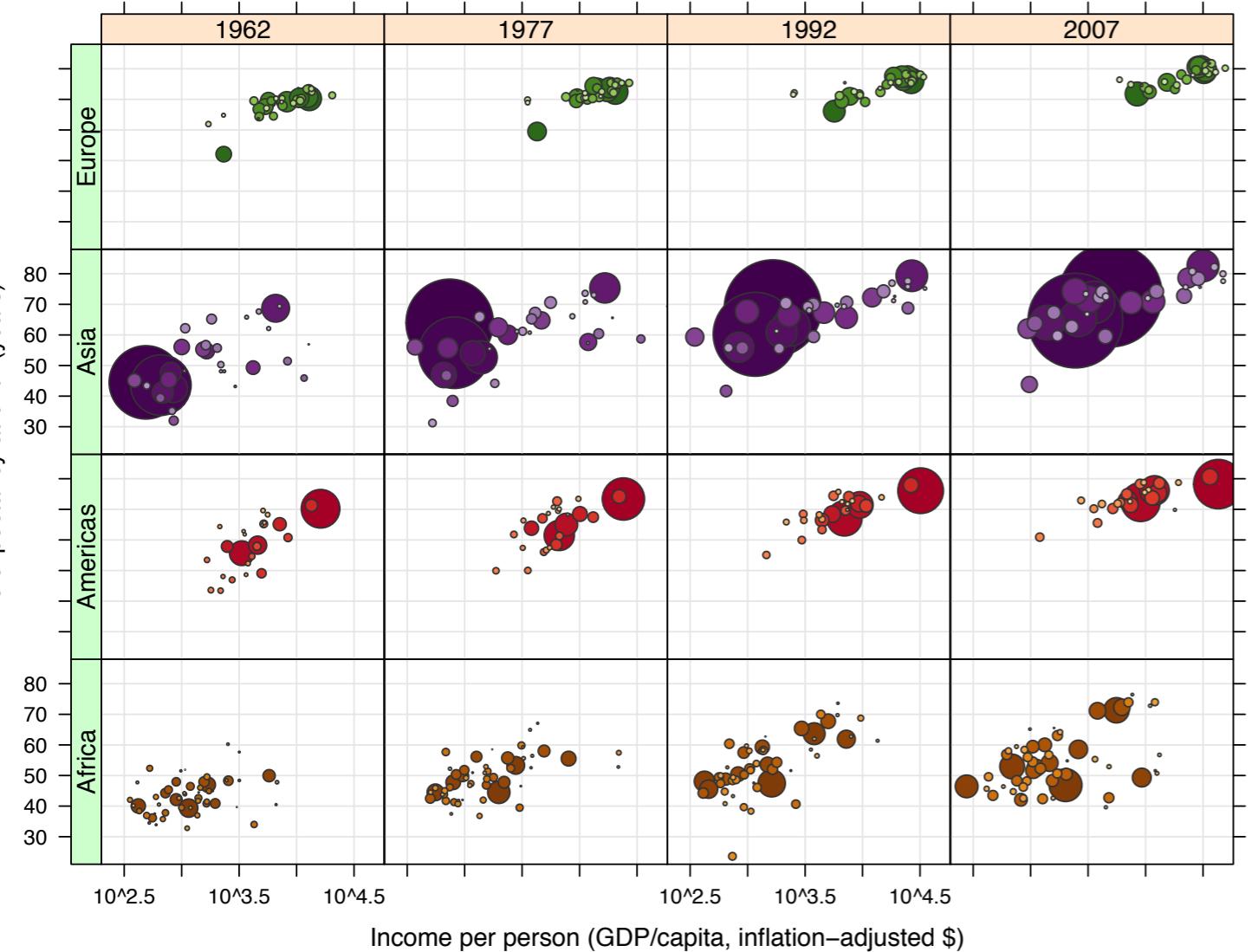
Please recall .....

## Two main goals for statistical graphics

- To facilitate comparisons.
- To identify trends.

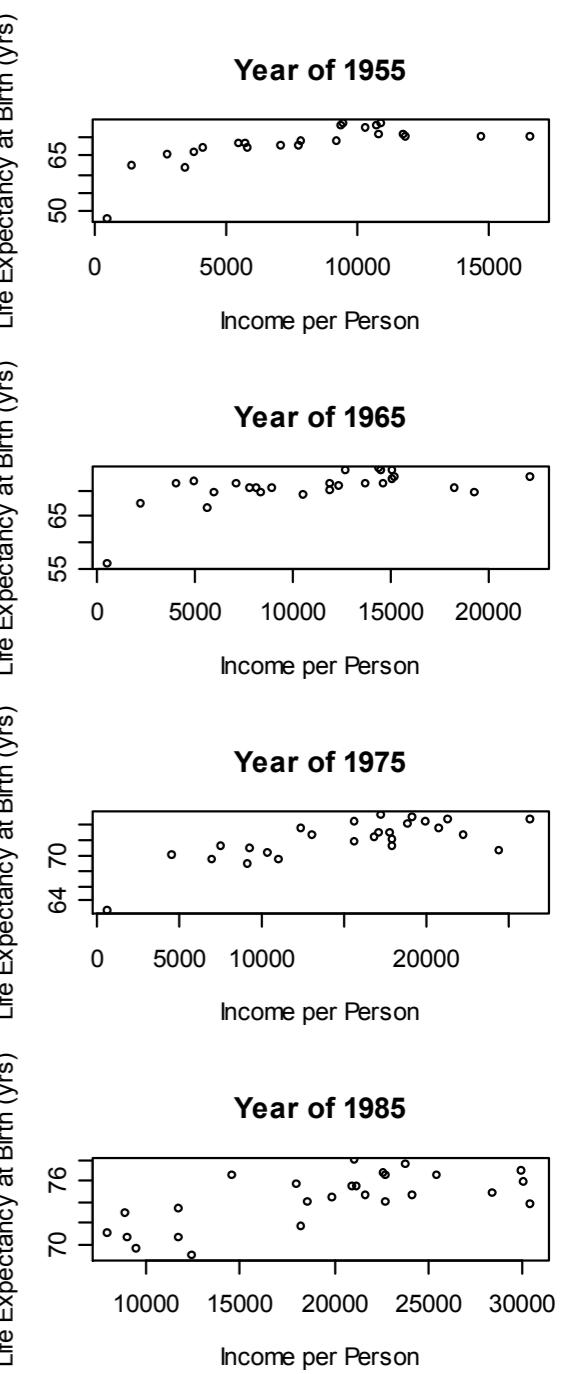
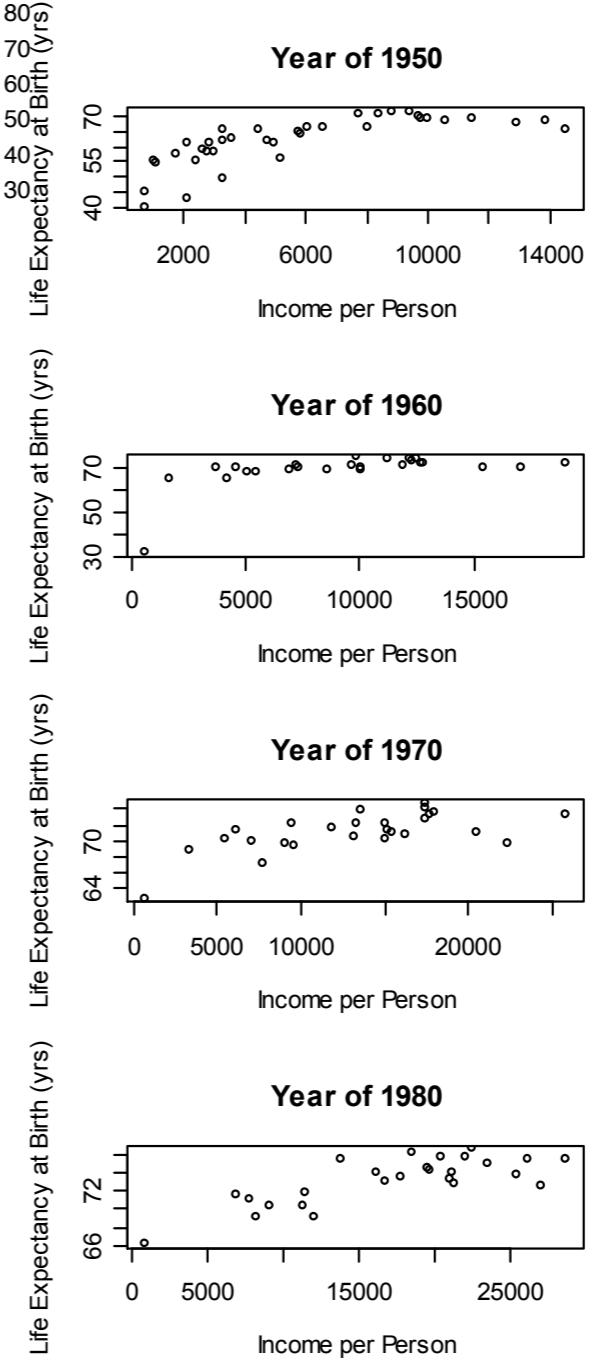
**lattice graphics are simply better  
than traditional graphics for  
achieving these goals**

# lattice



# traditional

## Assignment 1: Best Set of Graphs



# Trellis suite from S/S-PLUS, multi-panel conditioning

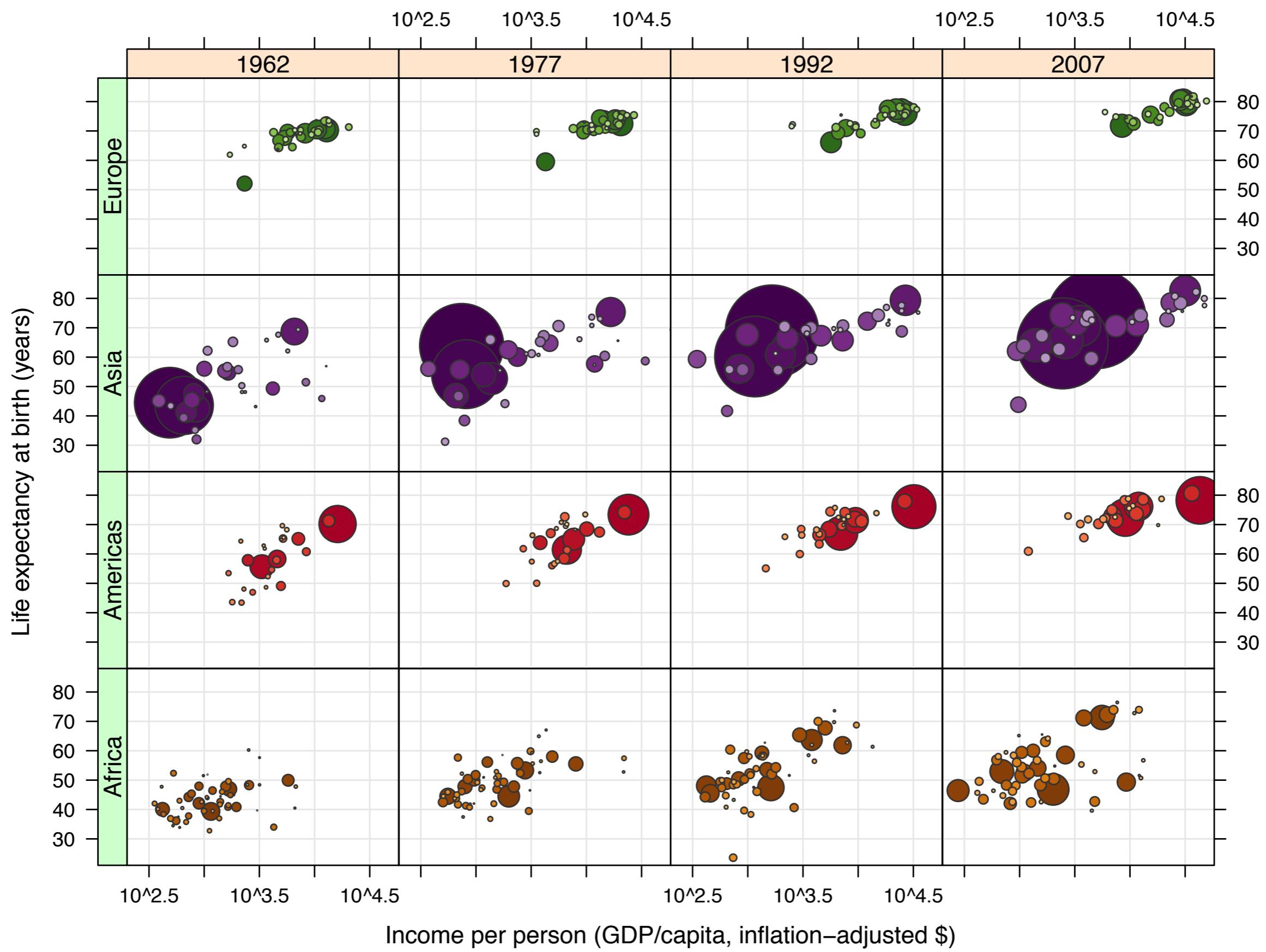
lattice: self-contained independent implementation of Trellis in R, using the low-levels tools of grid (vs. “traditional graphics”)

good news: less need to exert fine control

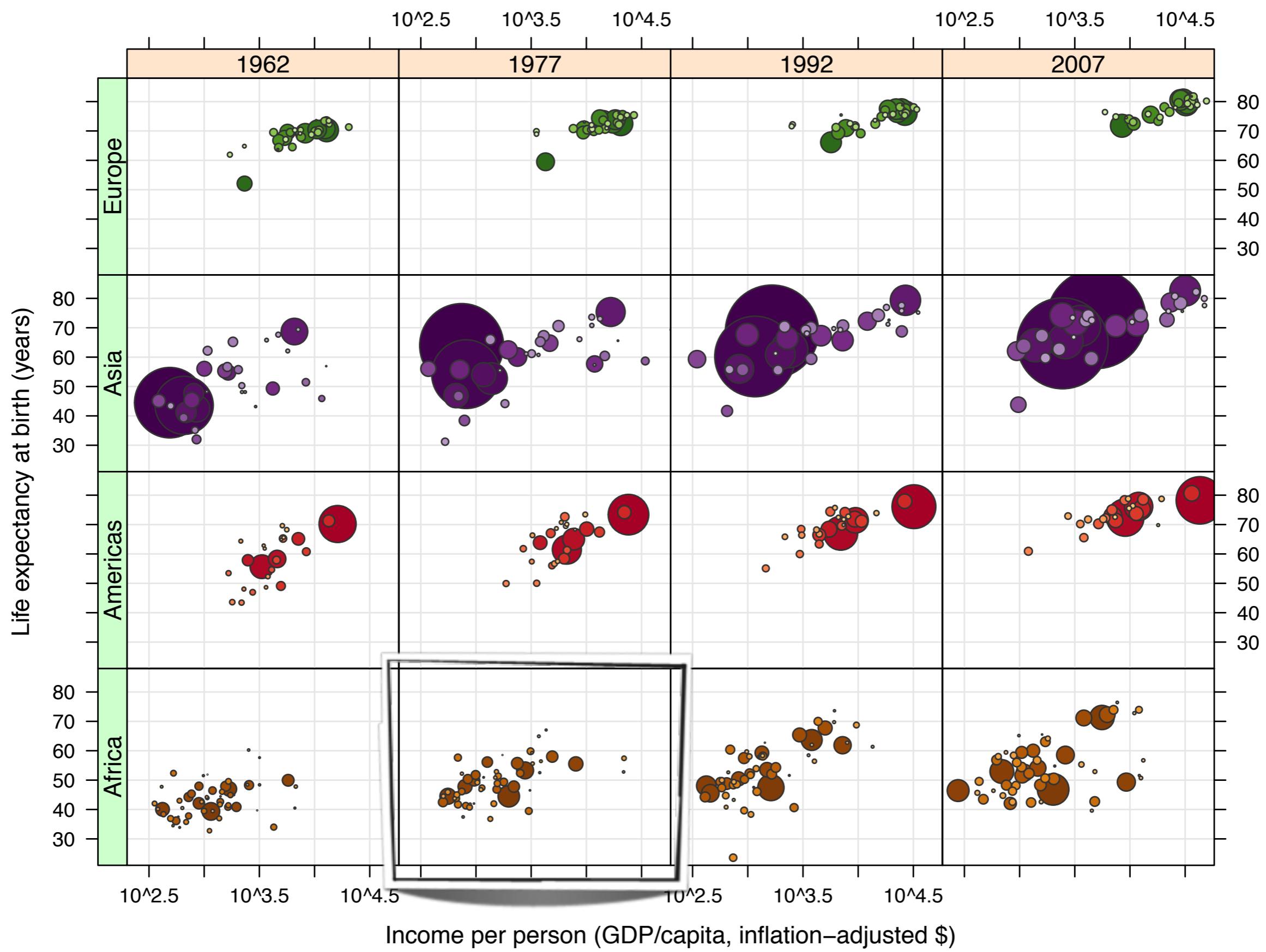
bad news: more difficult (?) and/or more unfamiliar (?) to exert fine control ... arises from the design of the system (“one calls says it all”) and from grid (which most of us don’t know much about)

bottom line: lattice is still worth mastering, no question

\*ggplot2 -- another interesting option that came on the scene later than lattice

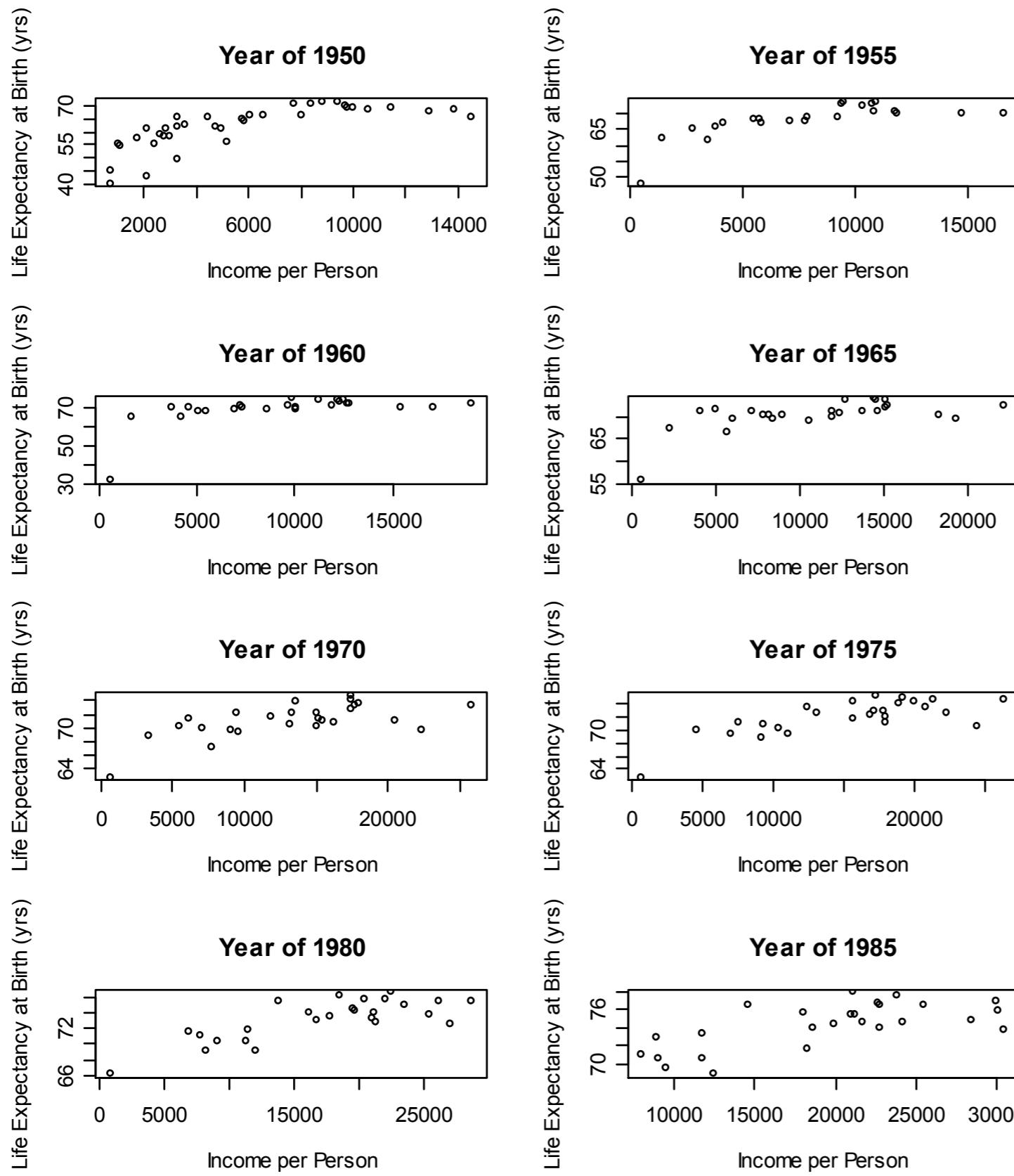


This is what is meant by multi-panel conditioning.



This is a panel.

## Assignment 1: Best Set of Graphs



An excellent demonstration of how traditional graphics fails at such tasks.

hello whitespace!

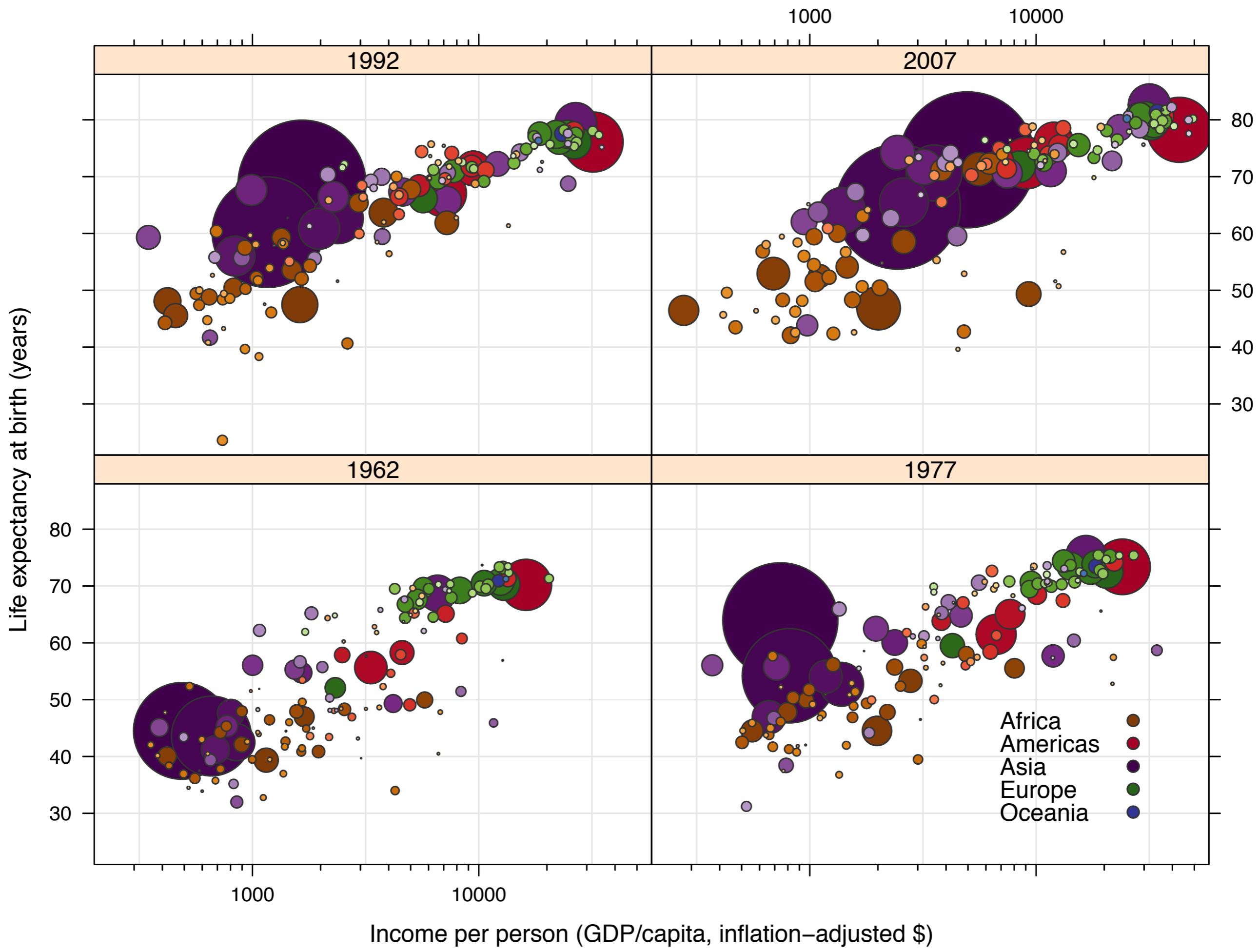
non-common axes

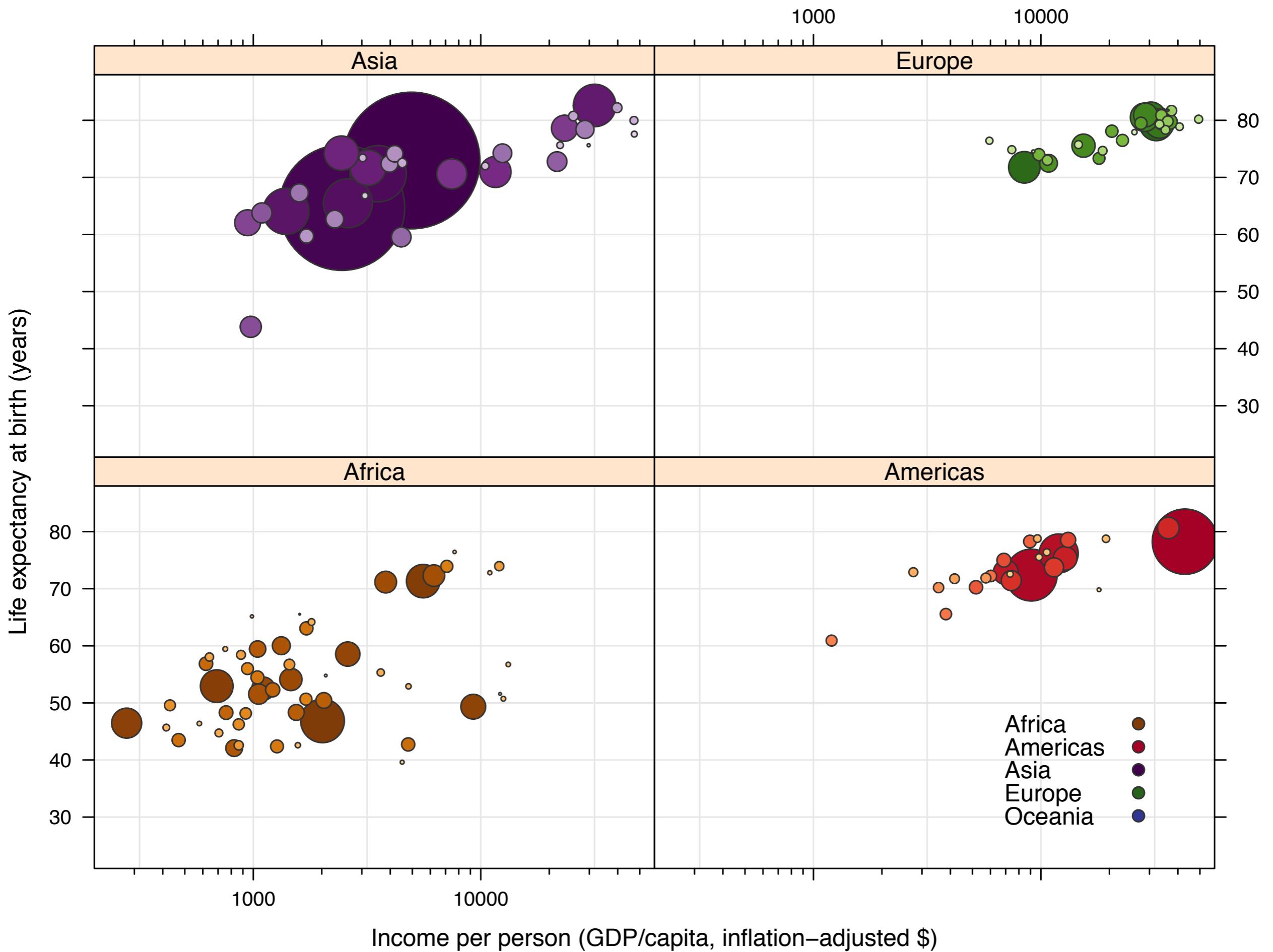
repeated elements, e.g.  
axis labels

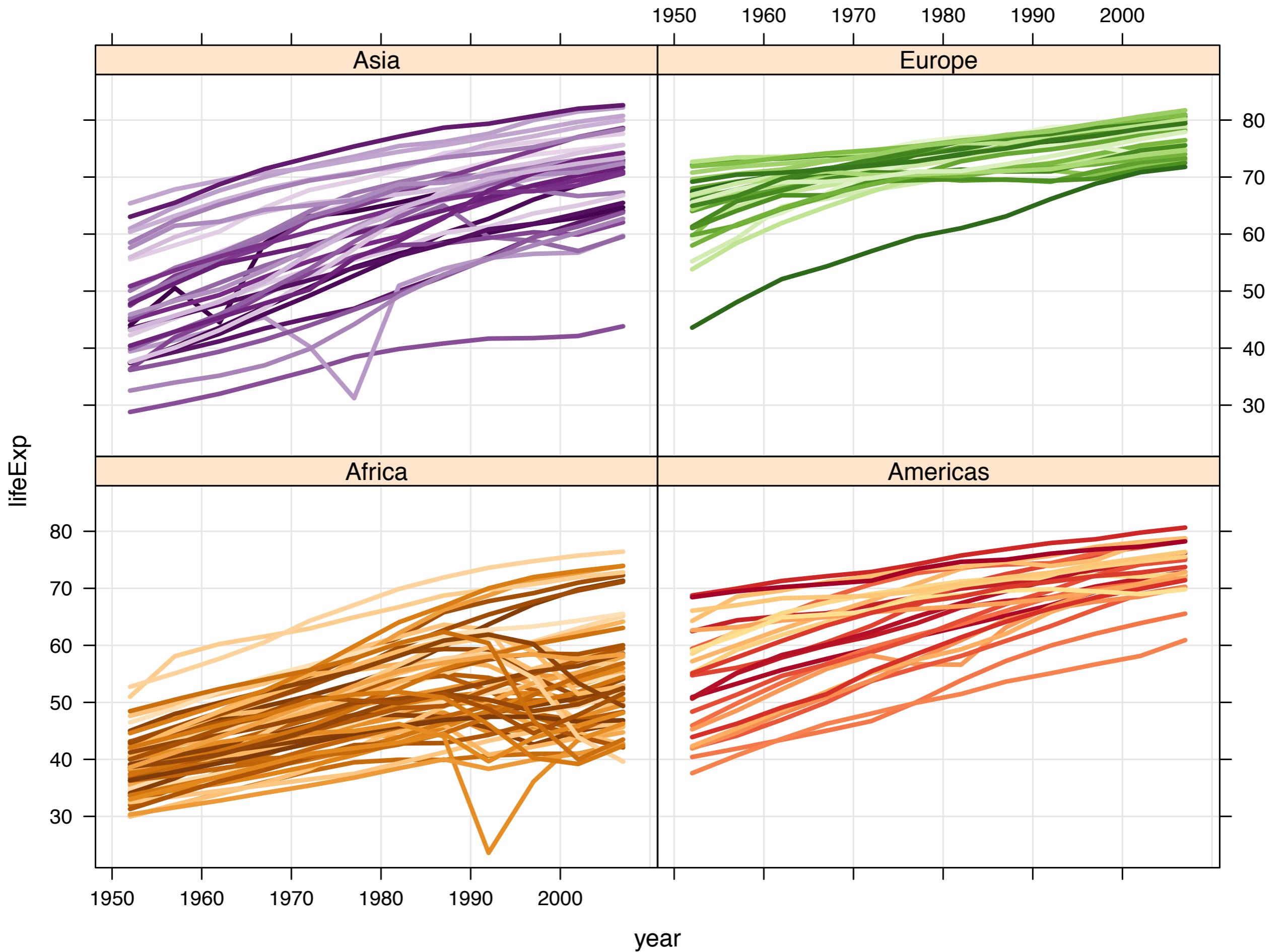
misallocation of space,  
too much spent on  
annotation, not enough  
on data

# **lattice: good for making cohesive groups of plots**

- Main reason for lattice (and its inspiration, Trellis) was to combine plots on one page in a sensible manner
- (In real life, multi-panel conditioning and other visual features are more used and more useful than animation)
- lattice is a **HUGE** improvement over base R graphics for this task
- lattice often produces better plots, at default settings, even for single plots
- It is harder to customize and adorn lattice plots, compared to base R graphics (but the need for this is also greatly reduced).





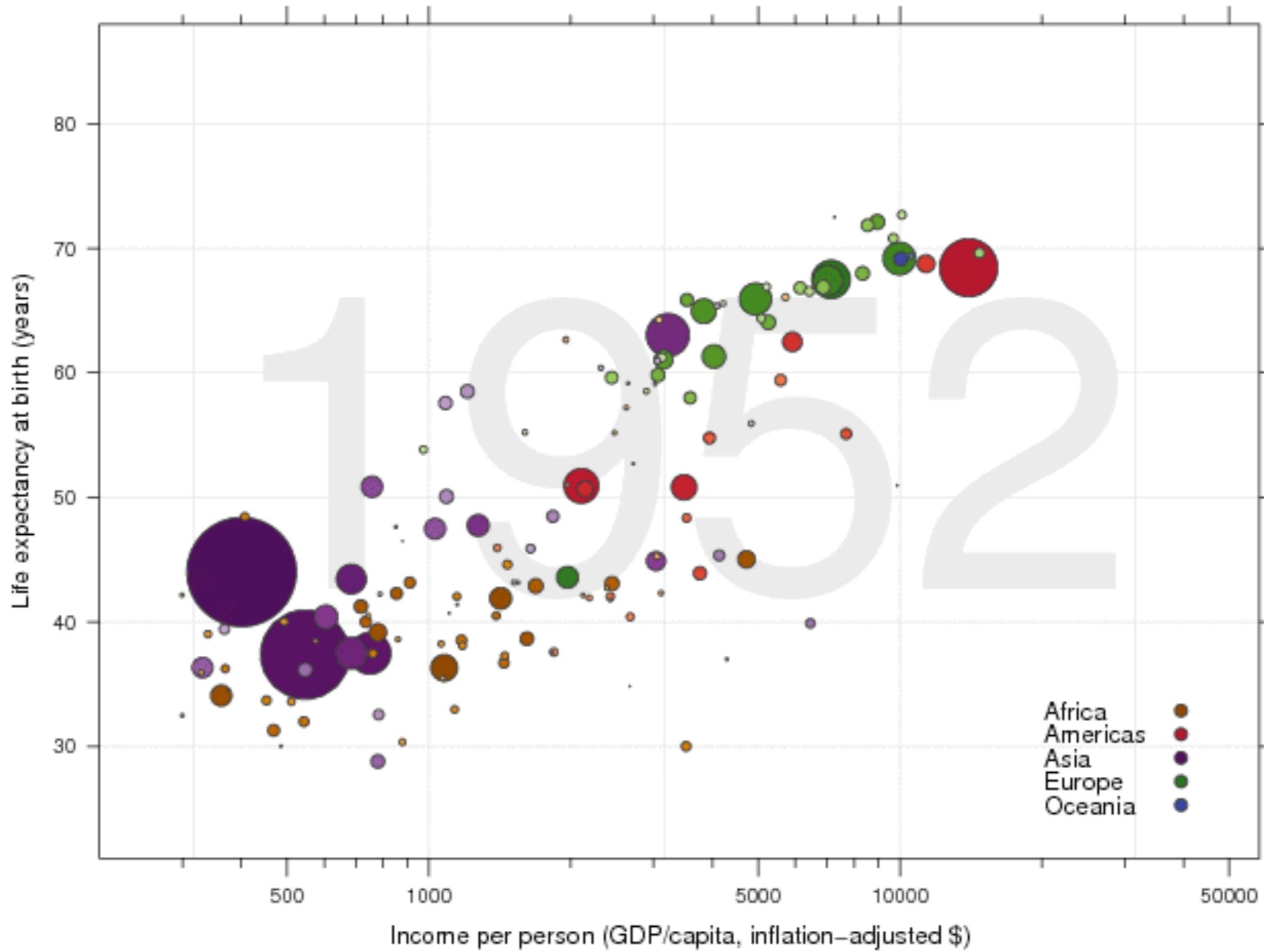


# lattice vs. base R graphics

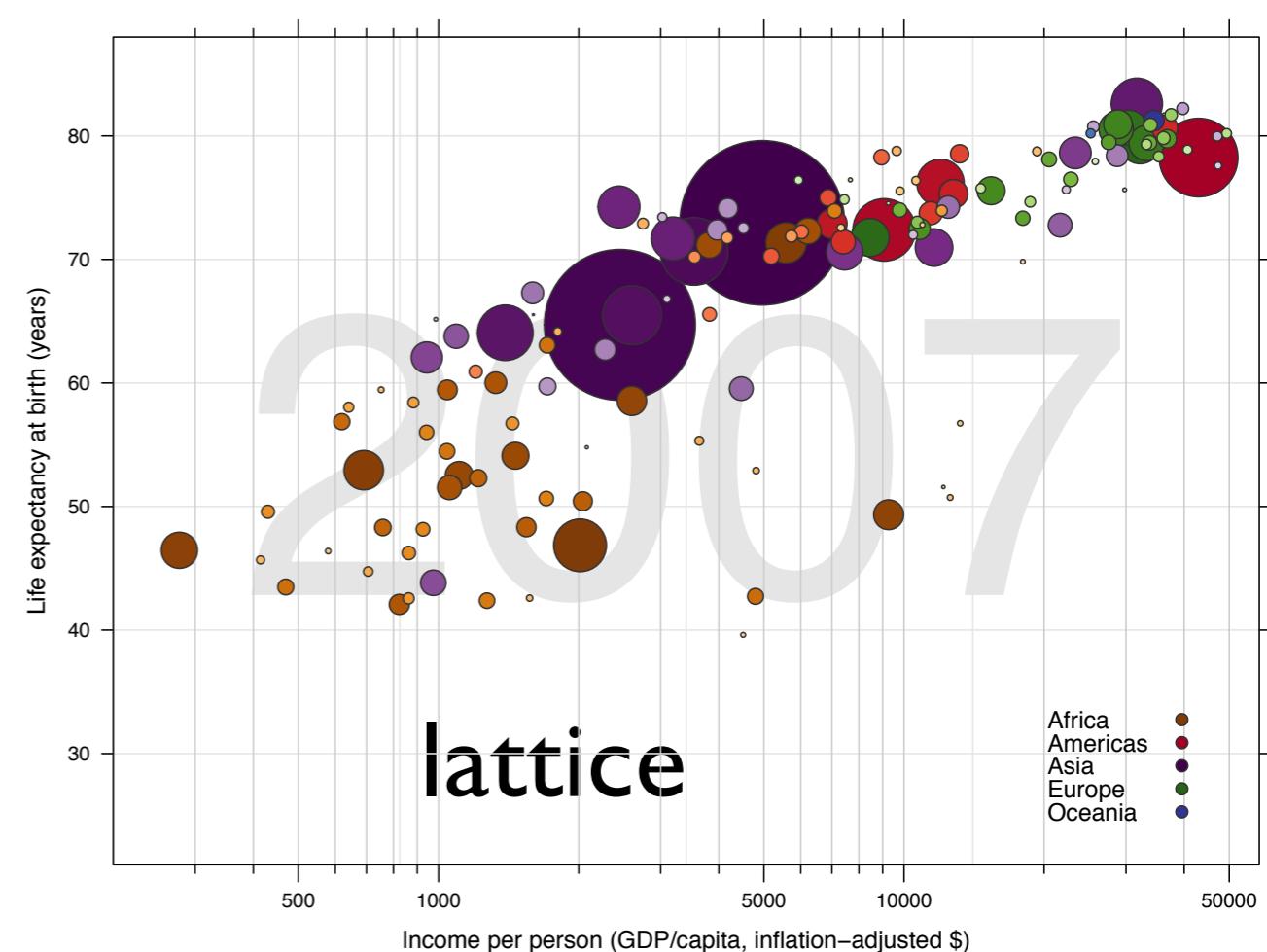
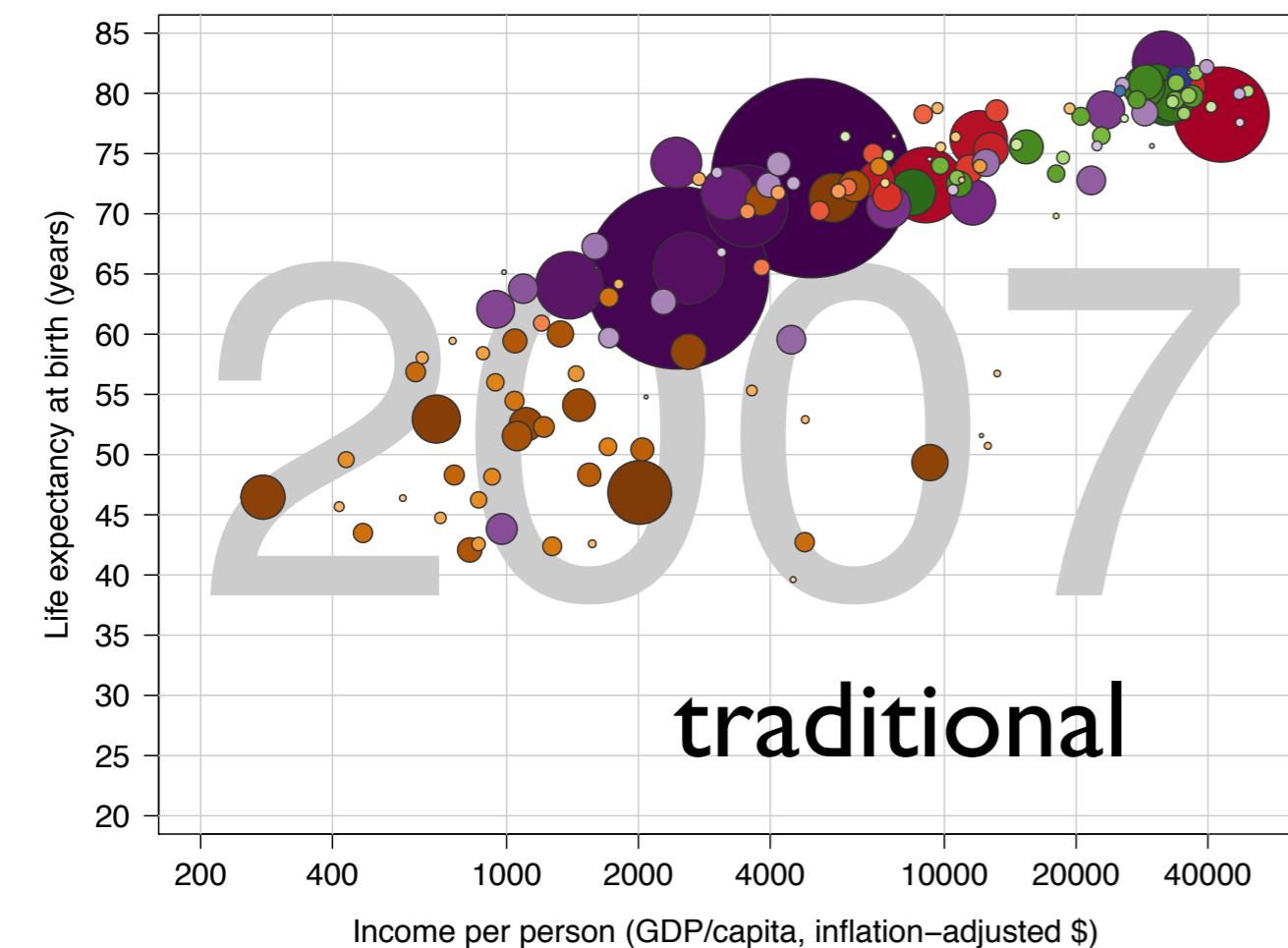
- Know which tool to use when.
- Good rule of thumb:
  - Want a decent plot in 10 mins? Use lattice. (Presumes you learn enough about lattice to get sthg decent in 10 mins.)
  - Want a highly customized plot in ~1 hr and you don't need multi-panel conditioning? Try base graphics. (Presumes you learn enough about adding elements to basic plots to achieve this.)
  - Want a customized plot in 1 day? Do you need multi-panel conditioning? Will you make similar plots again? Or will this be featured in a paper or talk? Use lattice and do what needs to be done, e.g. write a custom panel function.

# lattice vs. base R graphics

- Know what to expect, how to work.
- Traditional graphics: Call `plot()` to set up a coordinate system (and maybe print nothing else) and slowly add your elements.
- Lattice, 80% of the time: Call `xyplot()`, for example, and use oodles of arguments to specify what you want.
- Lattice, other 20% of the time: Re-define some of the standard functions or lists lattice uses to control certain elements, i.e. modify or extend the default to suit your purposes, and then call `xyplot()`, probably still using oodles of arguments.



Produced with lattice



# Scatterplot

See Ch. 5 in  
Sarkar (2008).

- Graph of two quantitative variables against each other in the plane.
- Single most important graphic in your statistical life.
  - Always plot the data BEFORE doing anything else.
- Things to consider and control:
  - join points? convey extra information via color or symbols or size?
  - juxtapose several scatterplots of  $Y$  vs  $X$  for different values of  $Z$ , a categorical variable?
  - add a regression line or scatterplot smooth?
  - use generalized scatterplot to avoid overplotting?

Read R documentation  
for `xyplot()` and  
`panel.xyplot()`

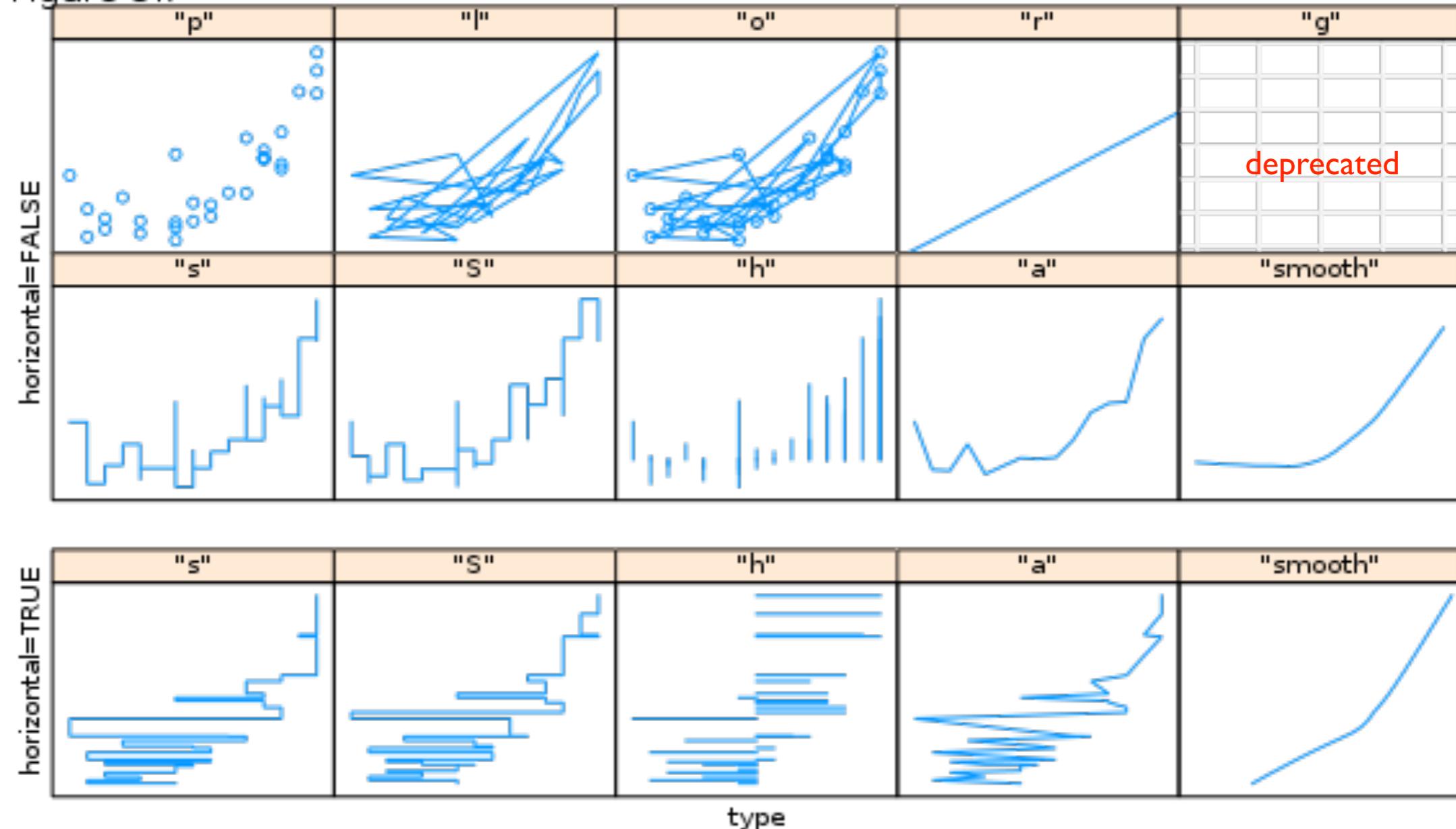
# `xyplot()`

- The first lattice function to learn and master.
- The `type` argument is incredibly useful. Use it!
- It is a character vector consisting of one or more of <read the help file for the values>. If `type` has more than one element, an attempt is made to combine the effect of each of the components.

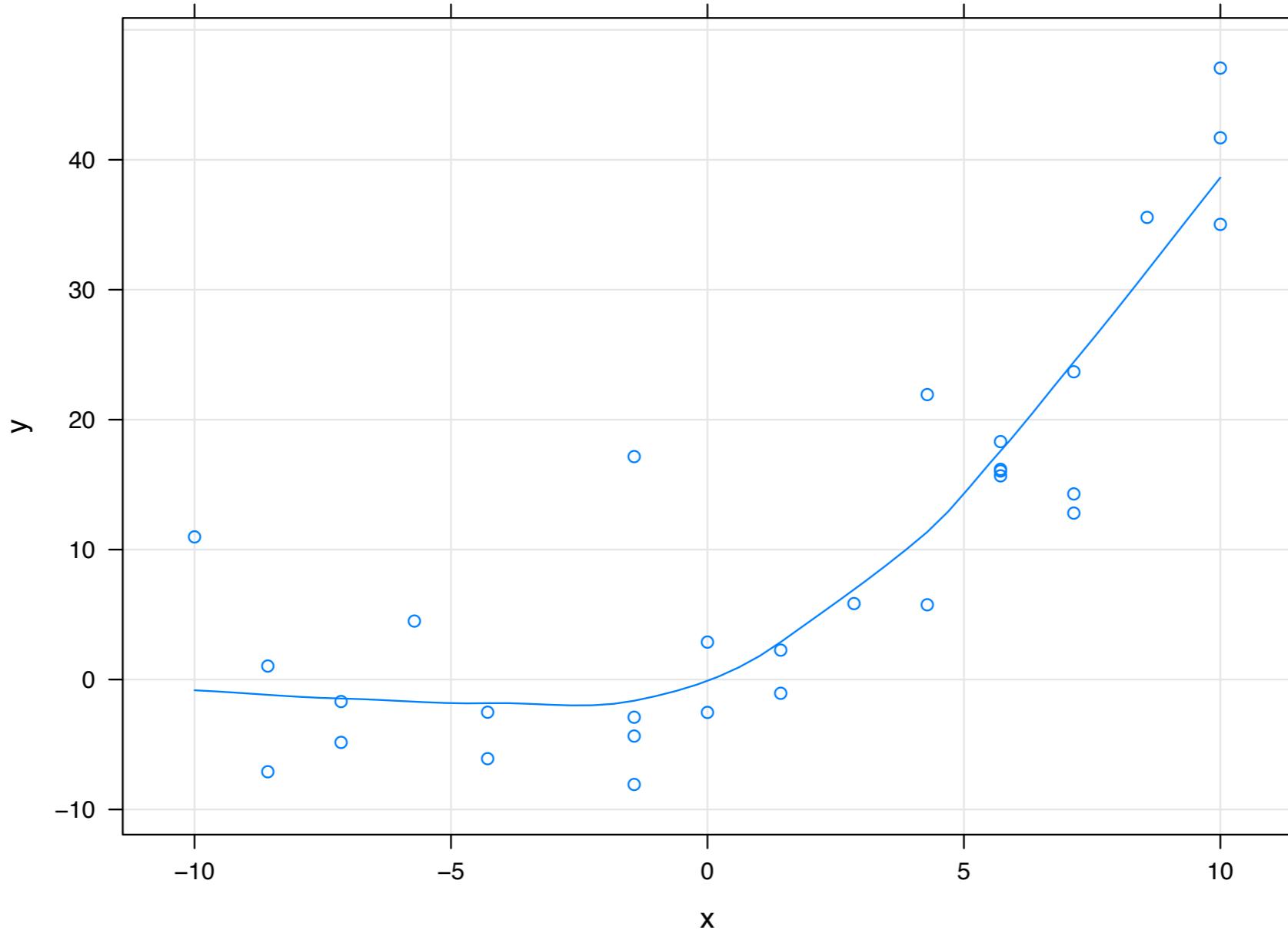
<code>type</code>	Effect	Panel function
"p"	Plot points	
"l"	Join points by lines	
"b"	Both points and lines	
"o"	Points and lines overlaid	
"S", "s"	Plot as step function	
"h"	Drop lines to origin ("histogram-like")	
"a"	Join by lines after averaging	<code>panel.average()</code>
"r"	Plot regression line	<code>panel.lmline()</code>
"smooth"	Plot LOESS smooth	<code>panel.loess()</code>
"g"	Plot a reference grid	<code>panel.grid()</code>
		<b>deprecated</b>

Table 5.1. The effect of various values of the `type` argument in `panel.xyplot()`.  
From Ch. 5 in Sarkar (2008).

Figure 5.7



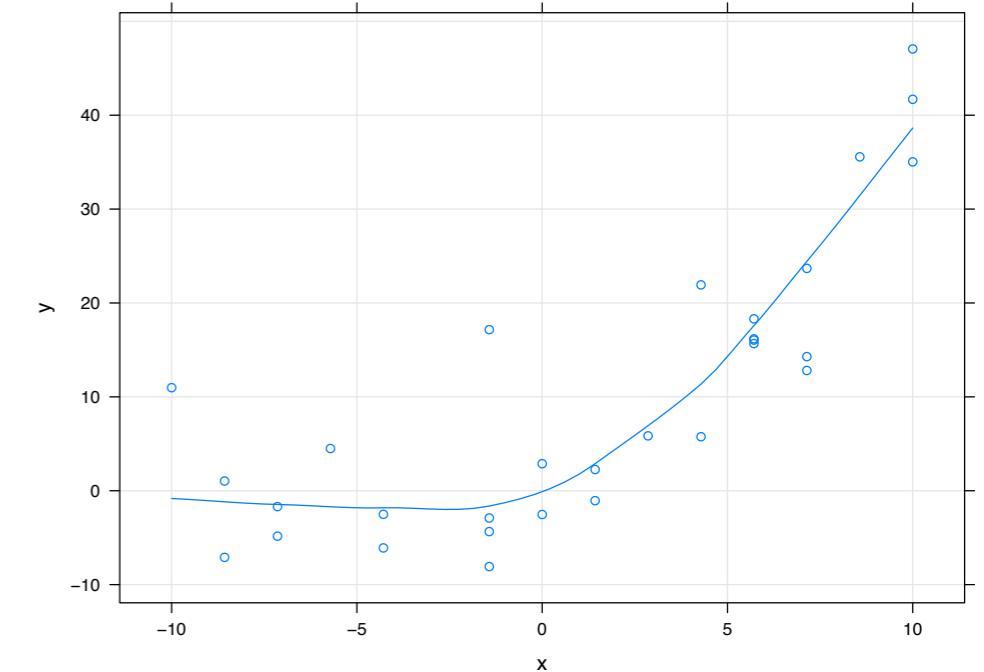
From Ch. 5 in Sarkar (2008).



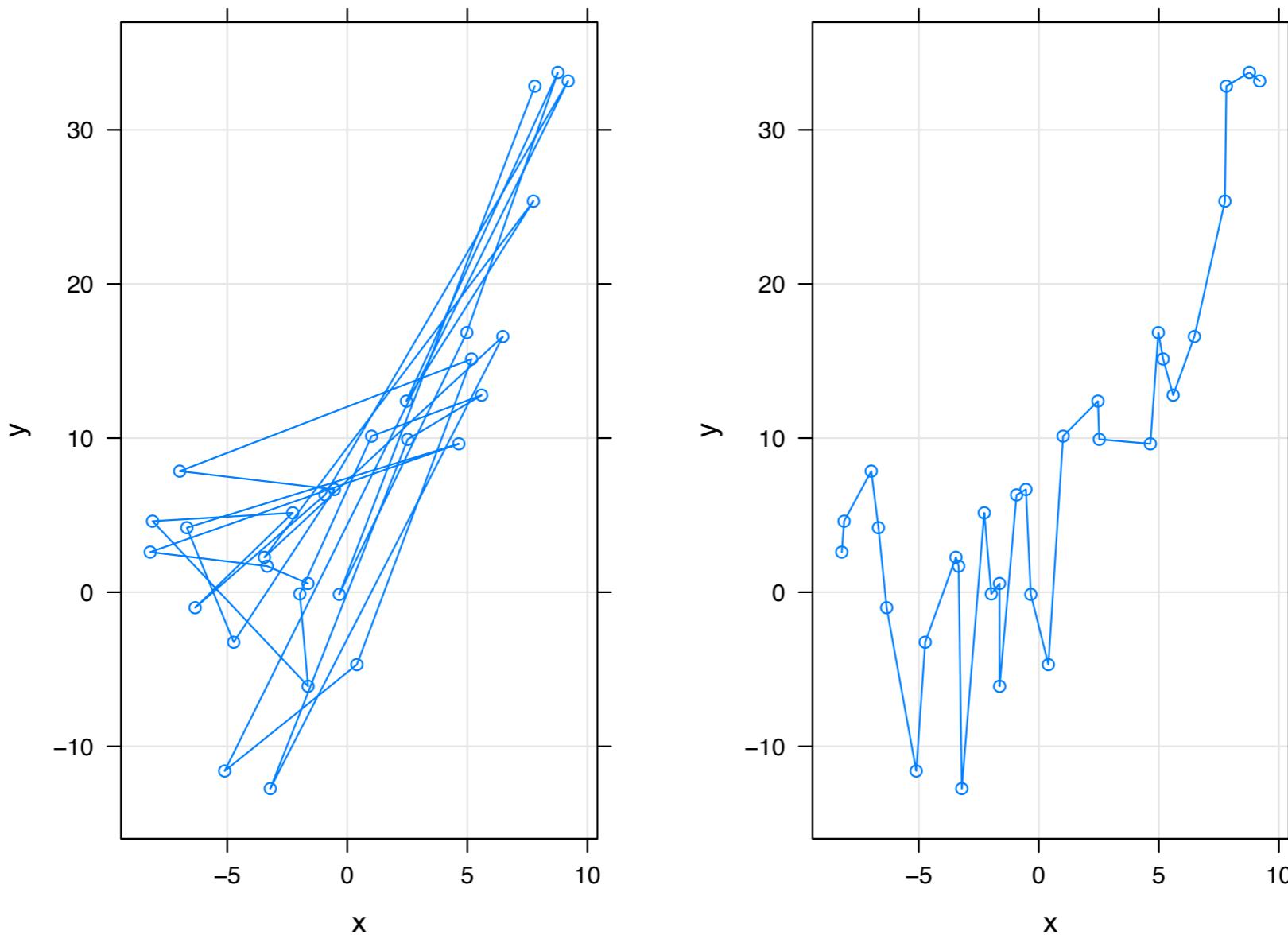
```
xyplot(y ~ x, type = c("g", "p", "smooth"))
```

Reference grids and other common elements greatly facilitate the interpretation and comparison of multi-panel displays.

If you want to get some quick info on  $Y \sim X$ , prior to full-blown model selection, the loess smooth is handy.



... type = c("g", "p", "smooth"))



```

x <- runif(n = 30, min = -10, max = 10)
y <- x + 0.25 * (x + 1)^2 + rnorm(length(x), sd = 5)
uglyPlot <- xyplot(y ~ x, type = c("g", "b"))
xOrd <- order(x); x <- x[xOrd]; y <- y[xOrd]
prettyPlot <- xyplot(y ~ x, type = c("g", "b"))
print(uglyPlot, pos = c(0, 0, 0.5, 1), more = TRUE)
print(prettyPlot, pos = c(0.5, 0, 1, 1), more = FALSE)

```

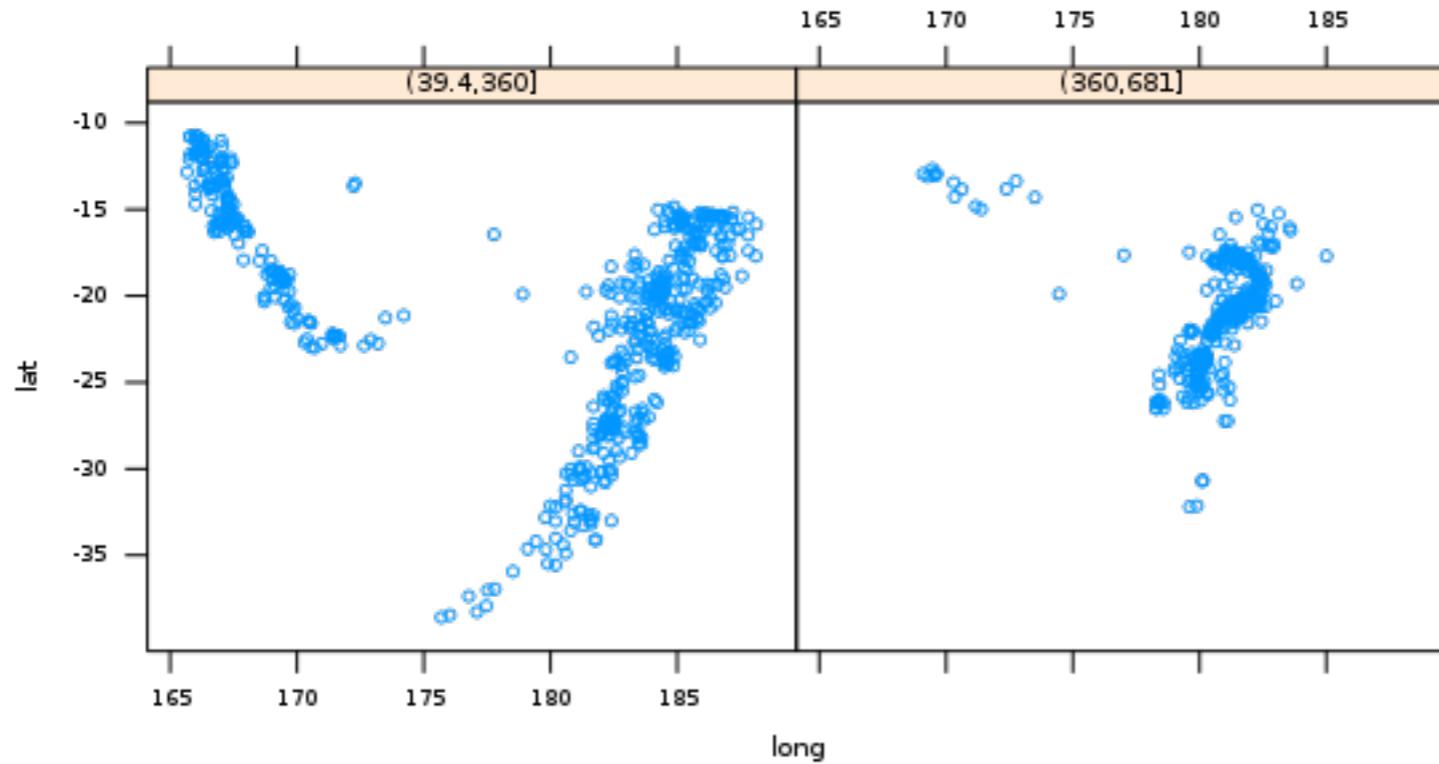
If you connect the dots, make sure to sort first!

# quakes data

See Sarkar (2008).

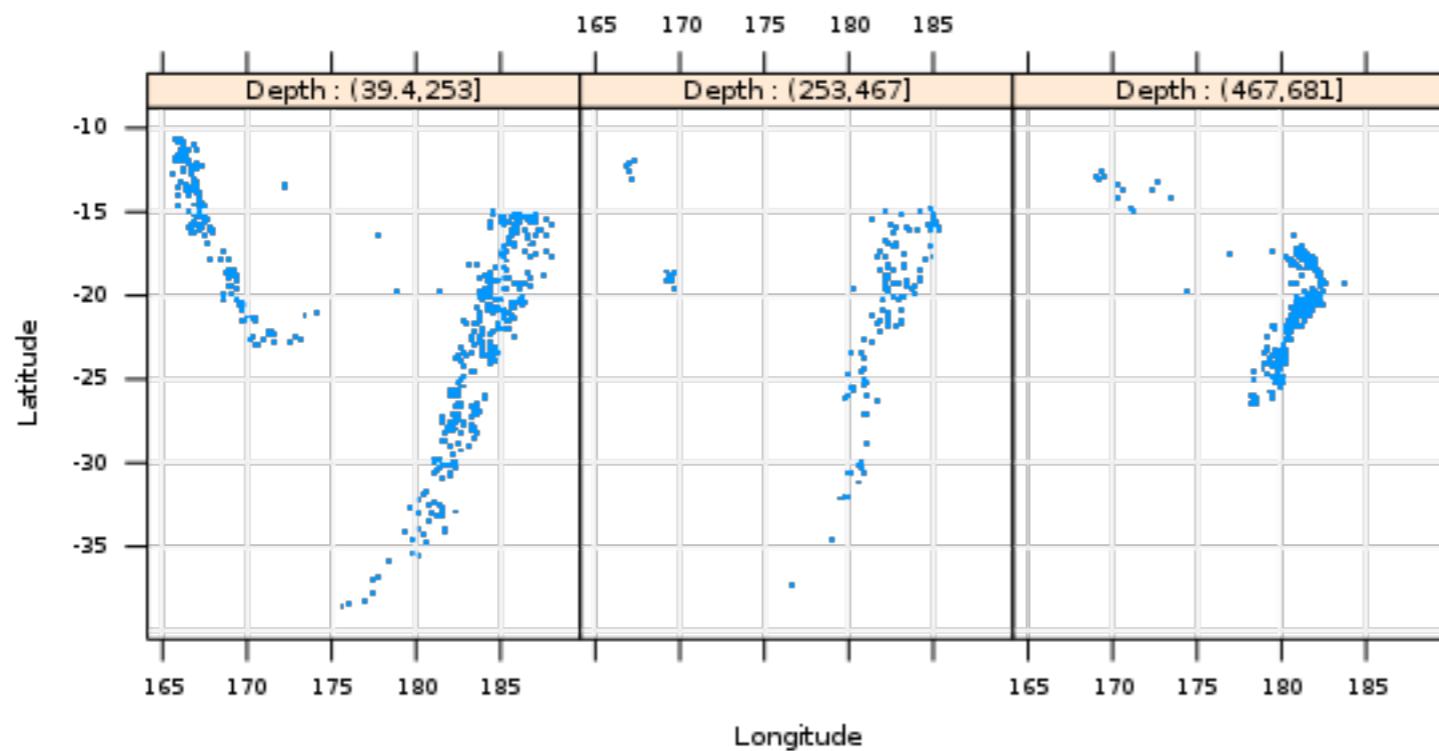
- Records location (latitude, longitude, depth) and magnitude of earthquakes near Fiji since 1964
- Latitude and longitude are (X,Y)
- The third quantitative variable, depth, can be turned into a ‘shingle’, for graphing purposes. See Sarkar 2.1.4.
- A ‘shingle’ is similar to converting a quantitative variable into a categorical one, but we potentially allow the intervals to overlap. Shingles are useful as the conditioning variable in lattice plots, i.e. the variable whose levels correspond to distinct panels.
- Functions to create shingles: `cut()`, `equal.count()`
- Code for following figures available [here](#)

Figure 5.1



Depth turned into a shingle w/ non-overlapping intervals, i.e. a factor.

Figure 5.2



Depth shingle has 3 levels  
Strip more informative  
Plot symbol is a dot  
Isometric scales  
Reference grid

Figure 5.3

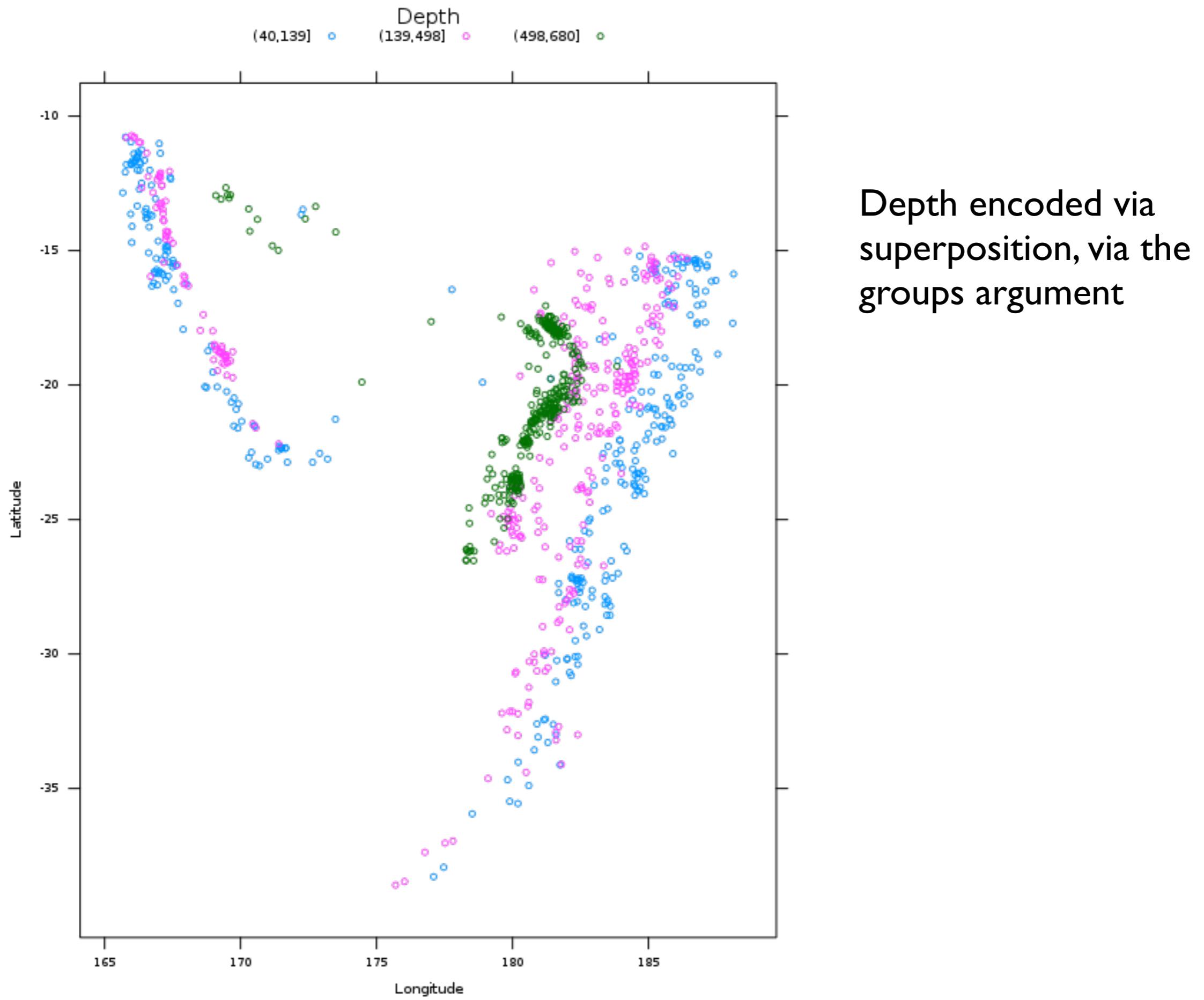
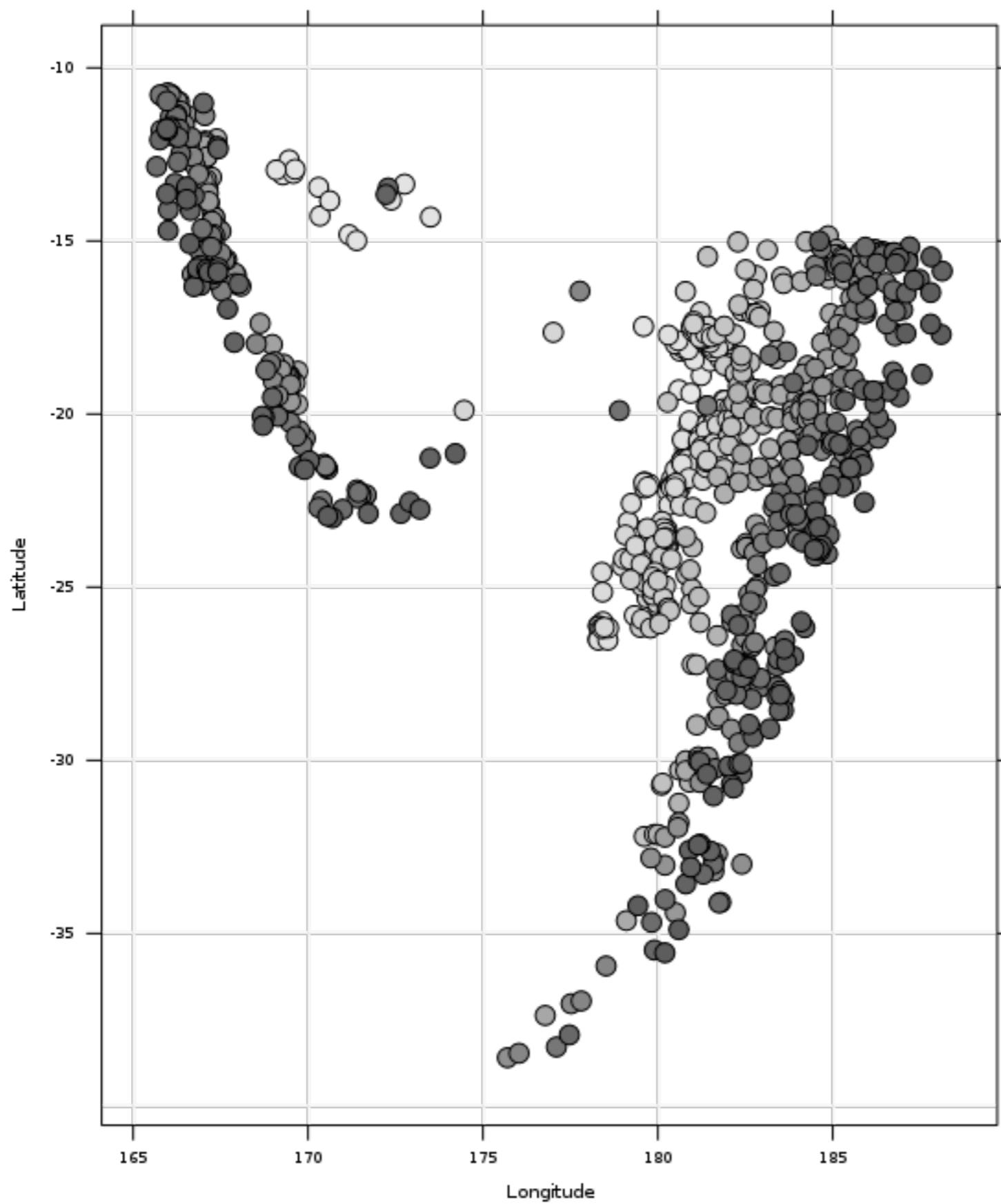
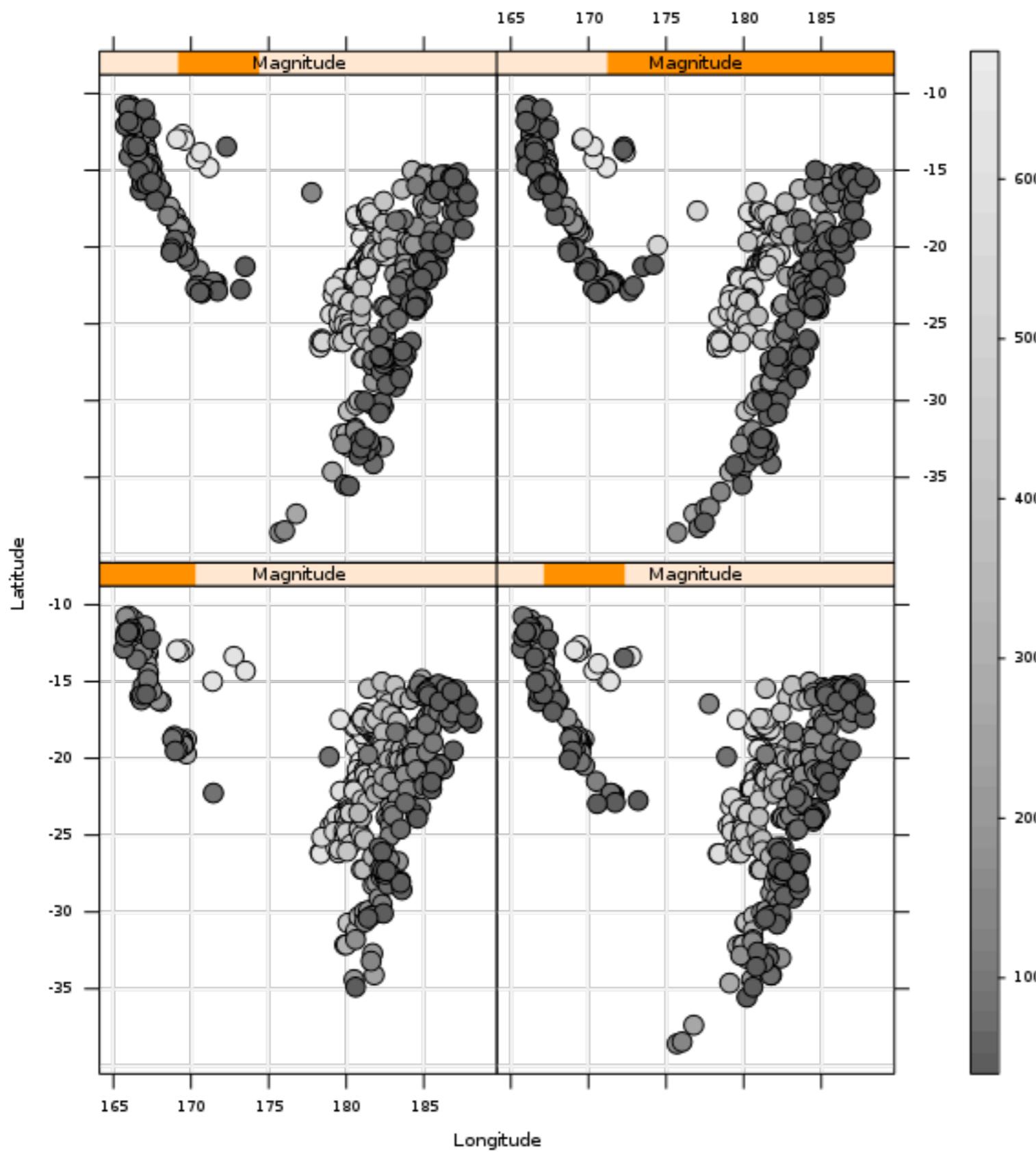


Figure 5.4



Depth encoded via color  
of shaded circle

Figure 5.6

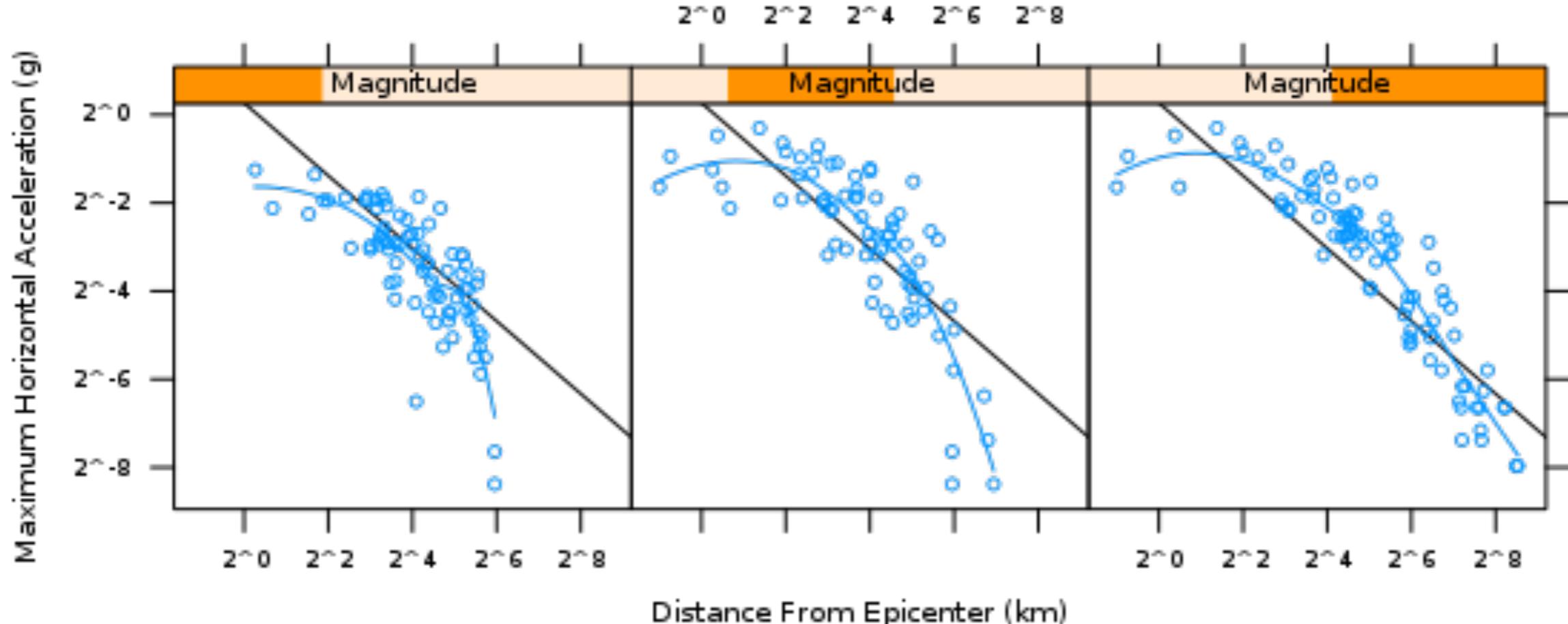


Depth encoded via color  
of shaded circle

AND

Magnitude treated as a  
shingle, i.e. the conditioning  
variable

Figure 5.10



```
library("locfit")
Earthquake$Magnitude <-
  equal.count(Earthquake$Richter, 3, overlap = 0.1)
coef <- coef(lm(log2(accel) ~ log2(distance), data = Earthquake))

## Figure 5.10
xyplot(accel ~ distance | Magnitude, data = Earthquake,
       scales = list(log = 2), col.line = "grey", lwd = 2,
       panel = function(...) {
         panel.abline(reg = coef)
         panel.locfit(...)
       },
       xlab = "Distance From Epicenter (km)",
       ylab = "Maximum Horizontal Acceleration (g)")
```

Magnitude used as a shingle  
w/ overlapping intervals.

Common regression line  
added as a visual reference

Alternative scatterplot  
smoother from locfit  
package.

# Lessons from `xyplot()` tour

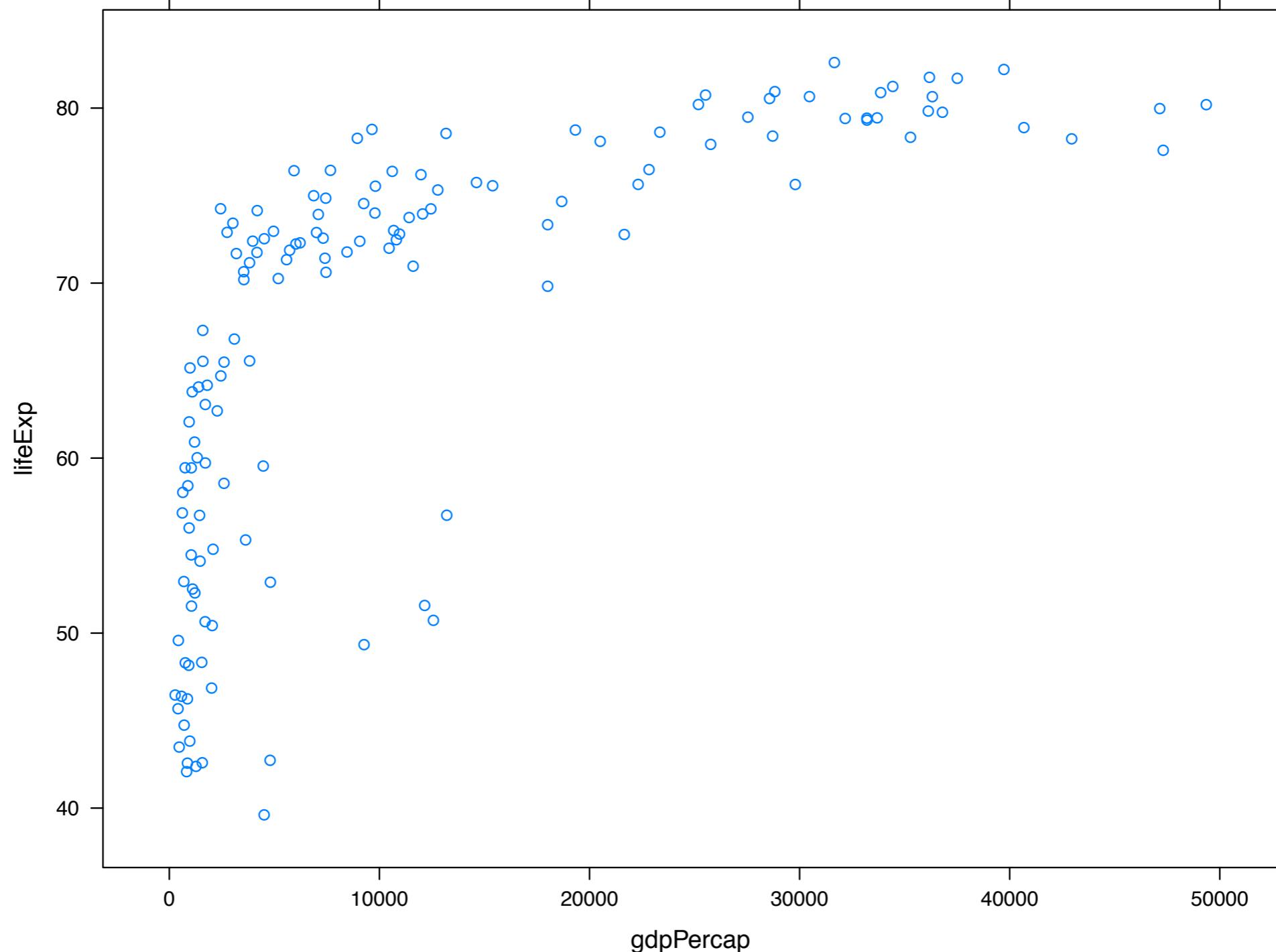
- Scatterplots are conceptually simple yet powerful.
- `xyplot()` produces nice scatterplots, even basic ones.
- Reference lines and, where appropriate, other common elements facilitate data interpretation.
- Various transformations and regression approaches can be enacted ‘on the fly’ by `xyplot`, such as the log transform and loess smoothing.
- The `type` argument is very useful for the above.

# Lessons from `xyplot()` tour (cont'd)

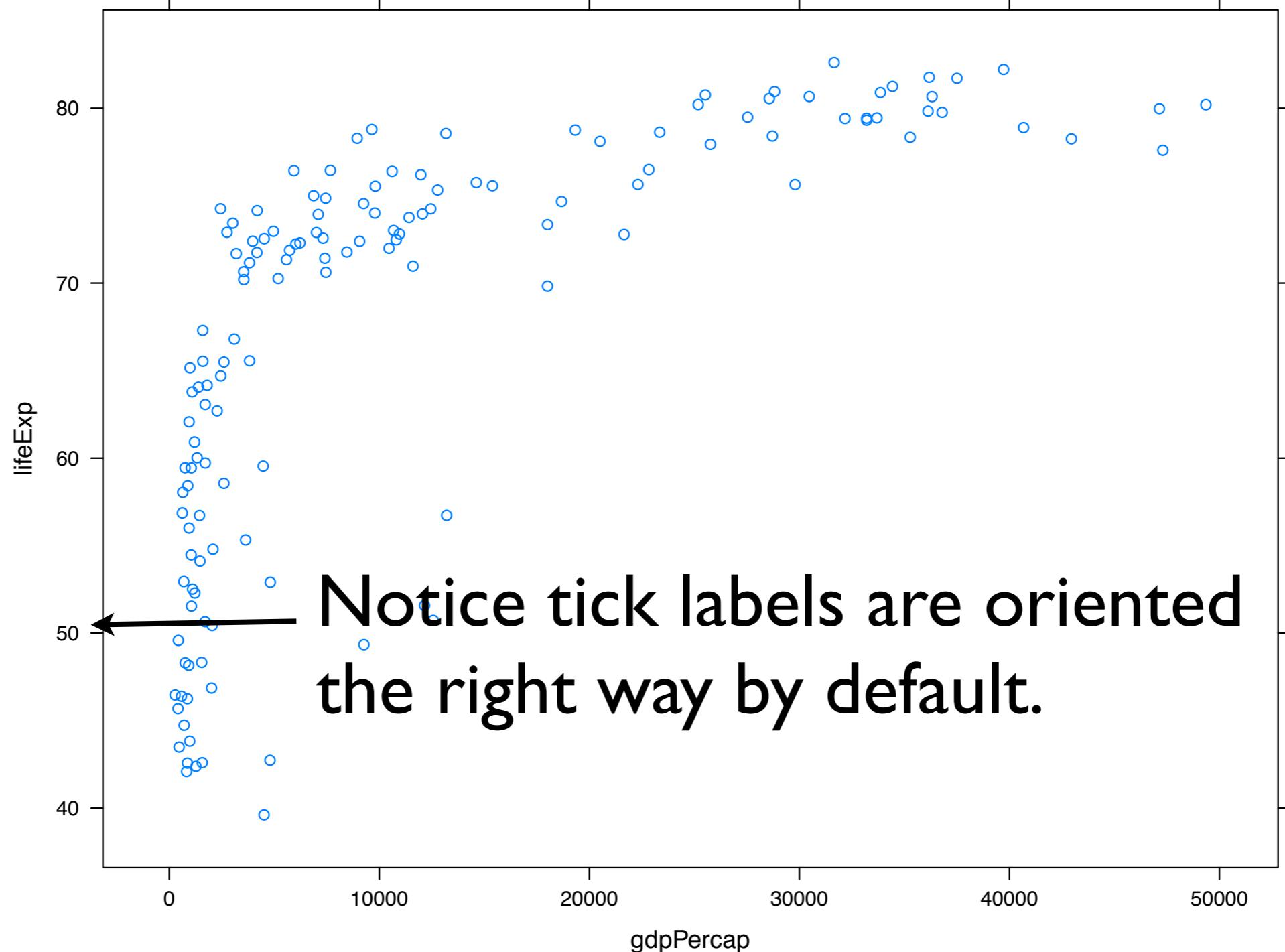
- The relationship between X and Y can be explored simultaneously for different values of a third variable via
  - conditioning into different panels
  - superposition with visual cues for the groups
- Consider treating an ancillary quantitative variable as a shingle.
- It is possible to depict four quantitative variables at once in a scatterplot. X and Y cover the two primary variables, the third can be conveyed (crudely) via a shingle, and the fourth encoded by color and/or size of plotting symbol.

# step by step development of the Gapminder figure using lattice graphics

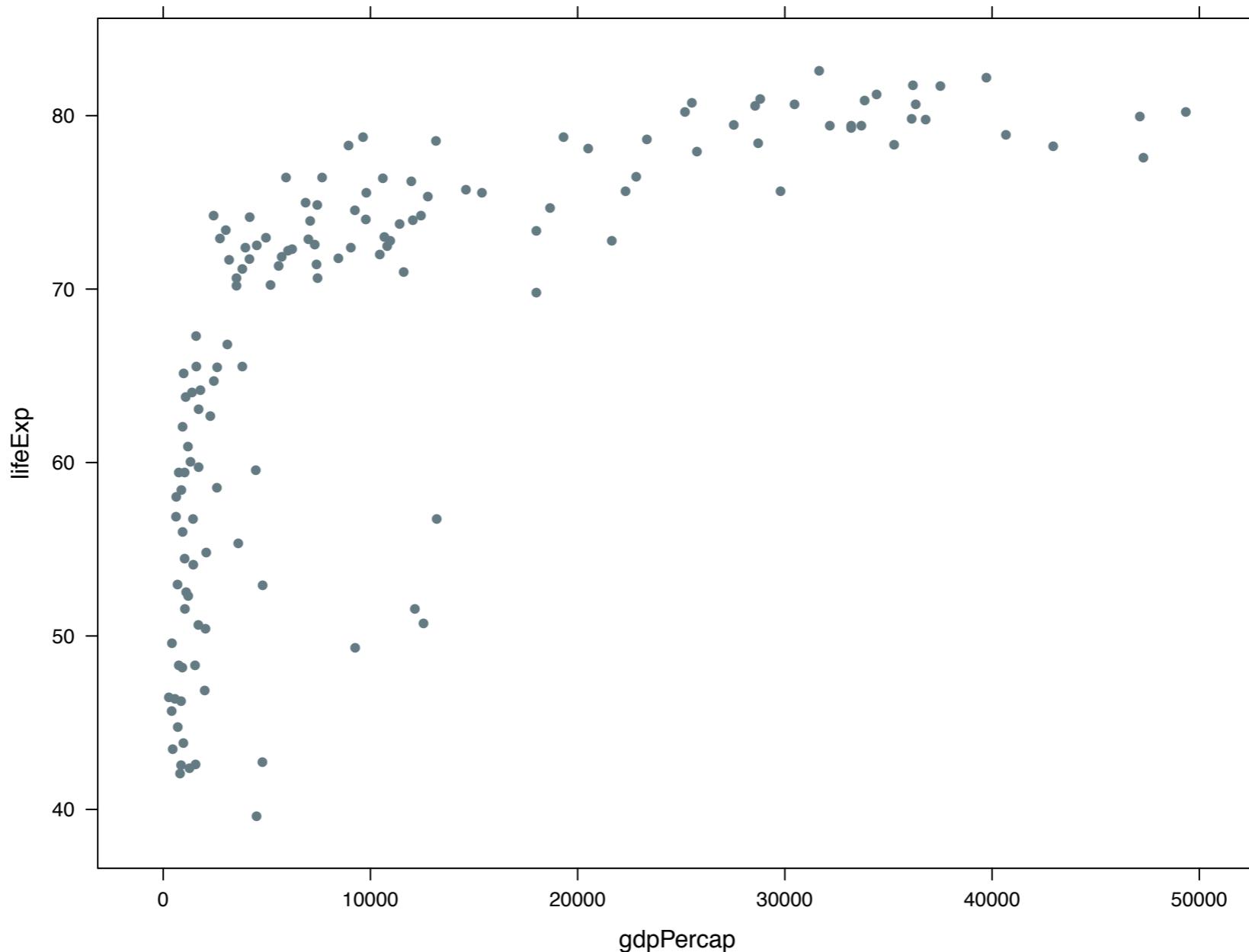
Note: most figures now here in two versions. One made under a default lattice theme and the other made under JB's lattice theme. Once the customizations gets profound enough, the difference goes away ....



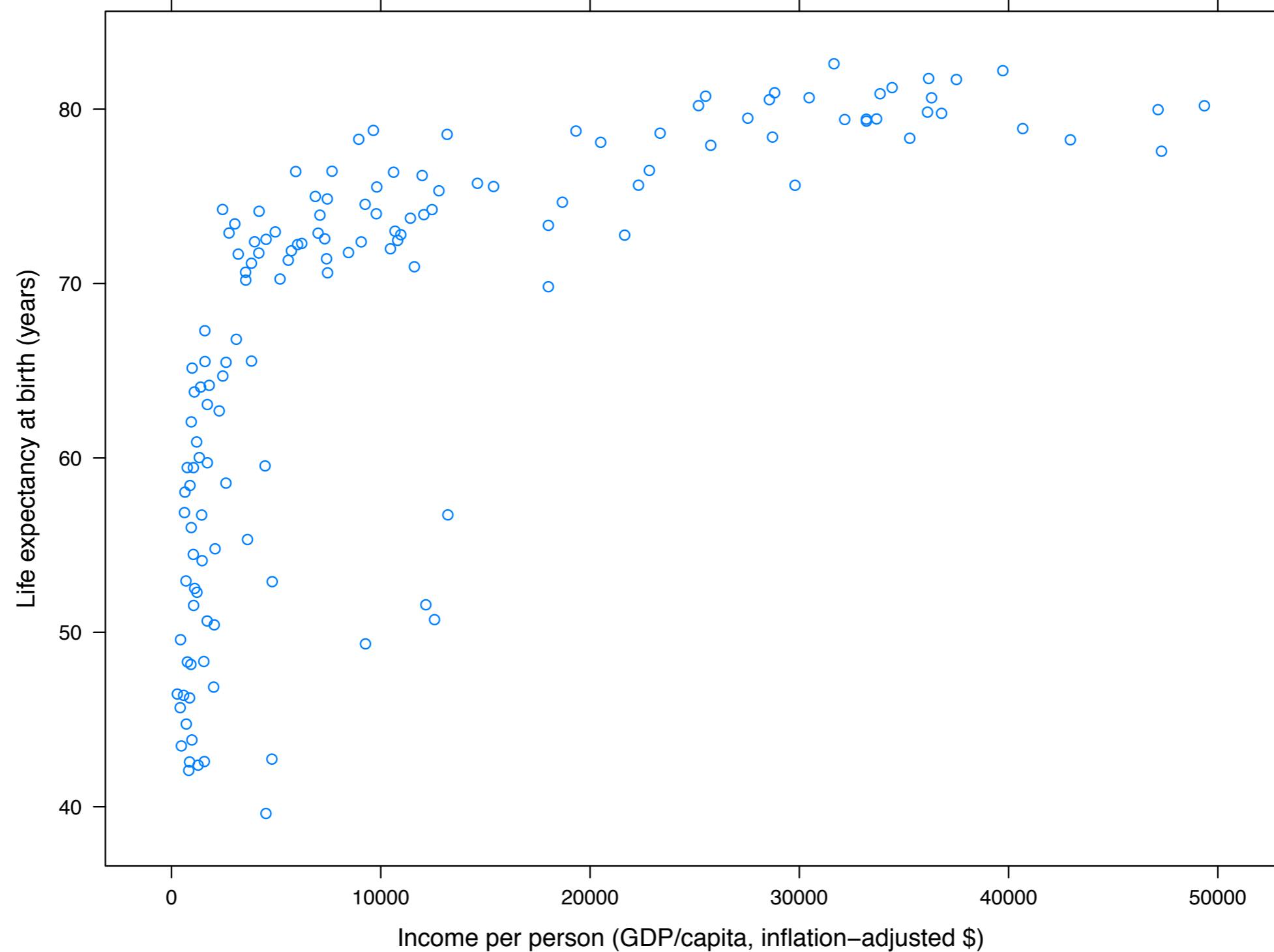
```
xyplot(lifeExp ~ gdpPercap, gDatOrdered, subset = year == jYear)
```



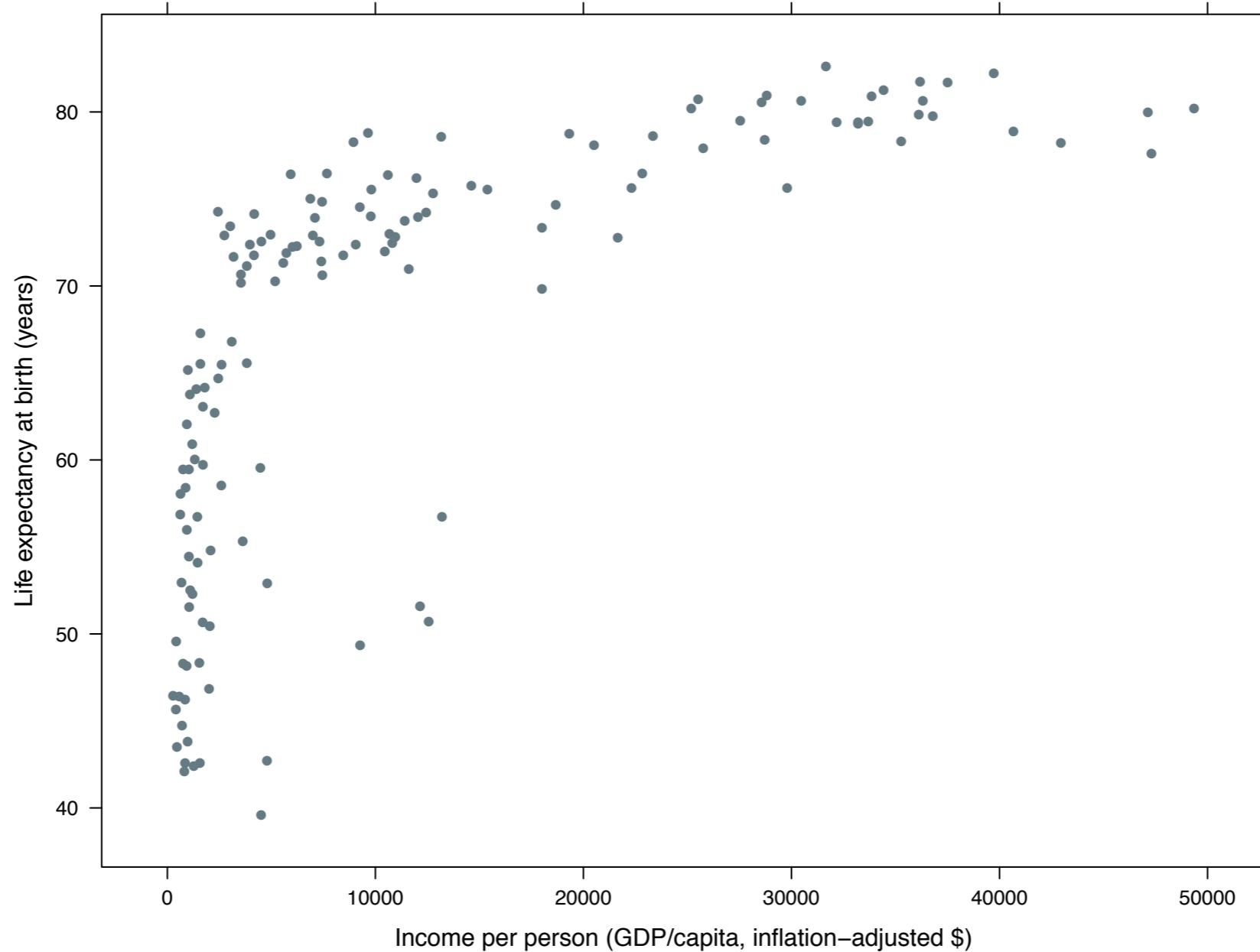
```
xyplot(lifeExp ~ gdpPercap, gDatOrdered, subset = year == jYear)
```



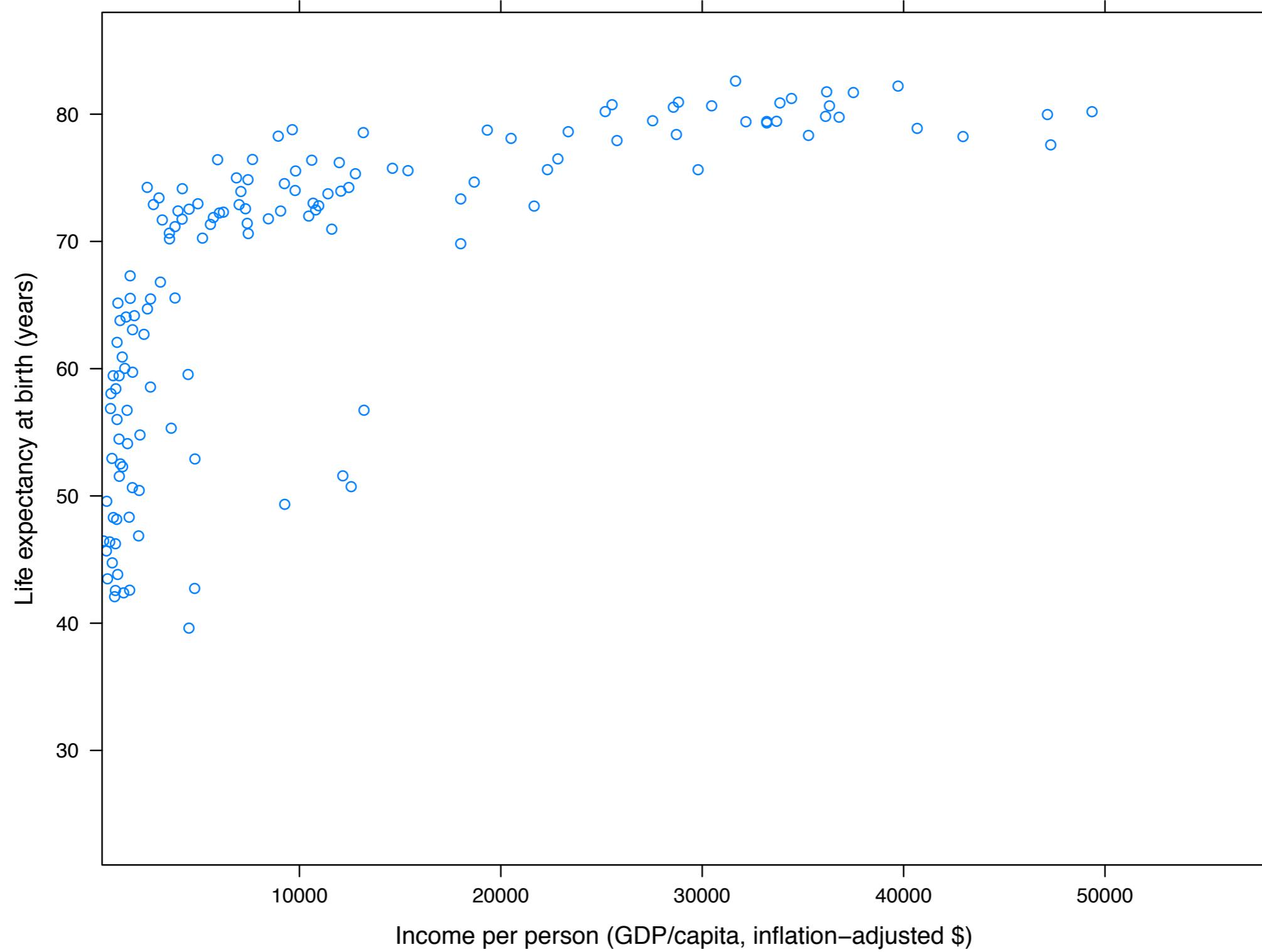
```
xyplot(lifeExp ~ gdpPerCap, gDatOrdered, subset = year == jYear)
```



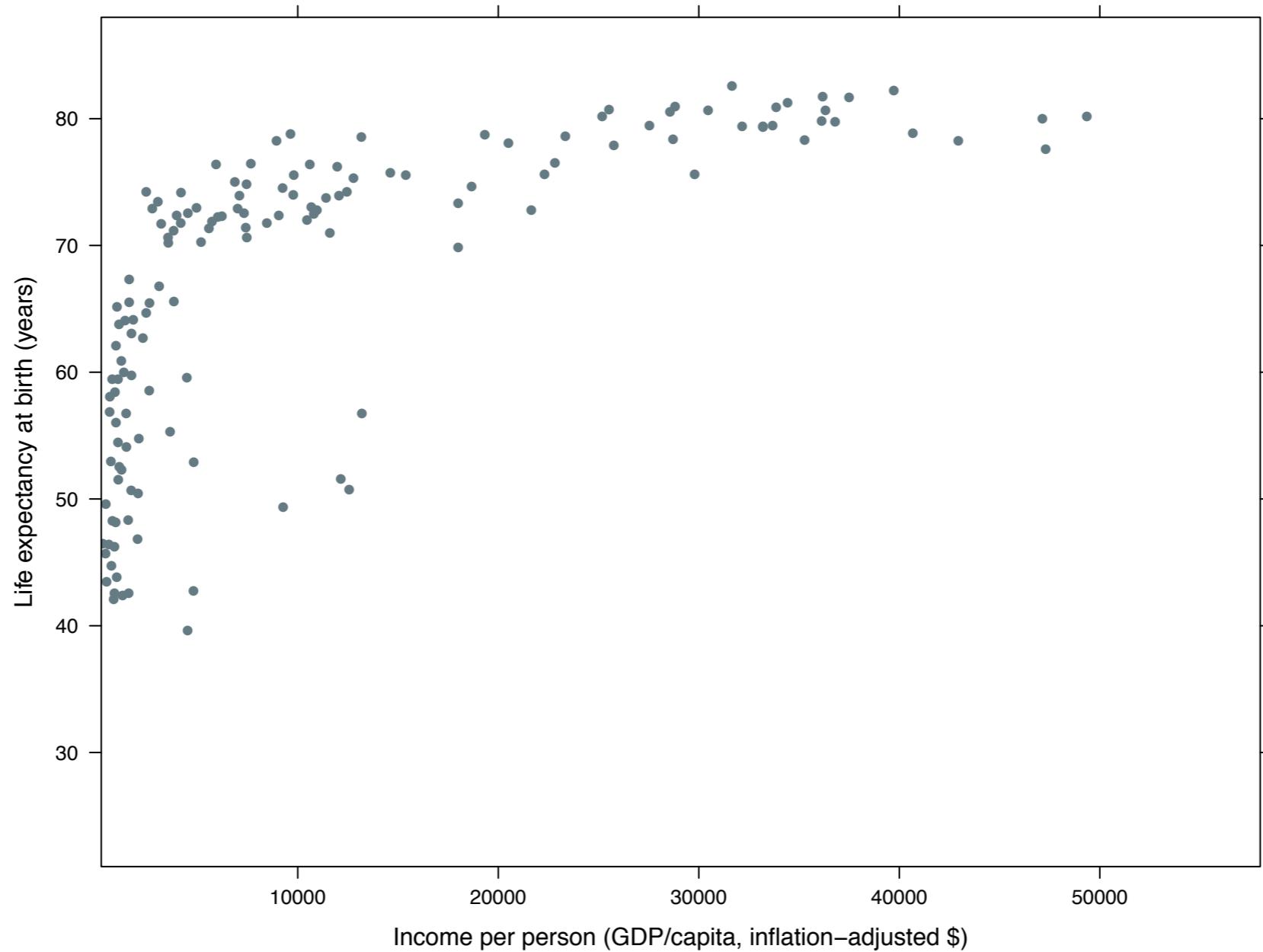
```
jXlab <- "Income per person (GDP/capita, inflation-adjusted $)"  
jYlab <- "Life expectancy at birth (years)"  
xyplot(lifeExp ~ gdpPercap, gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab)
```



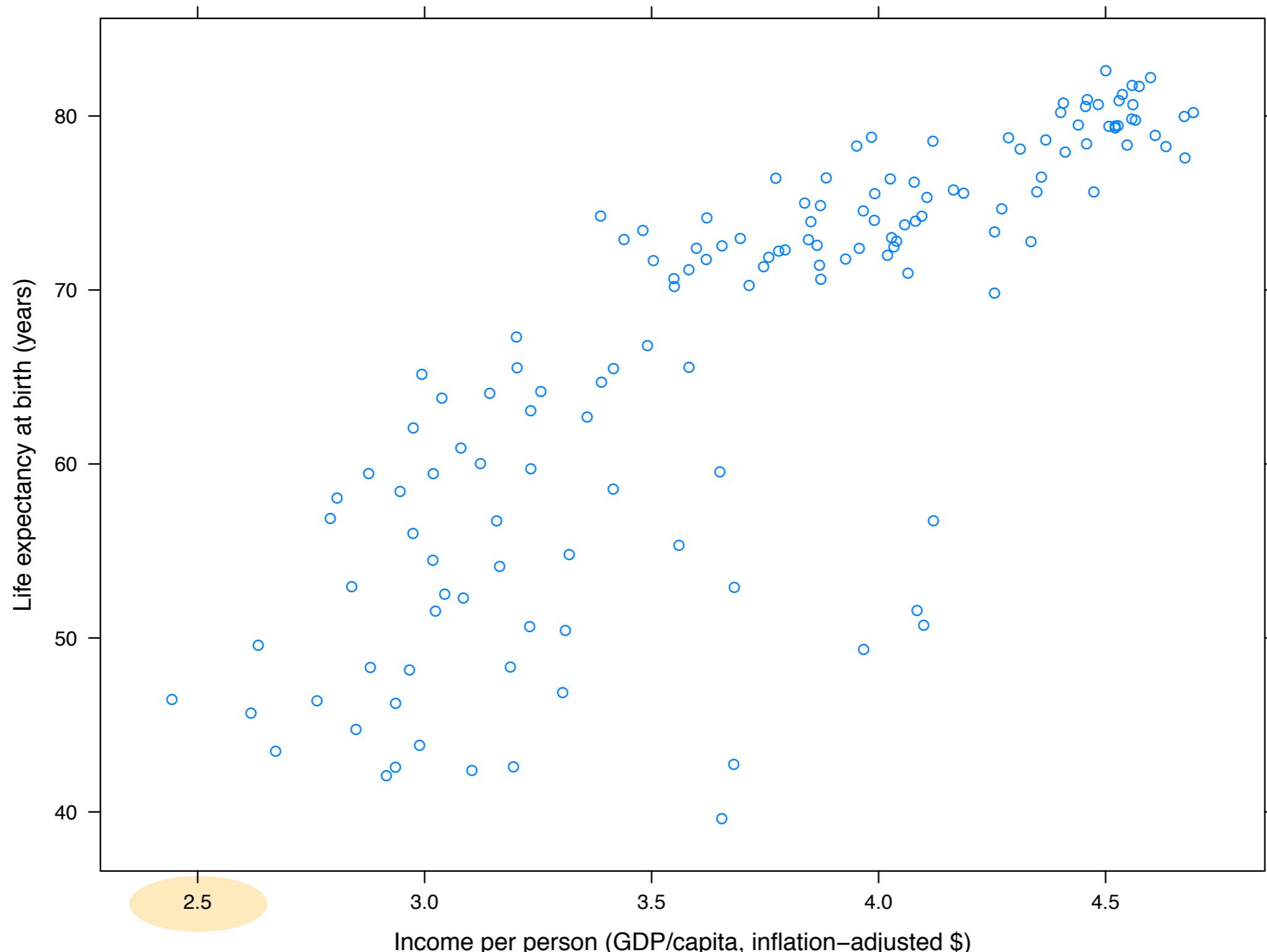
```
jXlab <- "Income per person (GDP/capita, inflation-adjusted $)"  
jYlab <- "Life expectancy at birth (years)"  
xyplot(lifeExp ~ gdpPercap, gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab)
```



```
jXlim <- c(200, 58000)
jYlim <- c(21, 88)
xyplot(lifeExp ~ gdpPerCap, gDatOrdered, subset = year == jYear,
       xlab = jXlab, ylab = jYlab,
       xlim = jXlim, ylim = jYlim)
```

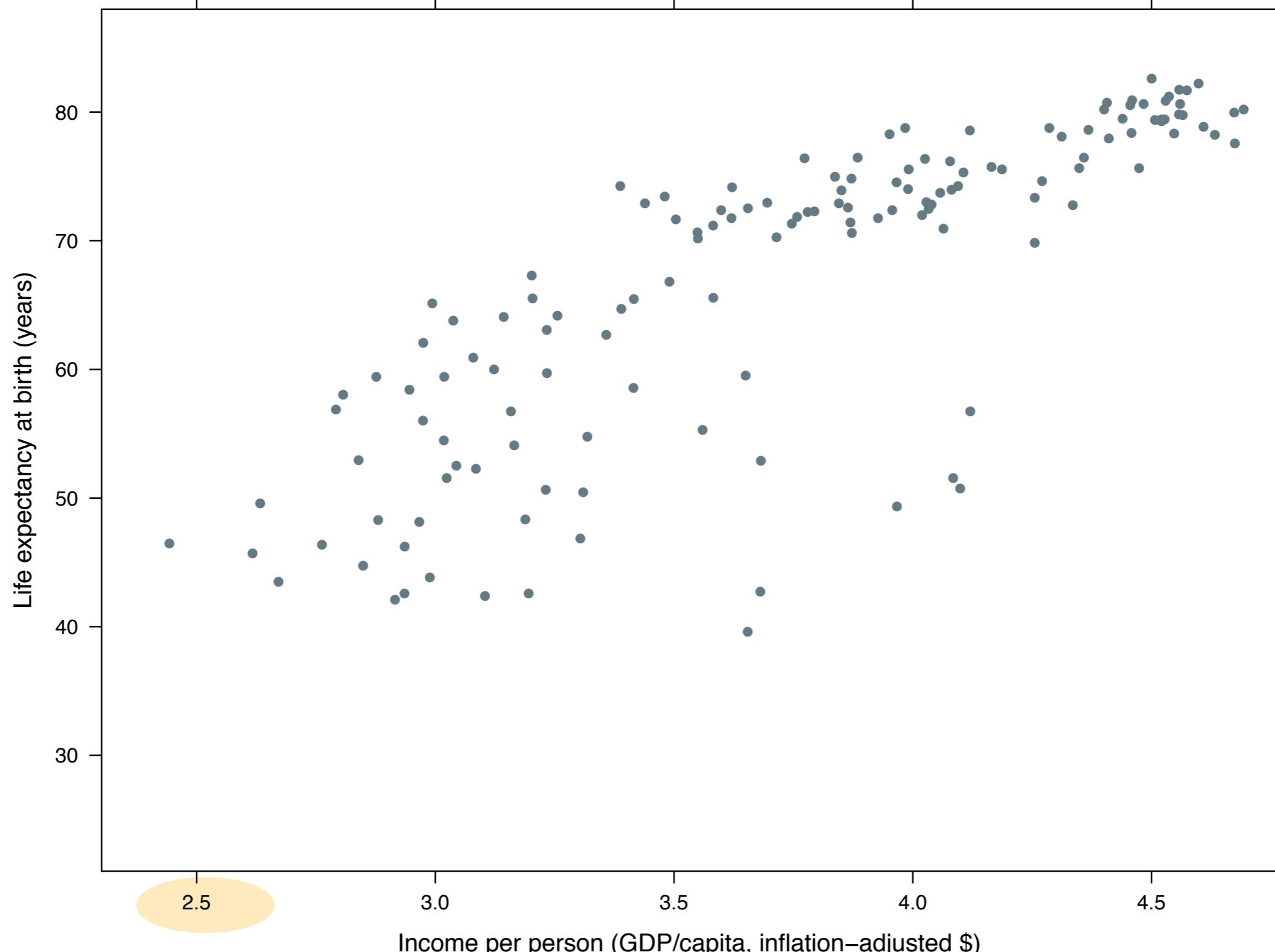


```
jXlim <- c(200, 58000)
jYlim <- c(21, 88)
xyplot(lifeExp ~ gdpPerCap, gDatOrdered, subset = year == jYear,
       xlab = jXlab, ylab = jYlab,
       xlim = jXlim, ylim = jYlim)
```



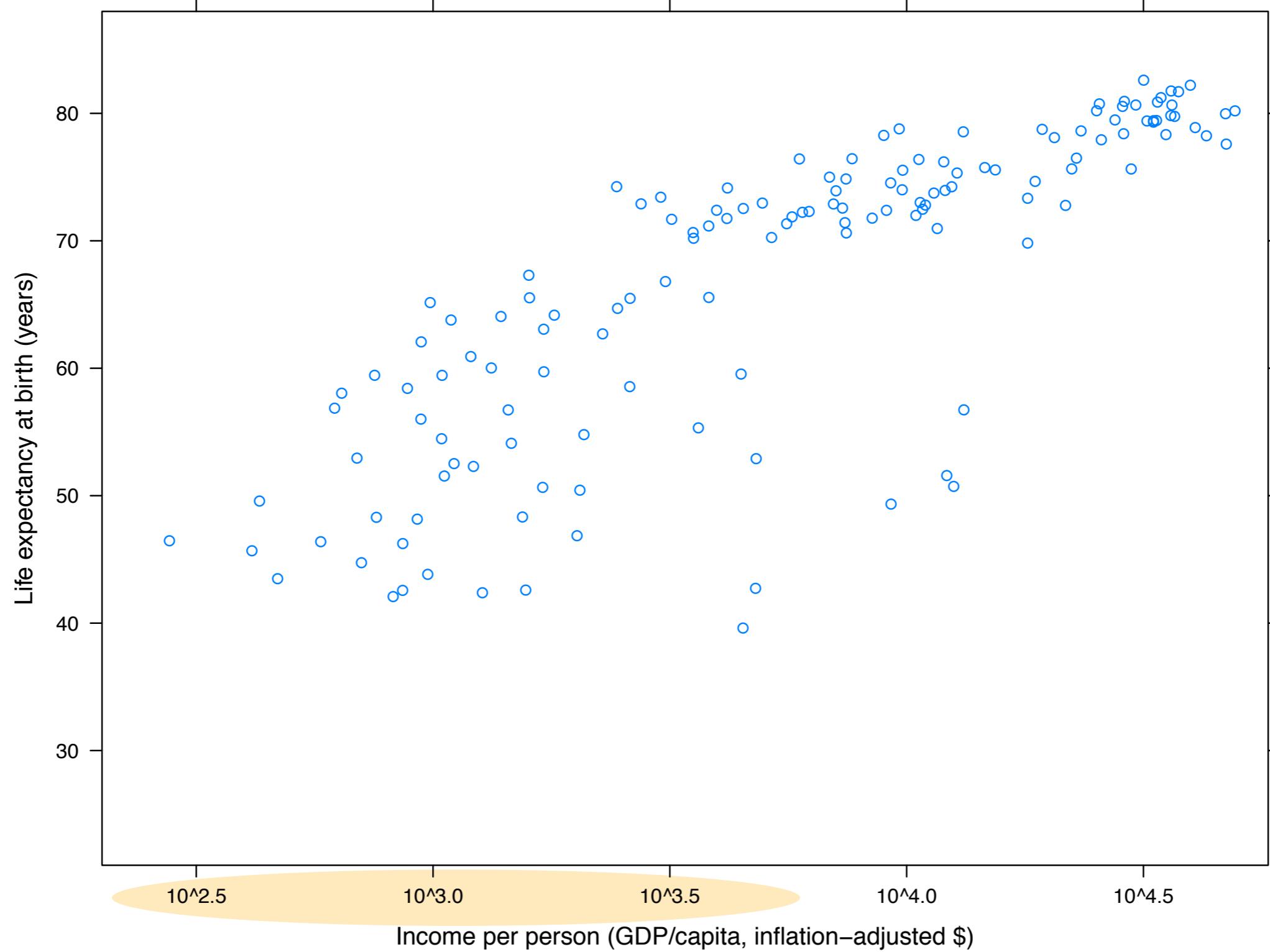
```
xyplot(lifeExp ~ log10(gdpPerCap), gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab,  
      xlim = log10(jXlim), ylim = jYlim))
```

It's still a bad idea to log 'by hand'.



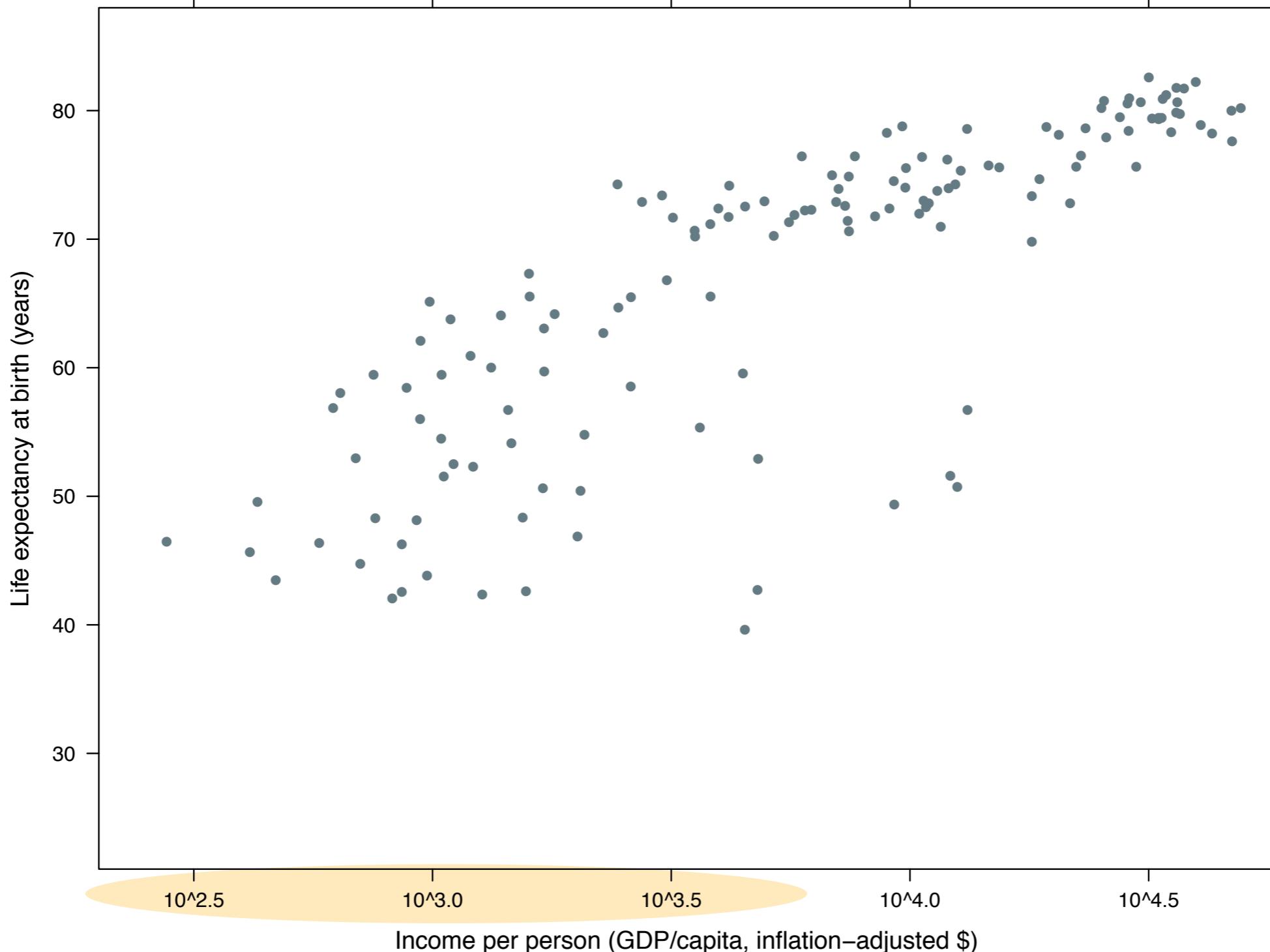
```
xyplot(lifeExp ~ log10(gdpPerCap), gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab,  
      xlim = log10(jXlim), ylim = jYlim))
```

It's still a bad idea to log 'by hand'.



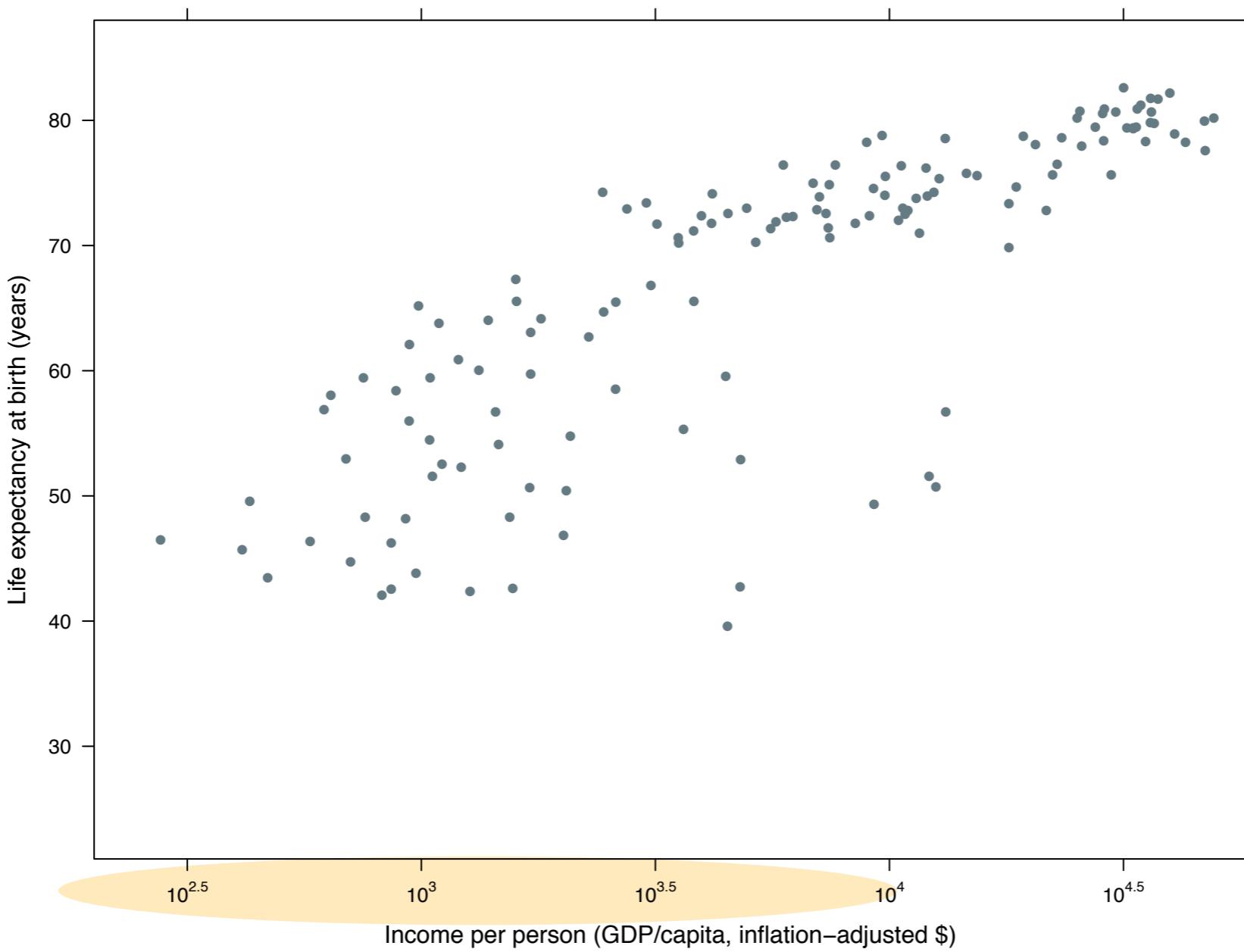
```
xyplot(lifeExp ~ gdpPerCap, gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab,  
      xlim = jXlim, ylim = jYlim,  
      scales = list(x = list(log = 10)))
```

**Use the scales argument to log transform an axis  
(among many many other things).**



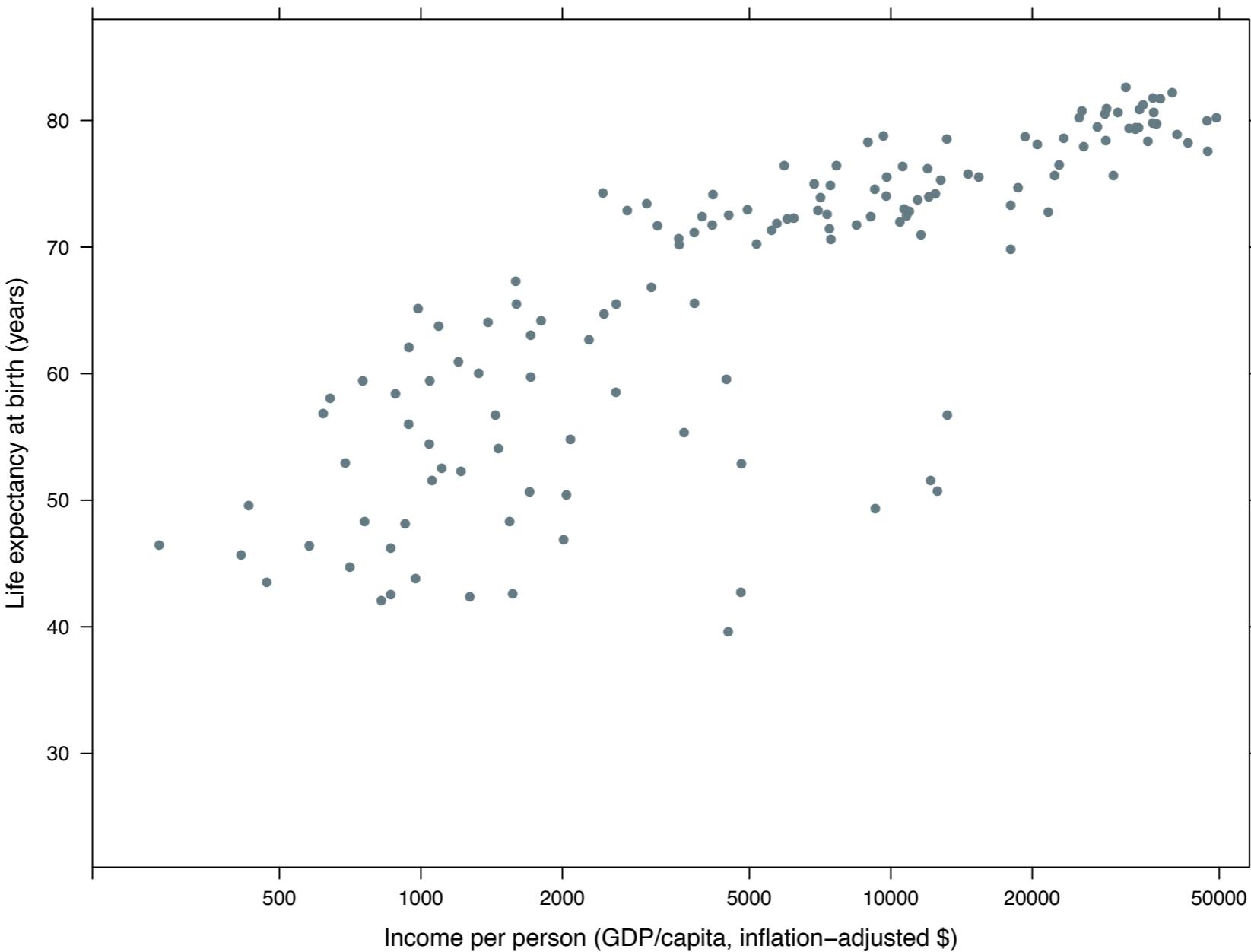
```
xyplot(lifeExp ~ gdpPerCap, gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab,  
      xlim = jXlim, ylim = jYlim,  
      scales = list(x = list(log = 10)))
```

Use the `scales` argument to log transform an axis  
(among many many other things).



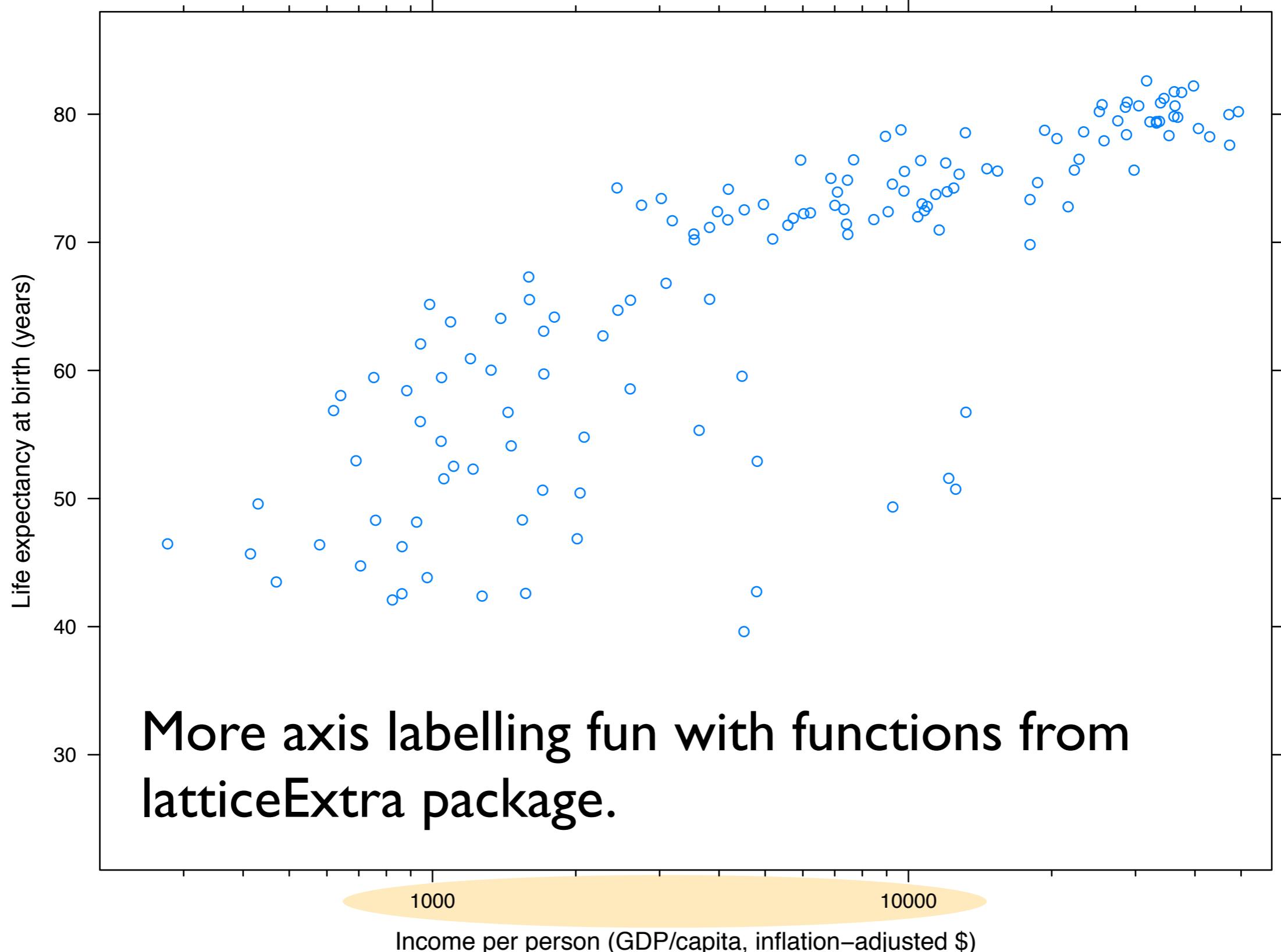
```
xyplot(lifeExp ~ gdpPercap, gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab,  
      scales = list(x = list(log = 10)),  
      xlim = jXlim, ylim = jYlim,  
      xscale.components = xscale.components.logpower)
```

`latticeExtra` package offers facilities for better log axis labelling; here an improved axis drawing function, `xscale.components.logpower`.

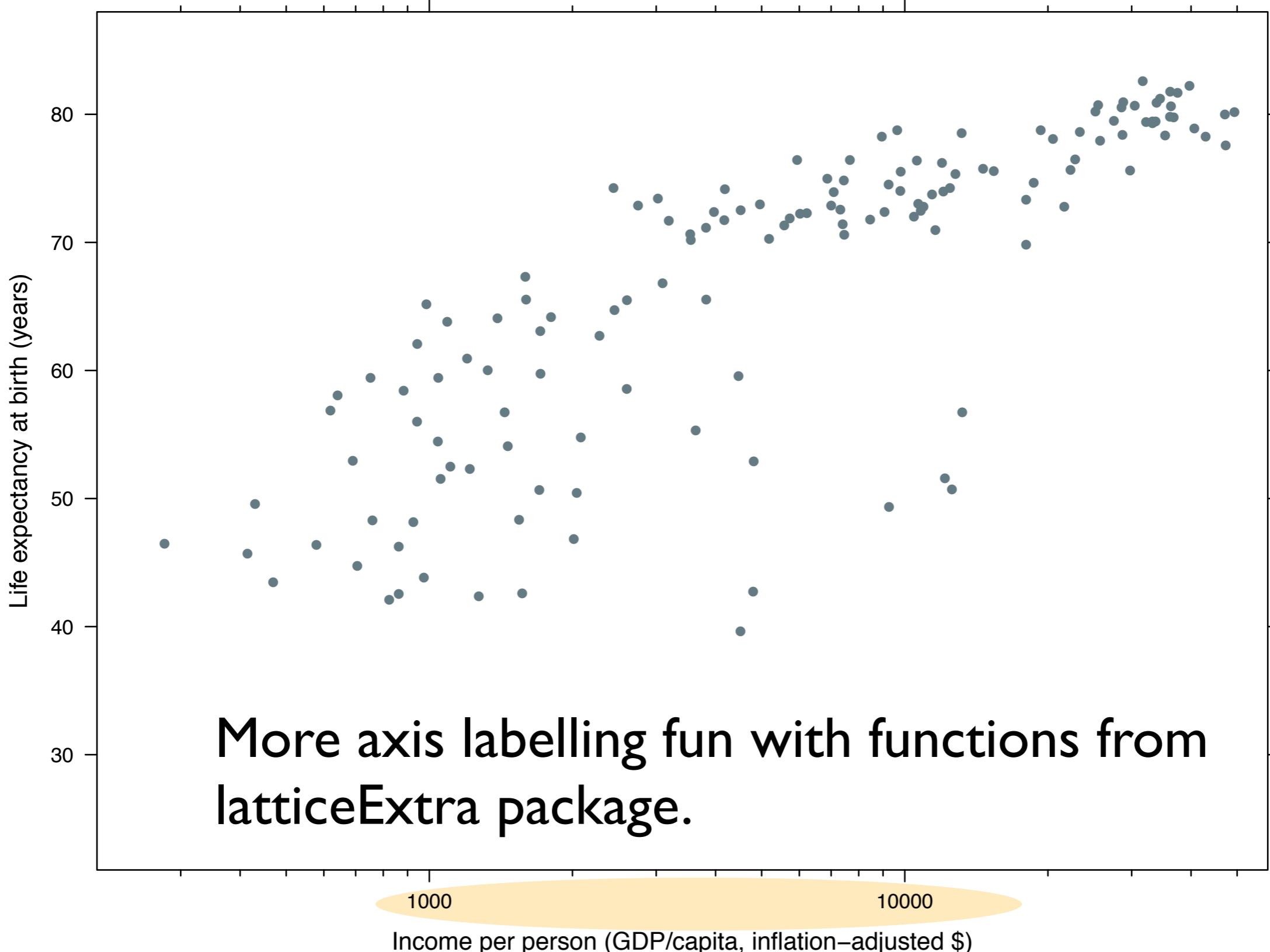


```
xyplot(lifeExp ~ gdpPercap, gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab,  
      xlim = jXlim, ylim = jYlim,  
      scales = list(x = list(log = 10), equispaced.log = FALSE))
```

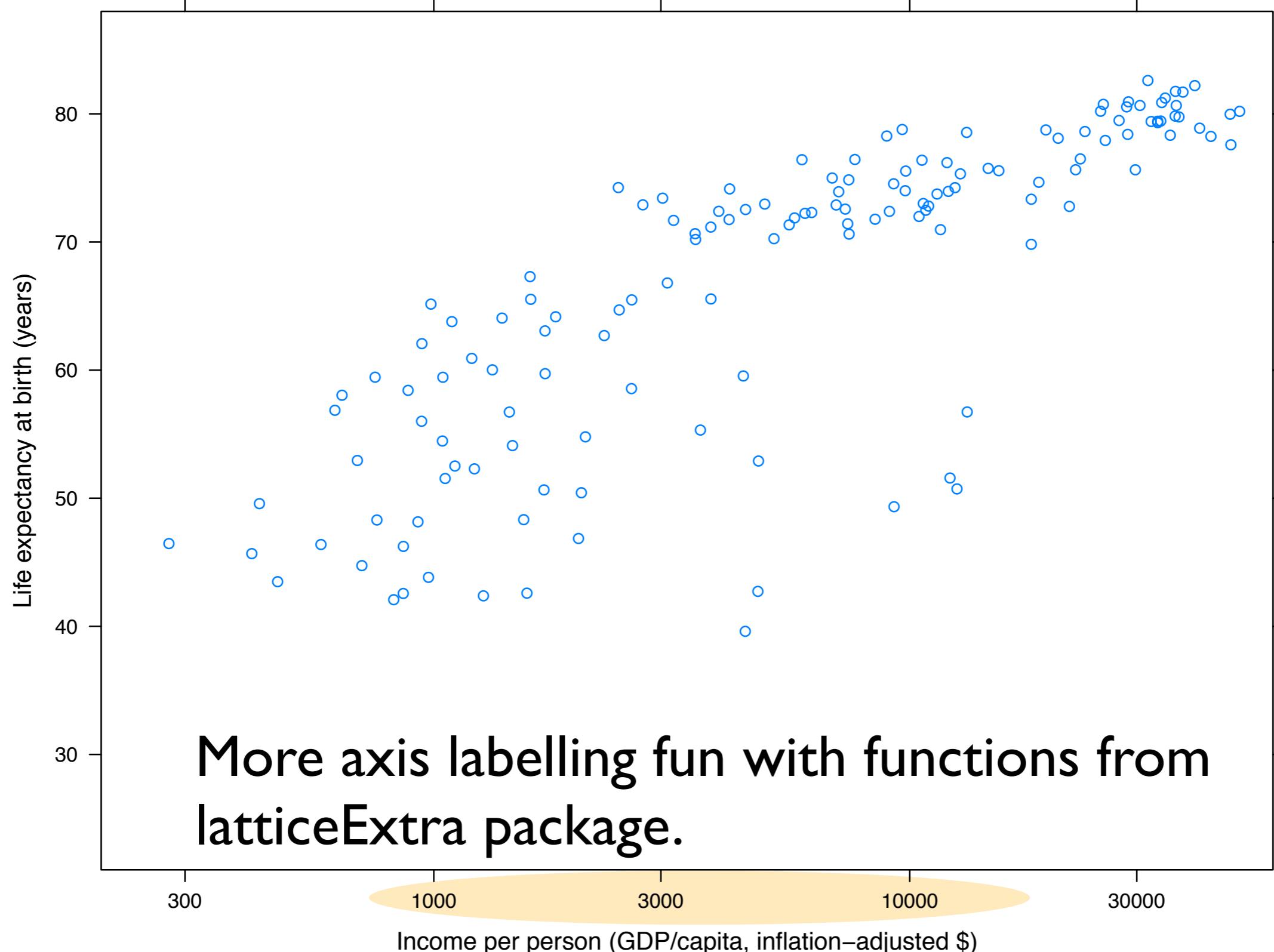
New argument for scales emulates log axis tick marking in traditional graphics.



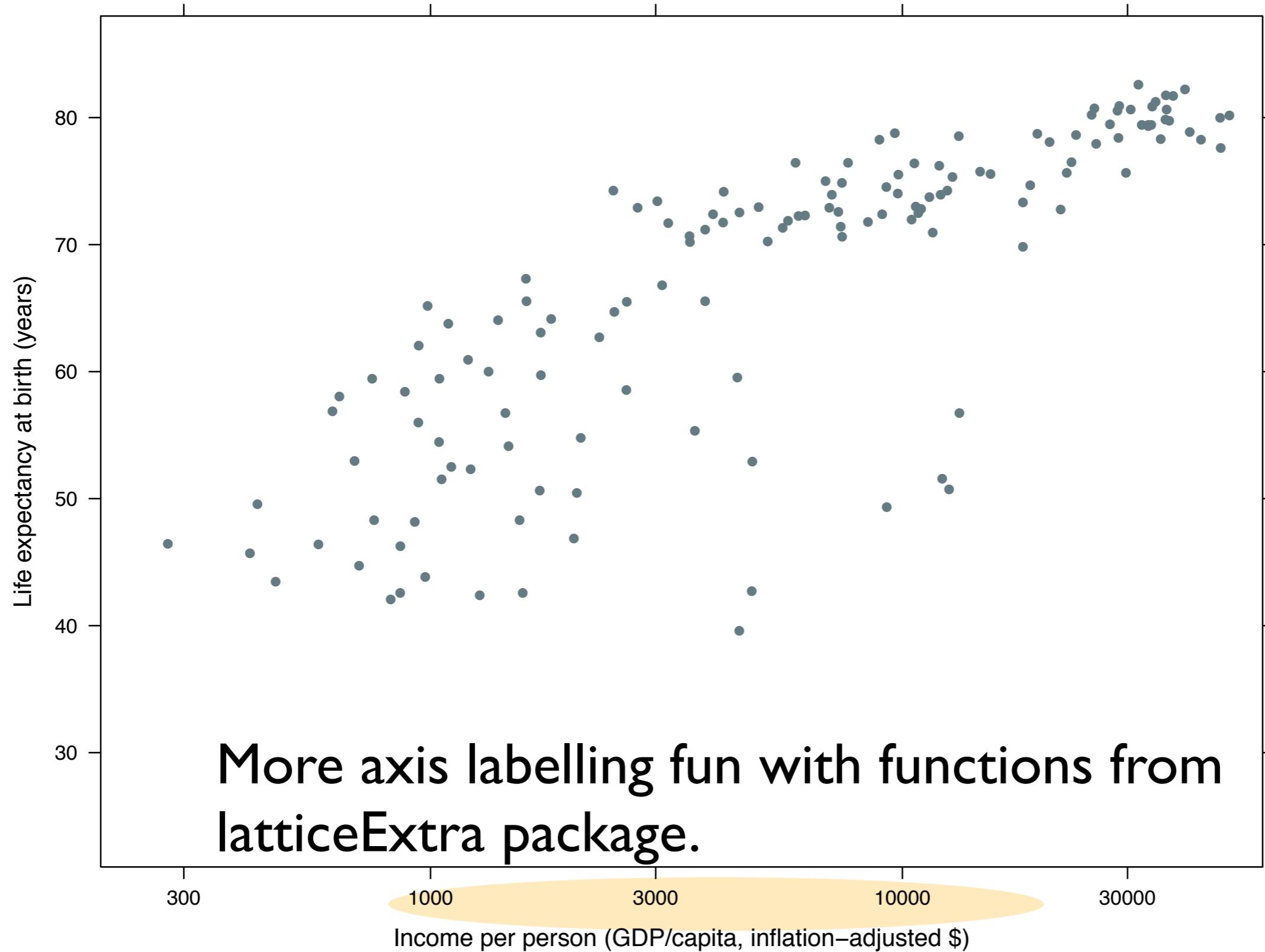
```
xyplot(lifeExp ~ gdpPercap, <snip, snip>,  
xscale.components = xscale.components.log10ticks)
```



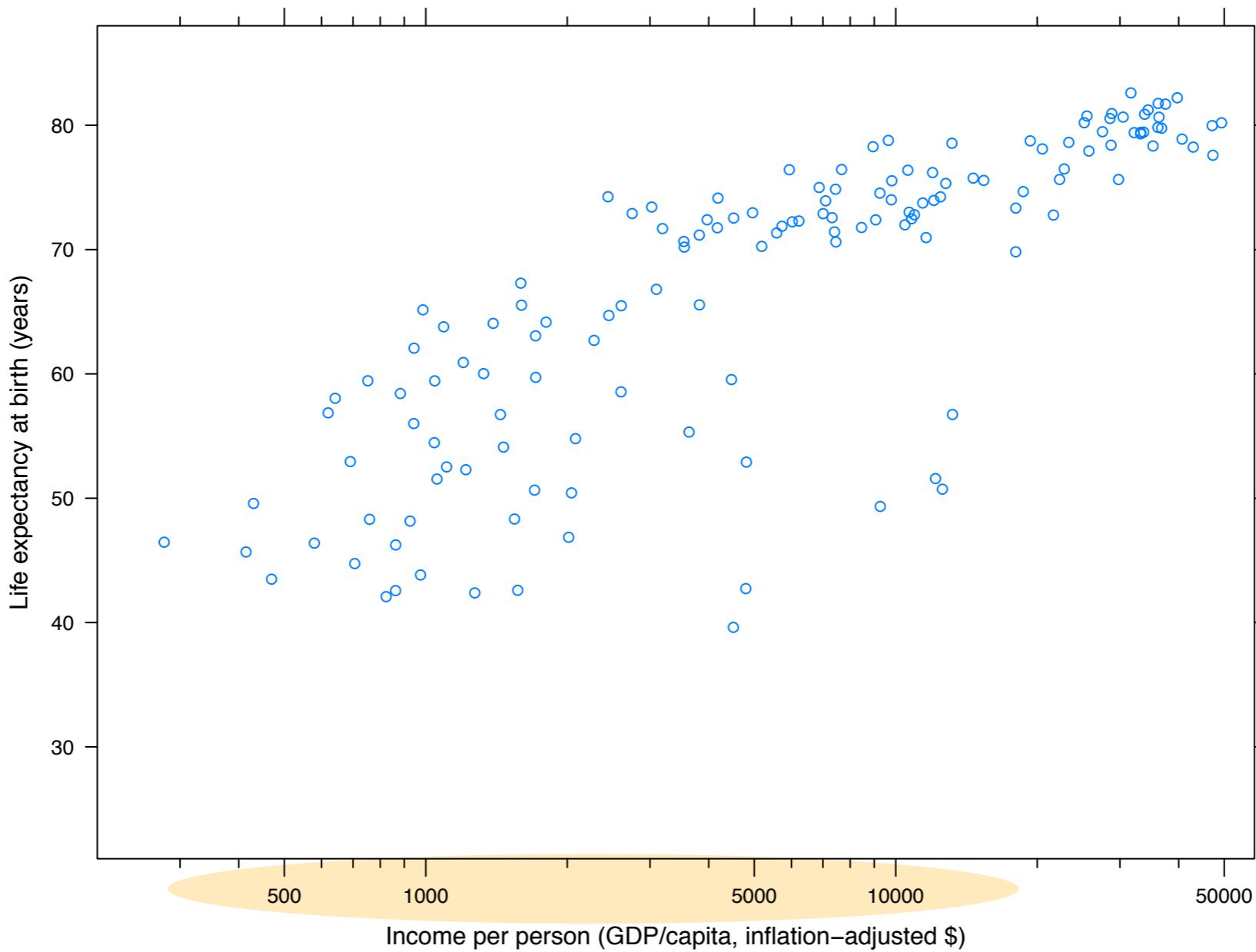
```
xyplot(lifeExp ~ gdpPerCap, <snip, snip>,  
      xscale.components = xscale.components.log10ticks)
```



```
xyplot(lifeExp ~ gdpPerCap, <snip, snip>,  
      xscale.components = xscale.components.log10.3)
```

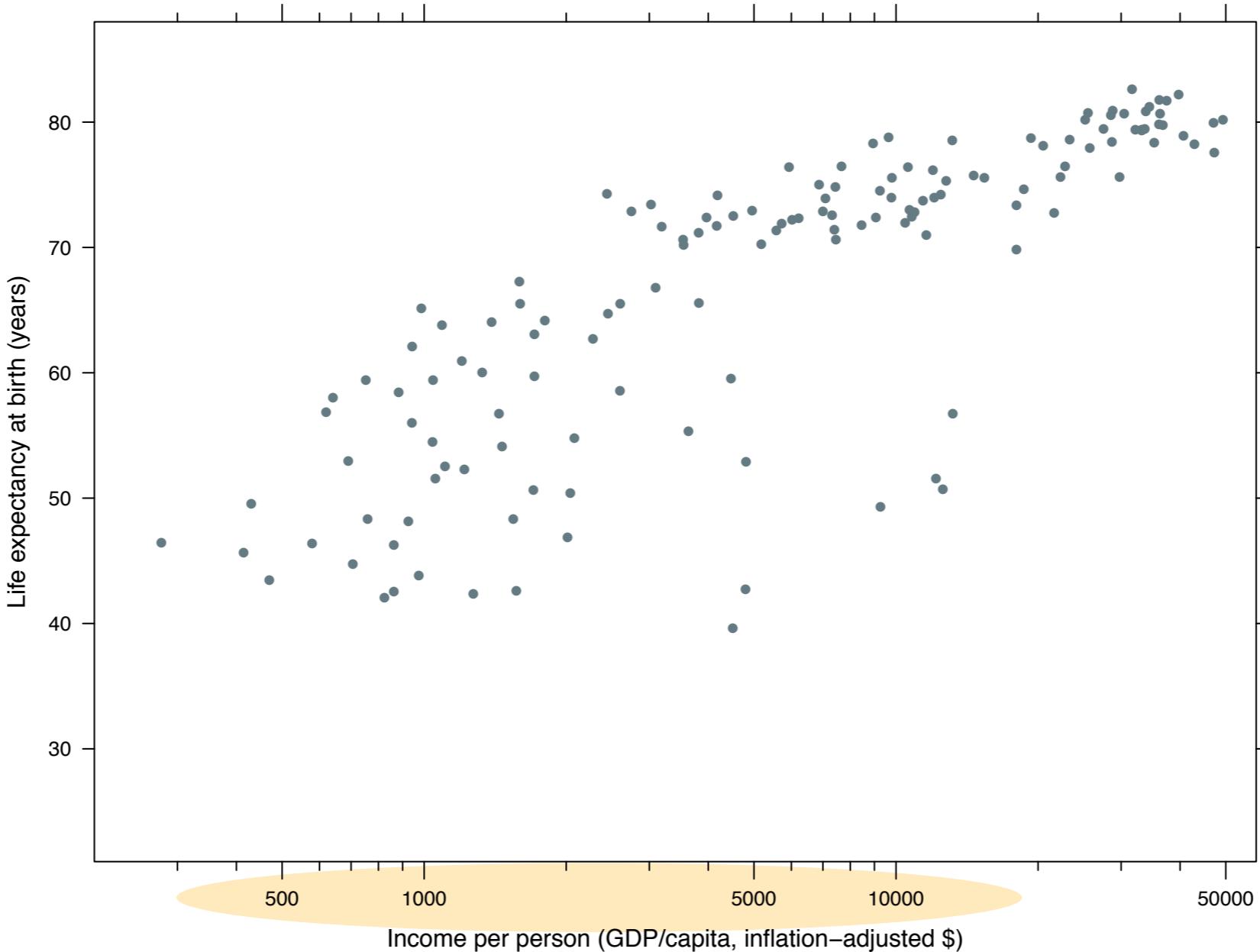


```
xyplot(lifeExp ~ gdpPerCap, <snip, snip>,  
      xscale.components = xscale.components.log10.3)
```



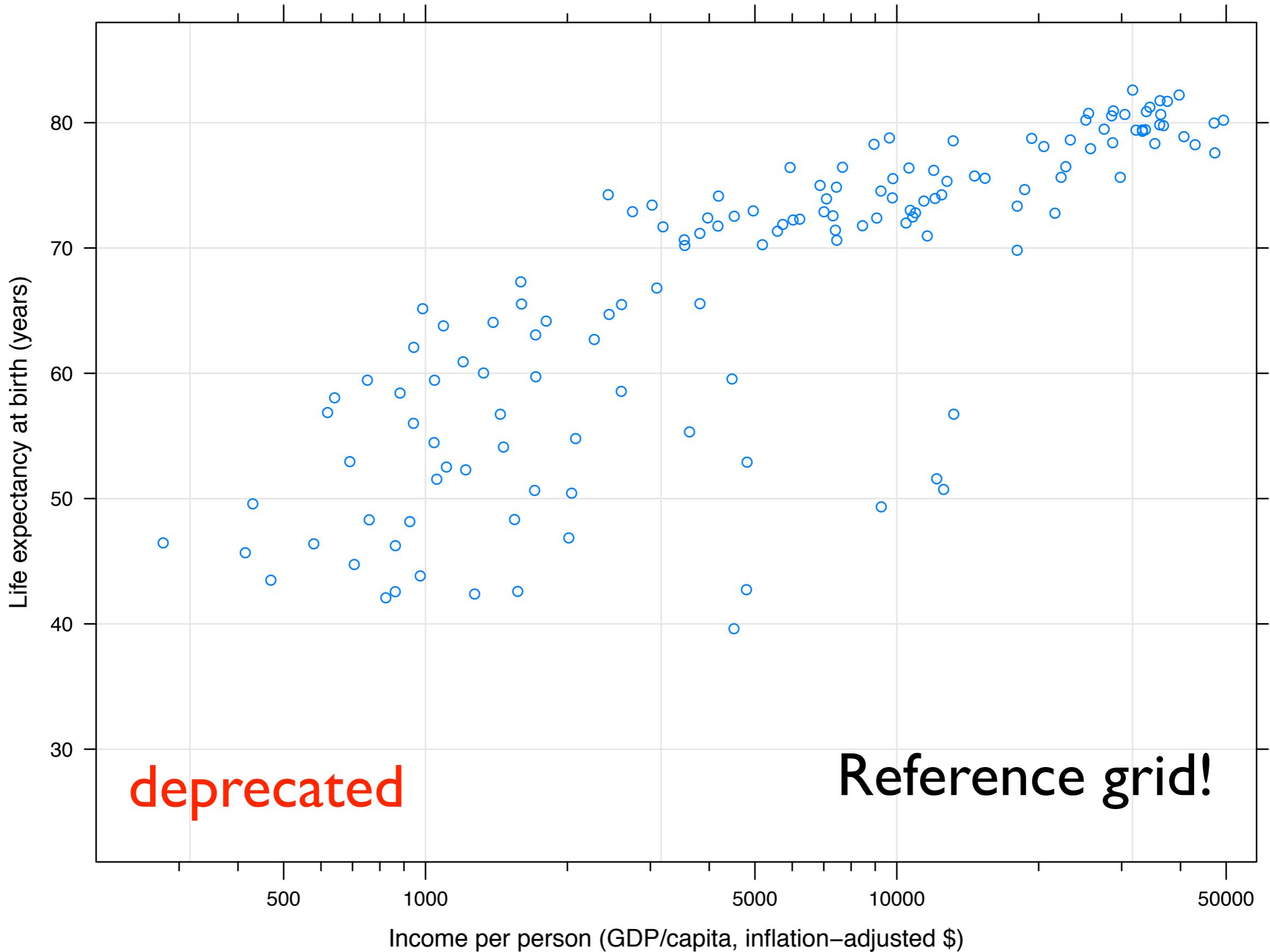
```
xyplot(lifeExp ~ gdpPercap, gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab,  
      scales = list(x = list(log = 10)),  
      xlim = jXlim, ylim = jYlim,  
      xscale.components = xscale.components.log10)
```

For even more control, can write a custom function for this. Kind of advanced / obscure. See the code.

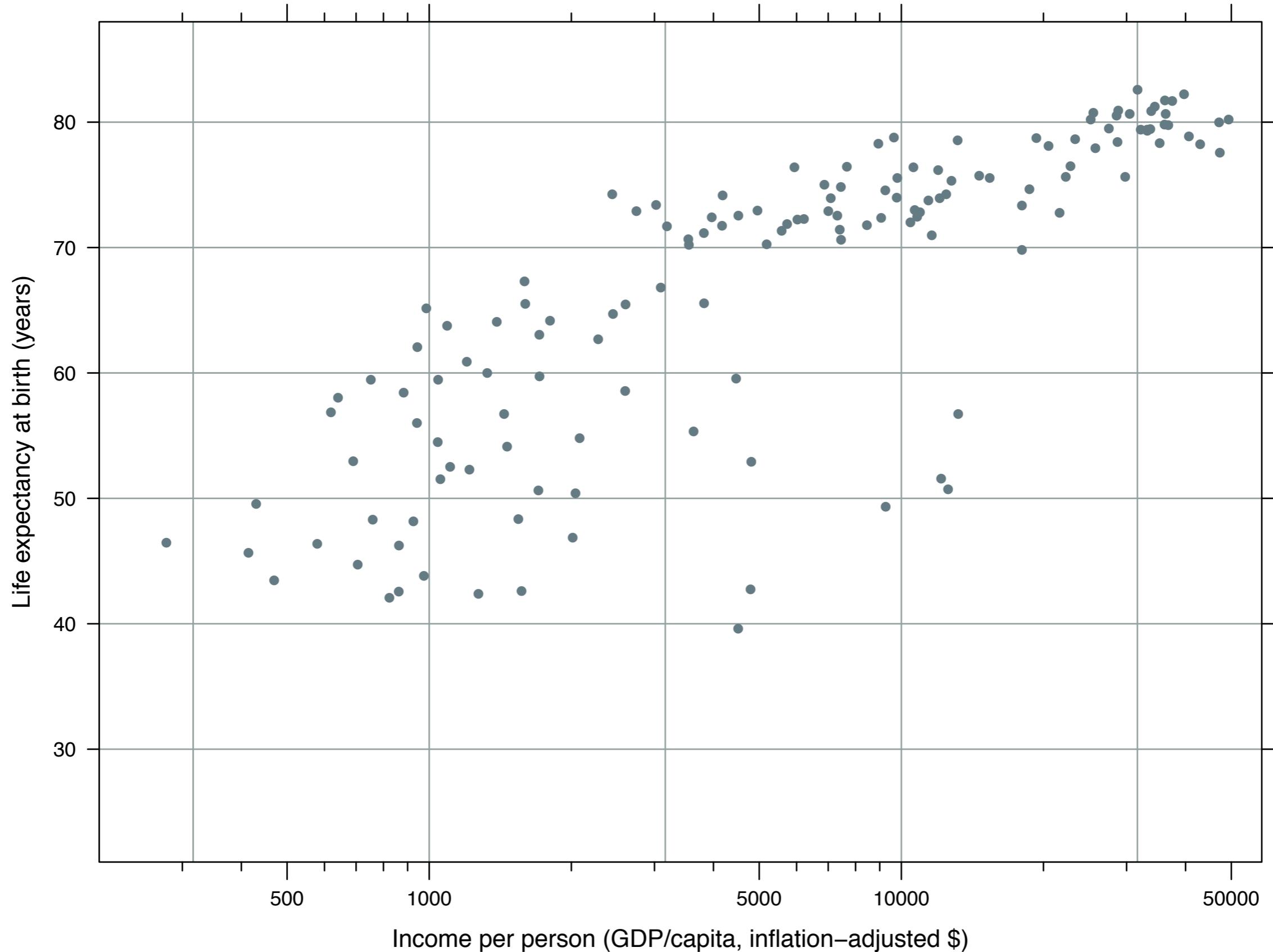


```
xyplot(lifeExp ~ gdpPercap, gDatOrdered, subset = year == jYear,  
      xlab = jXlab, ylab = jYlab,  
      scales = list(x = list(log = 10)),  
      xlim = jXlim, ylim = jYlim,  
      xscale.components = xscale.components.log10)
```

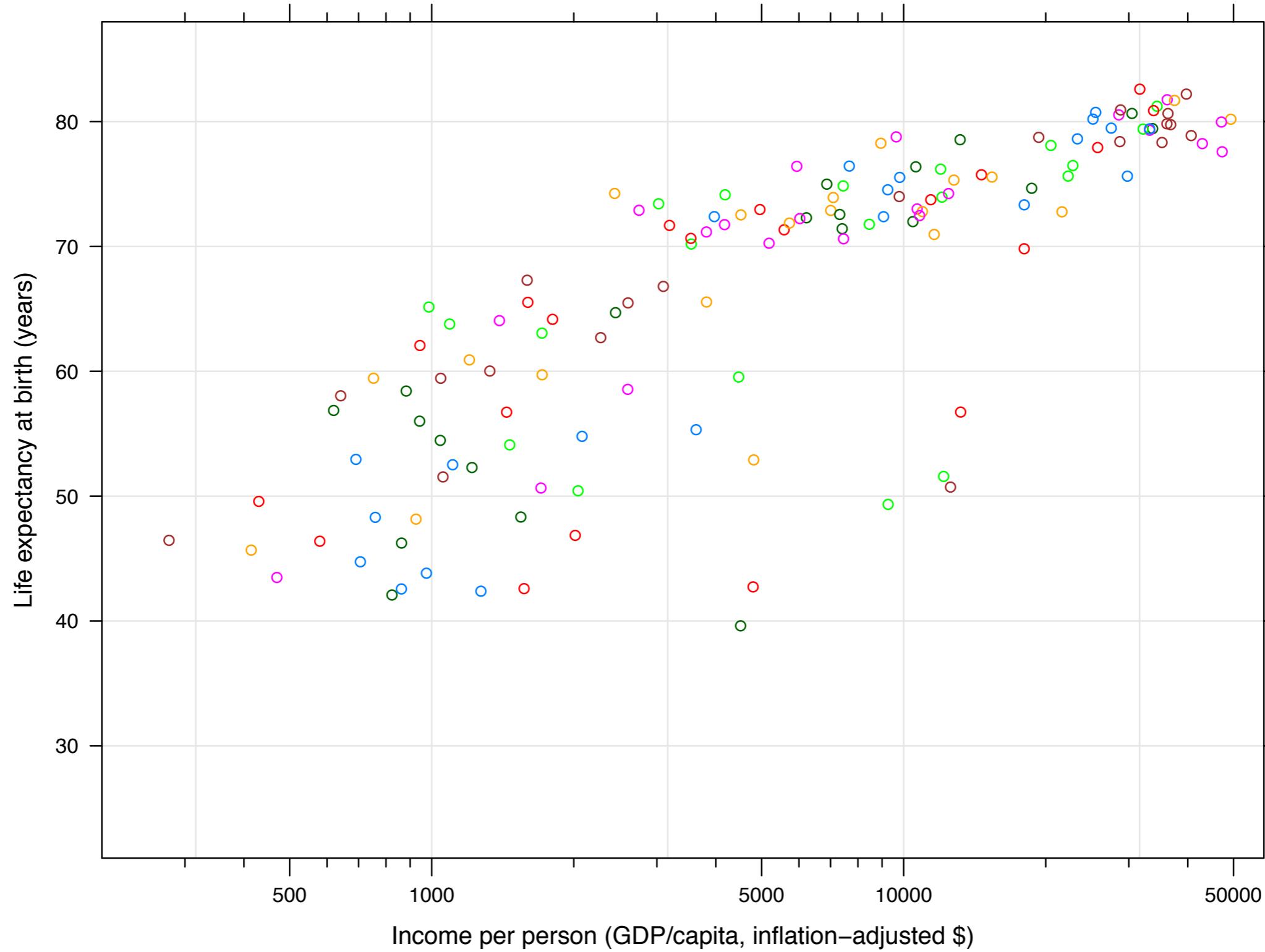
For even more control, can write a custom function for this. Kind of advanced / obscure. See the code.



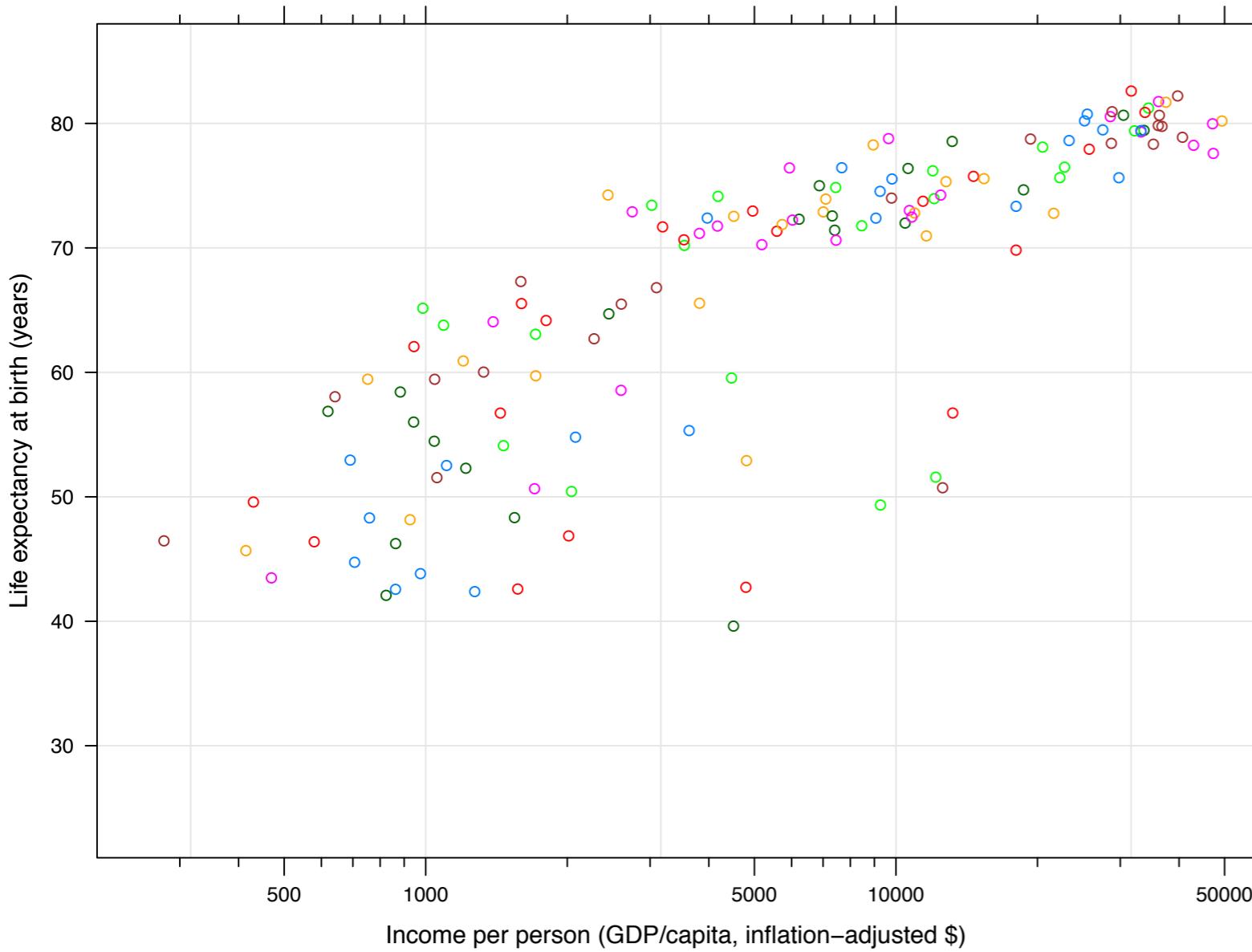
```
xyplot(lifeExp ~ gdpPerCap, <snip, snip>,
       type = c("p", "g"))
```



```
xyplot(lifeExp ~ gdpPerCap, <snip, snip>,  
      grid = TRUE)
```

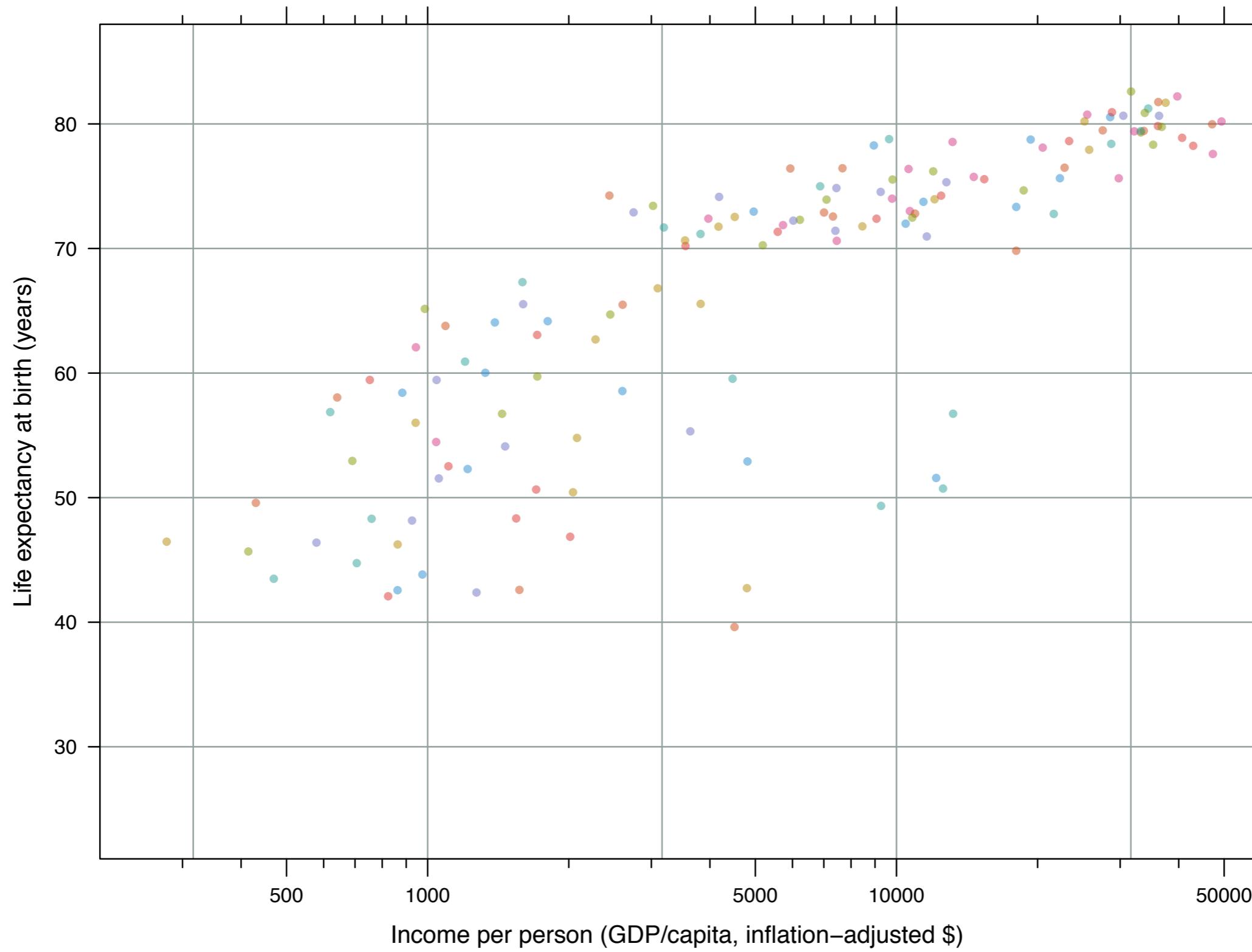


```
xyplot(lifeExp ~ gdpPerCap, ..., group = country)
```

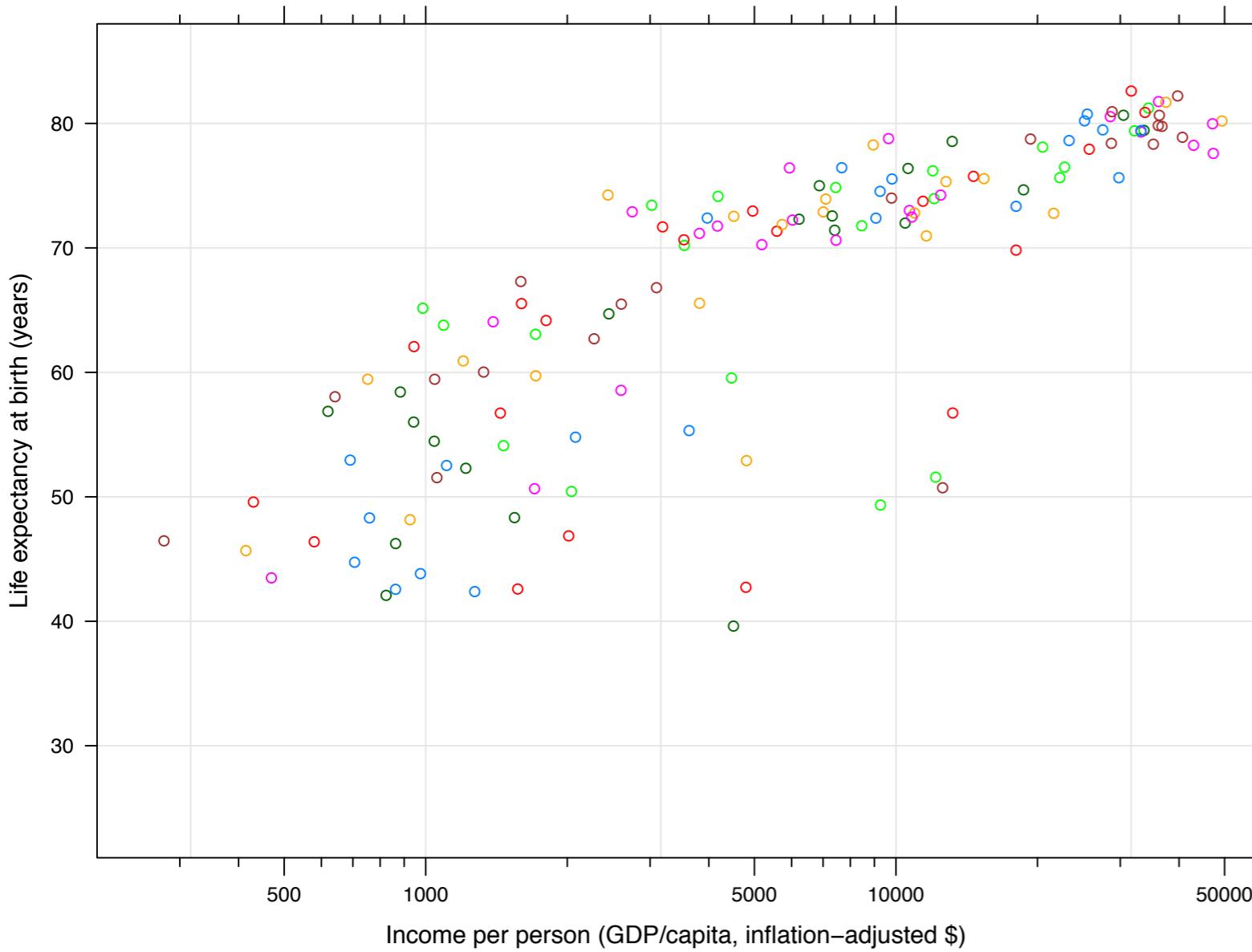


```
xyplot(lifeExp ~ gdpPercap, ..., group = country)
```

The group argument specifies a factor to be “superposed” in the plot, generally via color. Each level of the factor will be visually distinguished.

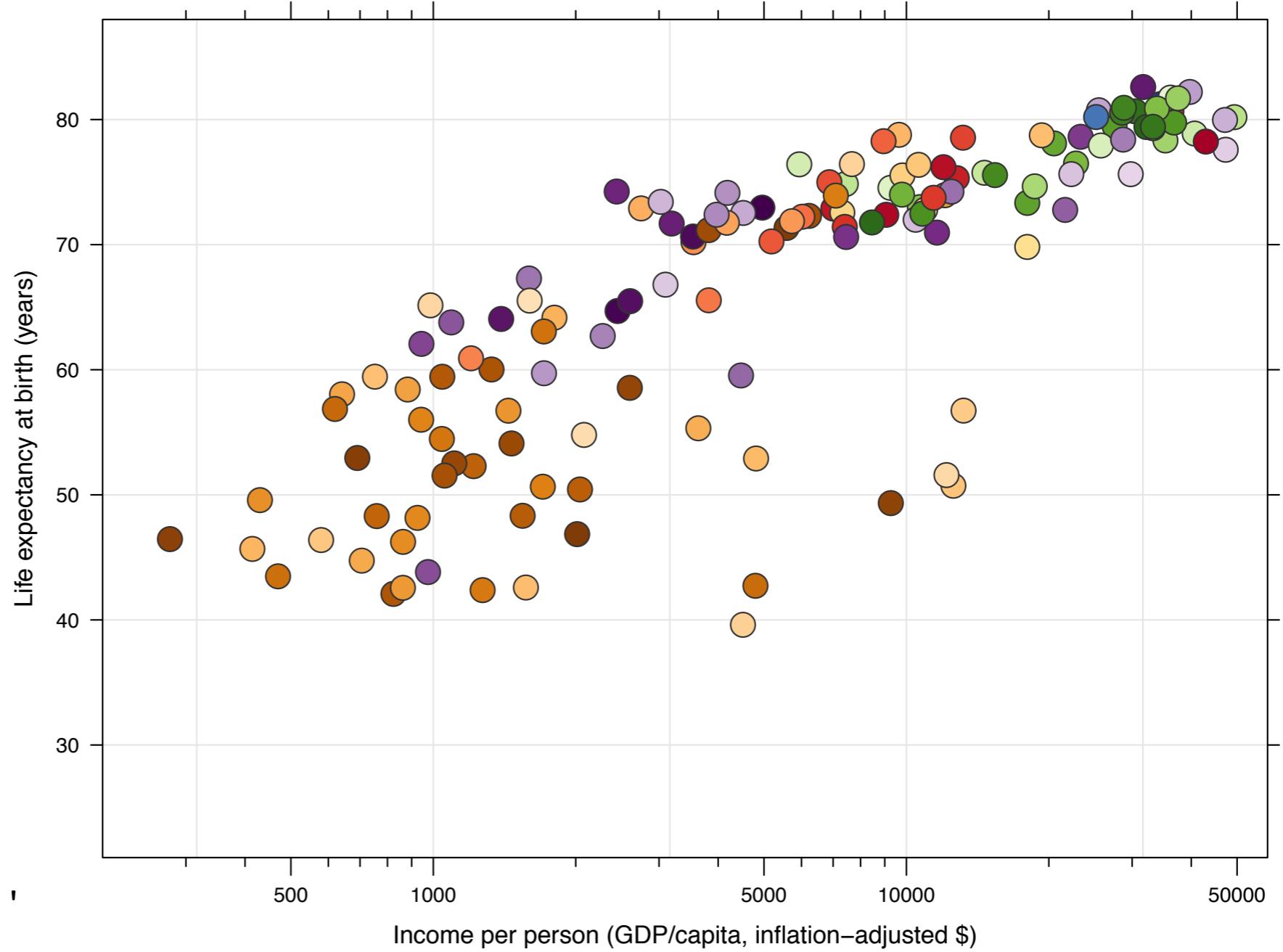


```
xyplot(lifeExp ~ gdpPercap, ..., group = country)
```



```
xyplot(lifeExp ~ gdpPercap, ..., group = country)
```

Here, we have two problems to solve: the default color palette is ugly and it is not large enough for our country factor.



```
jDarkGray <- 'grey20'
jPch <- 21
jGapminderPars <-
  list(superpose.symbol = list(pch = jPch, col = jDarkGray, cex = 2,
    fill = countryColors$color[match(levels(gDatOrdered$country),
      countryColors$country)]))

xyplot(lifeExp ~ gdpPerCap, ...
  group = country,
  par.settings = jGapminderPars)
```

You can specify your own color palette for a plot.

## Common Bivariate Trellis Plots

## Description:

This help page documents several commonly used high-level Lattice functions. 'xyplot' produces bivariate scatterplots or time-series plots.....

## Usage:

```
xyplot(x, data, ...)
```

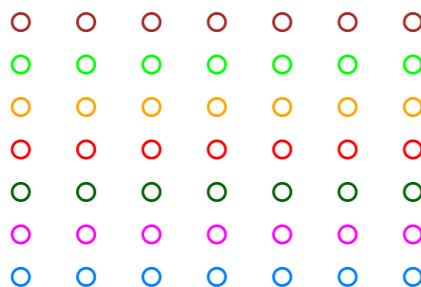
```
<snip, snip>
```

**Here's where one sets graphing symbols, line types, colors, alpha transparency, etc etc etc.**

'par.settings': A list that could be supplied to 'trellis.par.set'. When the resulting object is plotted, these options are applied temporarily for the duration of the plotting, after which the settings revert back to what they were before. This enables the user to attach some display settings to the trellis object itself rather than change the settings globally. See also the 'lattice.options' argument described above for a similar treatment of non-graphical options.

`show.settings()`

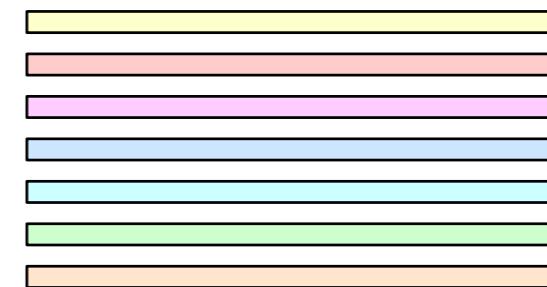
These are the settings in the default theme.



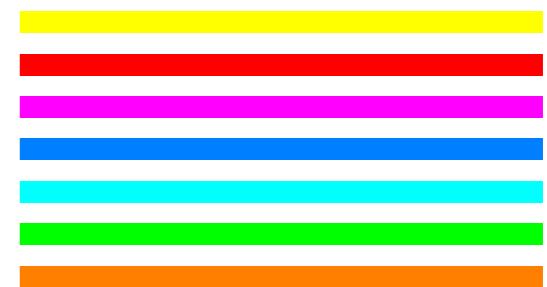
`superpose.symbol`



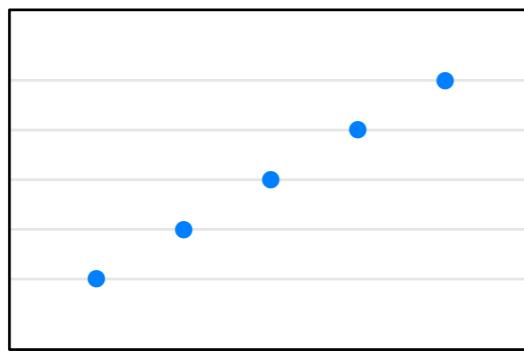
`superpose.line`



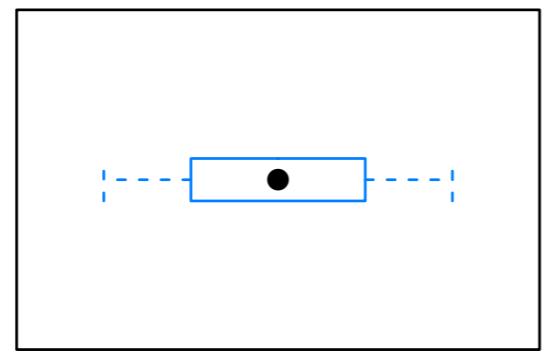
`strip.background`



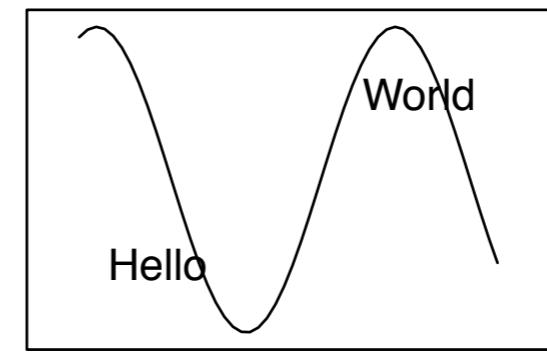
`strip.shingle`



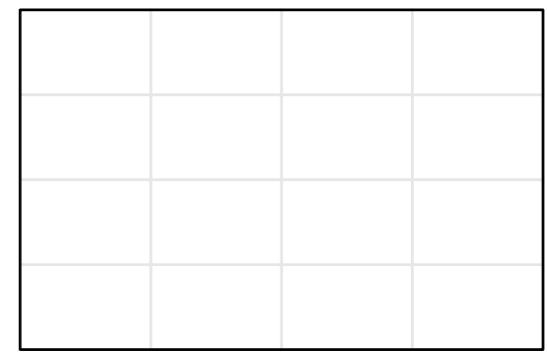
`dot.[symbol, line]`



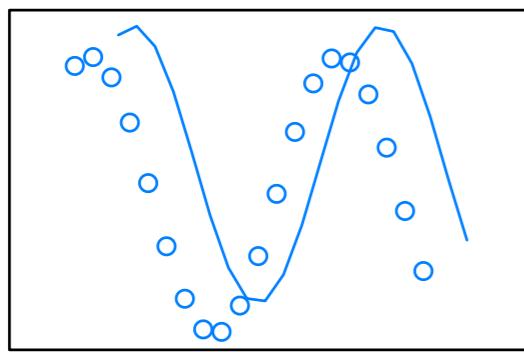
`box.[dot, rectangle, umbrella]`



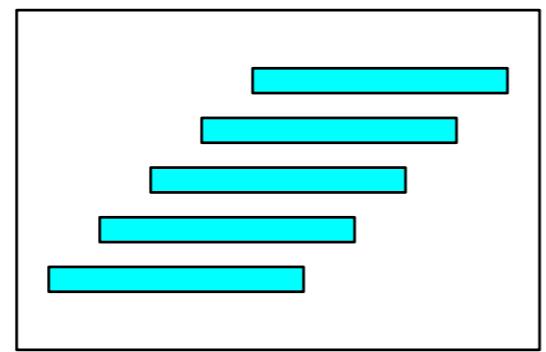
`add.[line, text]`



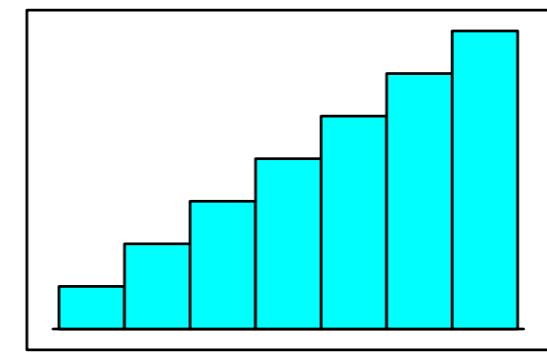
`reference.line`



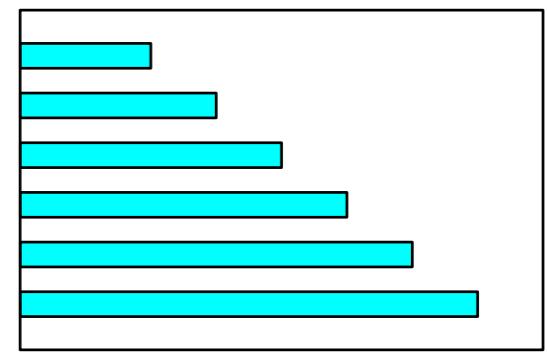
`plot.[symbol, line]`



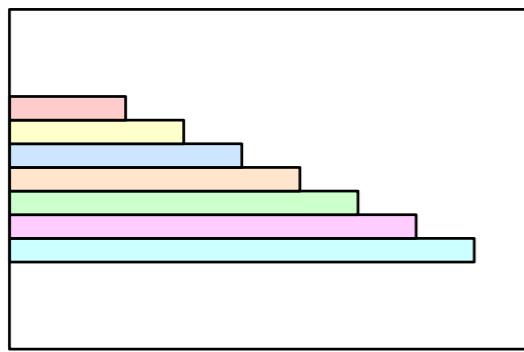
`plot.shingle[plot.polygon]`



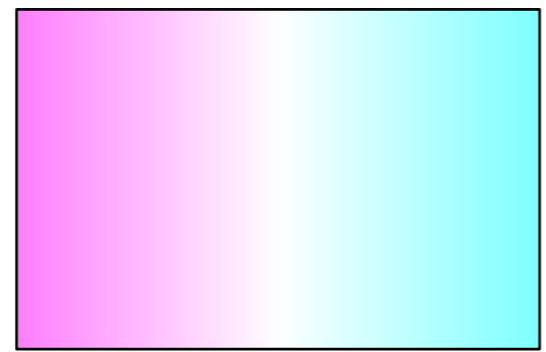
`histogram[plot.polygon]`



`barchart[plot.polygon]`



`superpose.polygon`



`regions`

```
> trellis.par.get()
```

```
<snip, snip>
```

```
$superpose.line
```

```
$superpose.line$alpha
```

```
[1] 1
```

```
$superpose.line$col
```

```
[1] "#0080ff" "#ff00ff" "darkgreen" "#ff0000" "orange" "#00ff00"
```

```
[7] "brown"
```

```
$superpose.line$lty
```

```
[1] 1 1 1 1 1 1 1
```

```
$superpose.line$lwd
```

```
[1] 1 1 1 1 1 1 1
```

```
$superpose.symbol
```

```
$superpose.symbol$alpha
```

```
[1] 1 1 1 1 1 1 1
```

```
$superpose.symbol$cex
```

```
[1] 0.8 0.8 0.8 0.8 0.8 0.8 0.8
```

```
$superpose.symbol$col
```

```
[1] "#0080ff" "#ff00ff" "darkgreen" "#ff0000" "orange" "#00ff00"
```

```
[7] "brown"
```

```
$superpose.symbol$fill
```

```
[1] "#CCFFFF" "#FFCCFF" "#CCFFCC" "#FFE5CC" "#CCE6FF" "#FFFFCC" "#FFCCCC"
```

```
$superpose.symbol$font
```

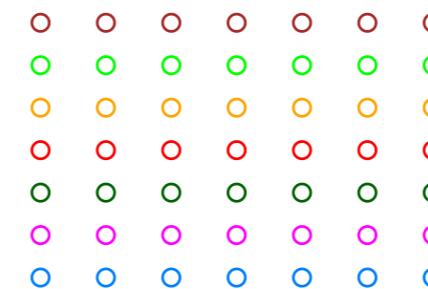
```
[1] 1 1 1 1 1 1 1
```

```
$superpose.symbol$pch
```

```
[1] 1 1 1 1 1 1 1
```

```
<snip, snip>
```

# These are the settings in the default theme.



superpose.symbol

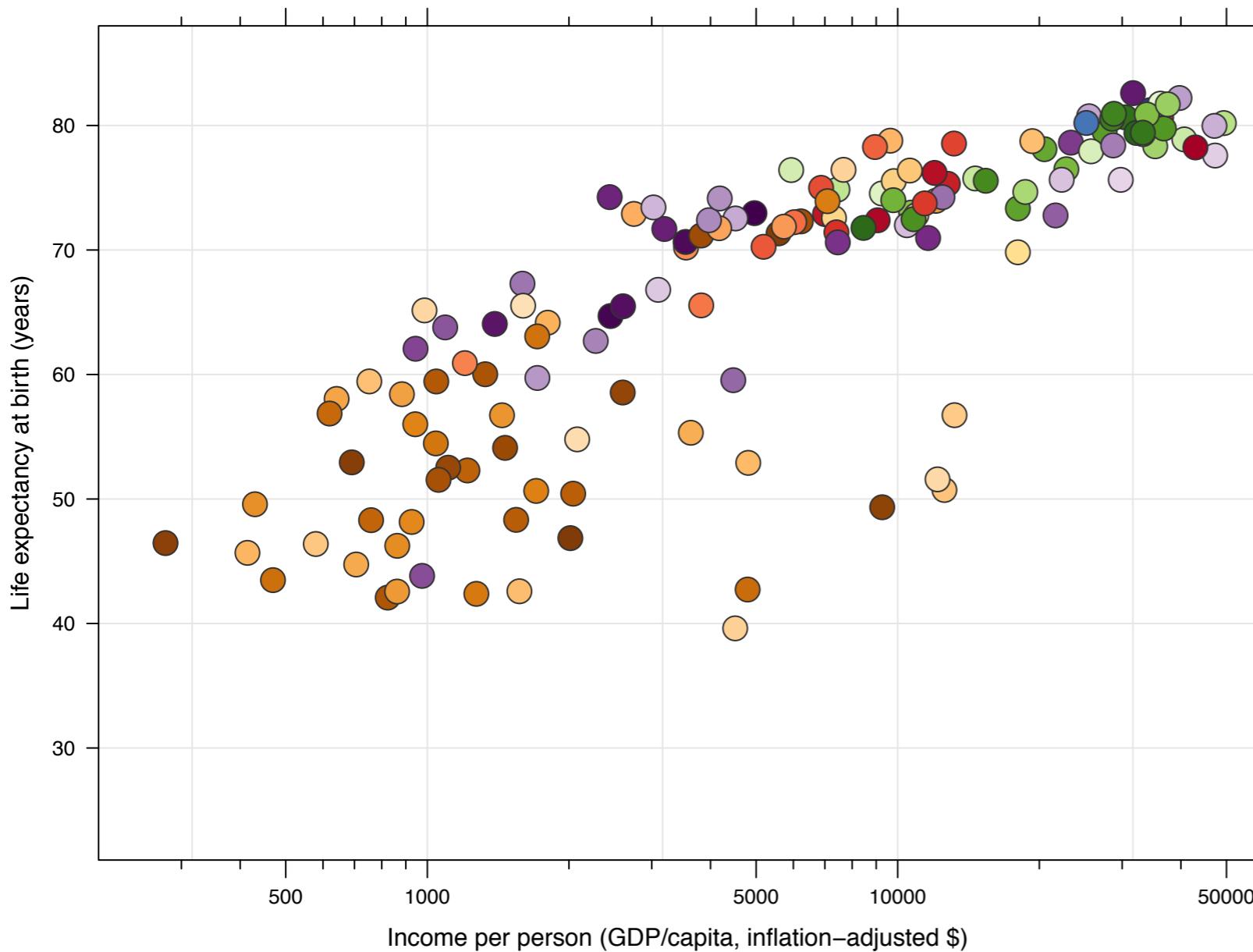


superpose.line

6 ▽ 7 ⊗ 8 \* 9 ♦ 10 ⊕ 11 ♠

0		"blank"
1	—	"solid"
2	- - - - -	"dashed"
3	· · · · ·	"dotted"

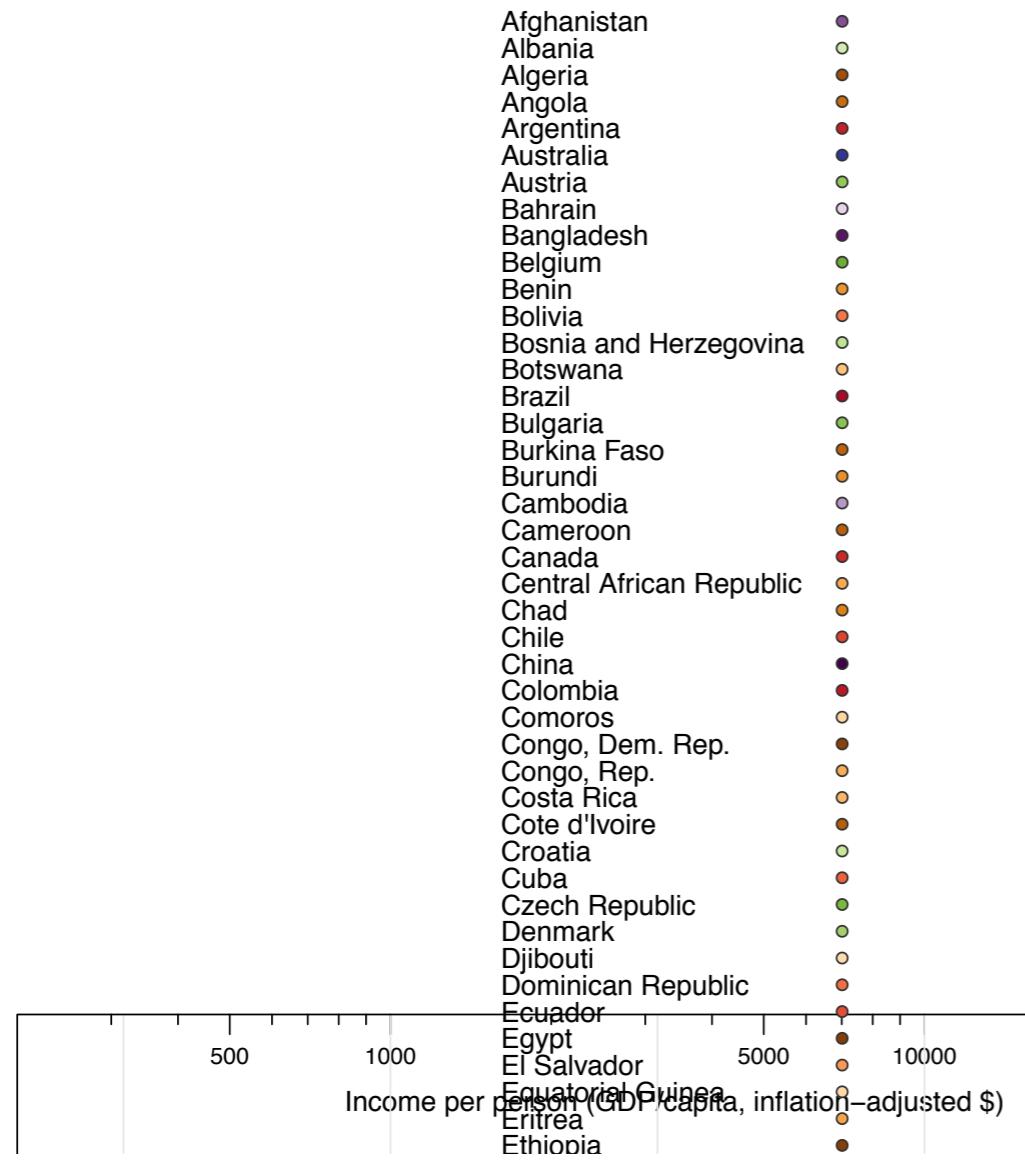
0 □ 1 ○ 2 △ 3 + 4 × 5 ◇



```
jGapminderPars <-
  list(superpose.symbol = list(pch = jPch, col = jDarkGray, cex = 2,
    fill = countryColors$color[match(levels(gDatOrdered$country),
      countryColors$country)]))

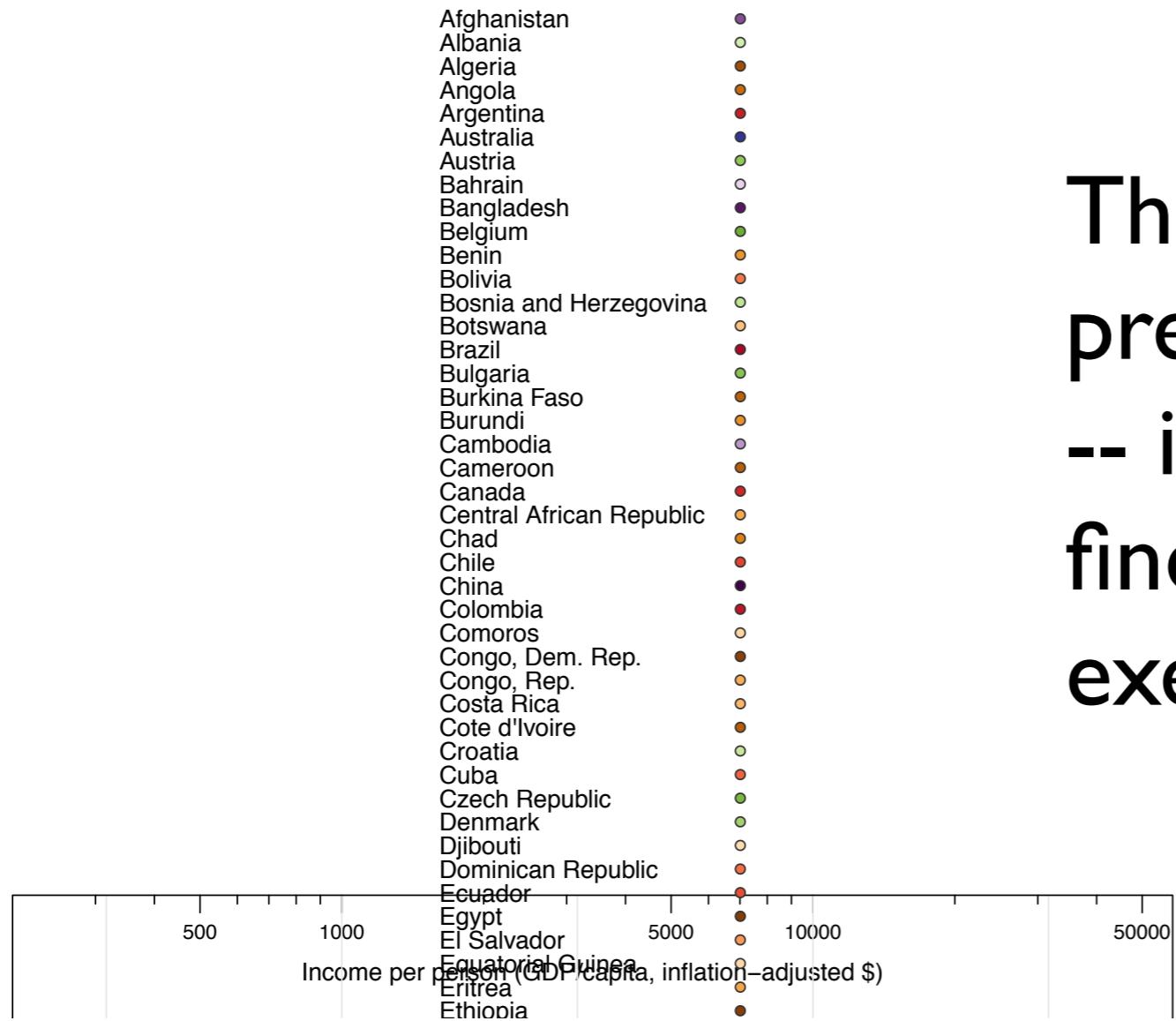
xyplot(lifeExp ~ gdpPerCap, ...
  group = country,
  par.settings = jGapminderPars)
```

I redefined the graphics settings controlling plot symbol, fill color, and size of superposed points.



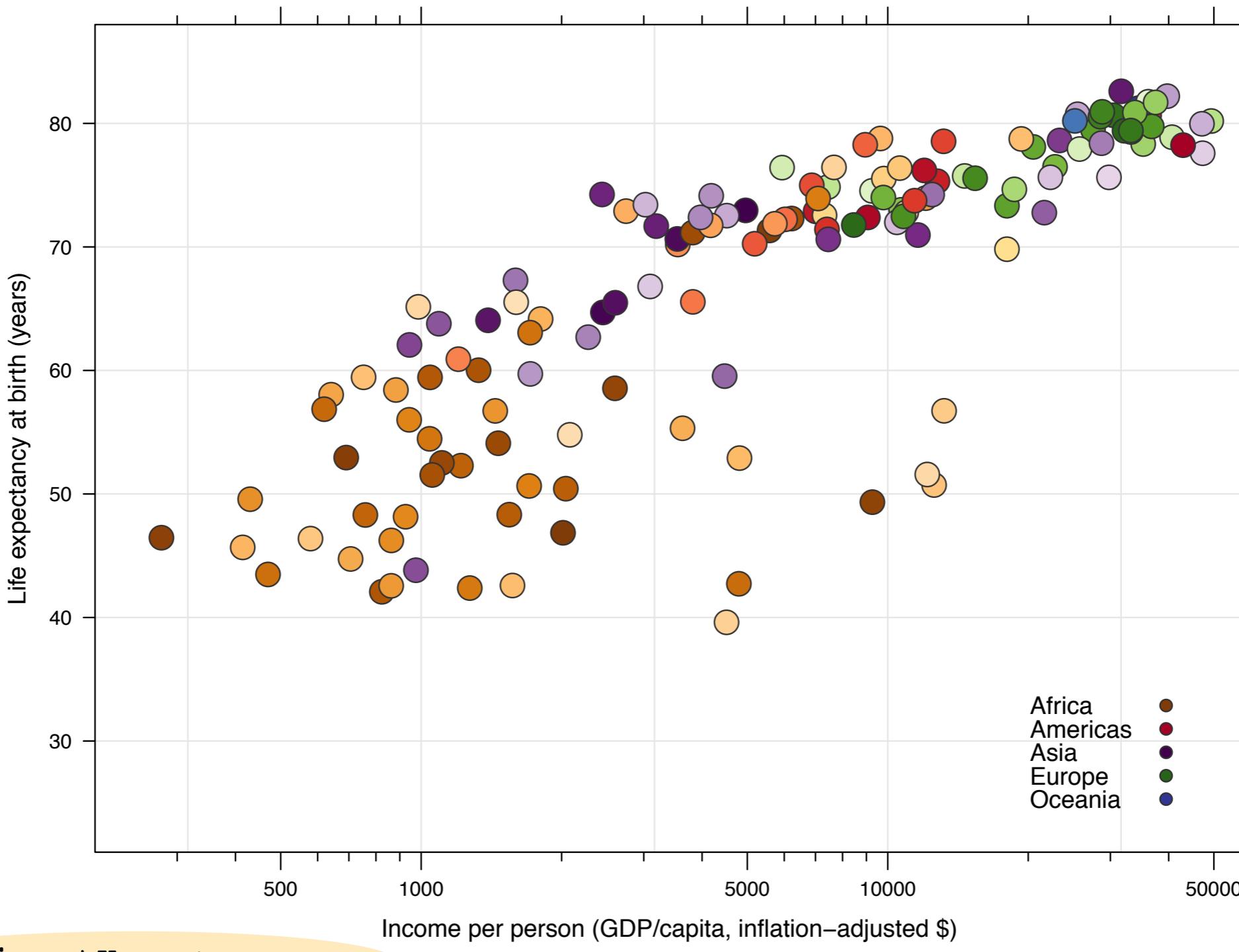
The `auto.key` argument is incredibly useful for adding a legend automatically. In many situations, `auto.key = TRUE` works perfectly. In many others, `auto.key` can be used to provide a bit more info, e.g. specifying the location of the legend, while letting everything else happen automatically.

```
xyplot(lifeExp ~ gdpPercap, ...
       group = country,
       par.settings = jGapminderPars,
       auto.key = TRUE)
```



The Gapminder data presents some tricky issues -- it's a case where even finer control must be exerted.

```
xyplot(lifeExp ~ gdpPercap, ...
       group = country,
       par.settings = jGapminderPars,
       auto.key = TRUE)
```

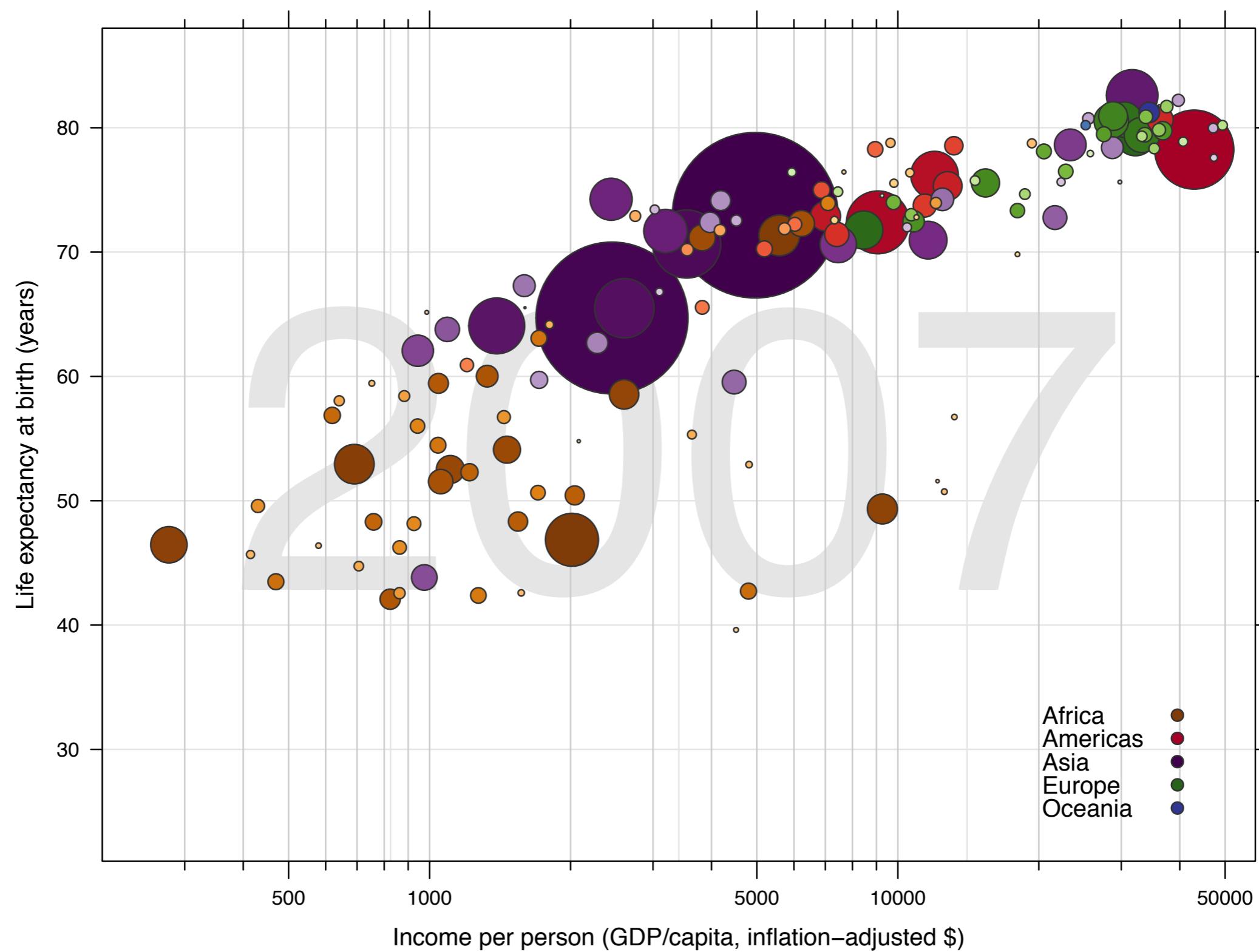


```

continentKey <-
  with(continentColors,
    list(x = 0.95, y = 0.05, corner = c(1, 0),
         text = list(as.character(continent)),
         points = list(pch = jPch, col = jDarkGray, fill = color)))
}

xyplot(lifeExp ~ gdpPerCap, ...,
       key = continentKey)

```



Finally ... sizing the circles s.t. area = pop. Again, it's a bit tricky -- not a problem encountered in a typical figure, so don't be discouraged. How to do?

```
yDat <- subset(gDatOrdered, year == jYear)
xyplot(lifeExp ~ gdpPercap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(x, y, ...) {
         grid.text(jYear,
                   gp = gpar(fontsize = jFontSize, col = jLightestGray))
         panel.grid(h = -1, v = 0, col.line = jLightGray)
         panel.abline(v = log10(logTicks(c(10, 50000), loc = 1:9)),
                       col.line = jLightGray)
         panel.points(x, y, pch = jPch,
                      cex = jPopRadFun(yDat$pop)/jCexDivisor,
                      col = jDarkGray, fill = yDat$color)
       })
})
```

Example of redefining the panel function ‘on the fly’. Though it looks geeky, this is in fact a very common way to exert control over lattice figures.

If the control you need cannot be exerted by clever use of arguments or (temporarily) changing some graphics settings, then you may need to rewrite the panel function.

This is not as awful as it sounds.

Borrowing heavily from Sarkar's talk "Lattice Tricks for the Power UseR" from UseR! 2007 conference .....

Things to know about panel functions:

Panel functions are functions (!)

They are responsible for graphical content inside panels

They get executed once for every panel

Every high level function has a default panel function

e.g., `xyplot()` has default panel function `panel.xyplot()`

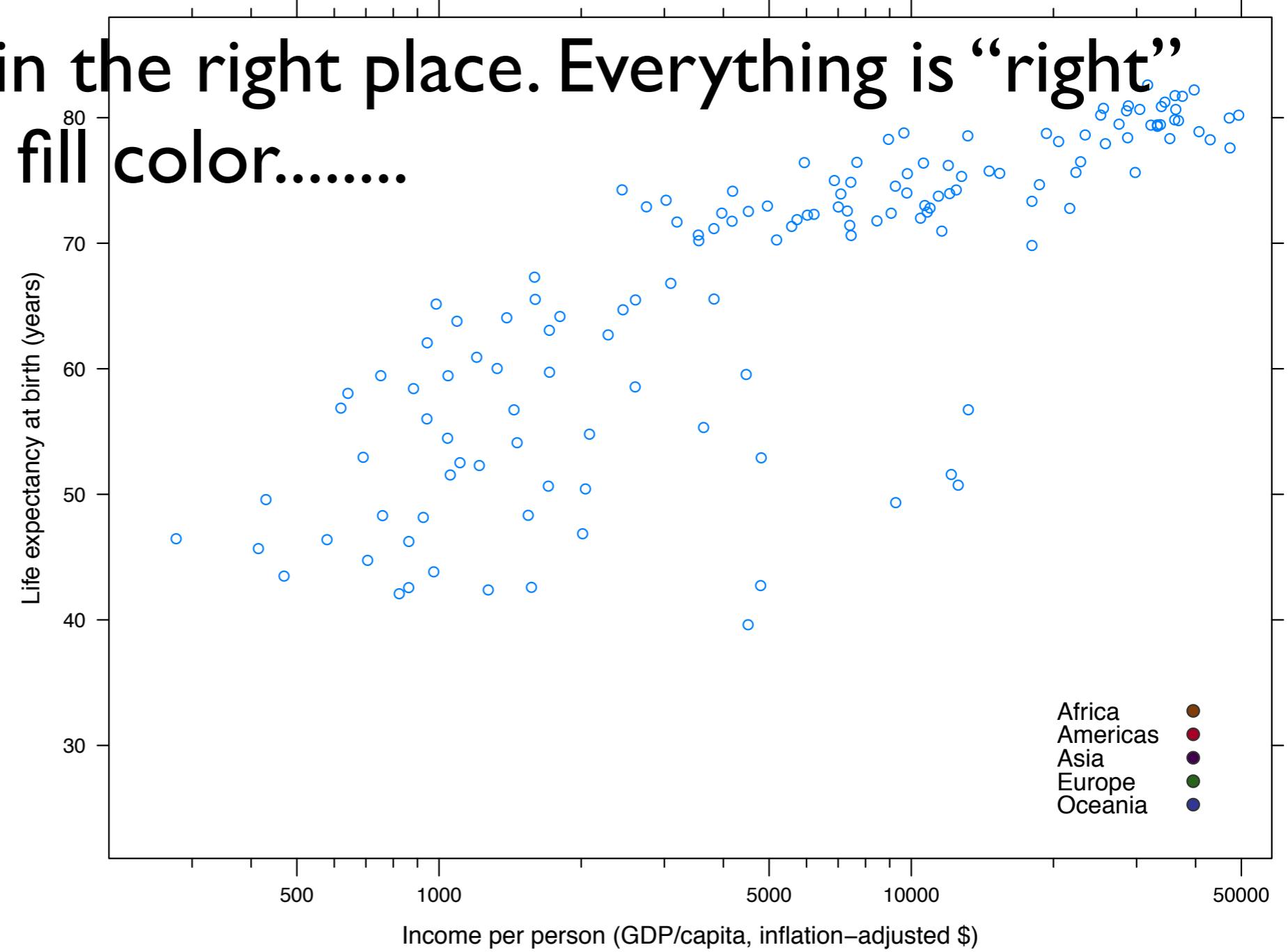
## Workflow for rewriting (e.g. extending) a panel function:

Start with a graphing command (e.g. a call to `xyplot`) that “works”, i.e. it has nothing wrong with it -- it’s just missing some cool features.

Rewrite the panel function but without adding any functionality!

Gradually add the elements you need.

The points are in the right place. Everything is “right” except size and fill color.....



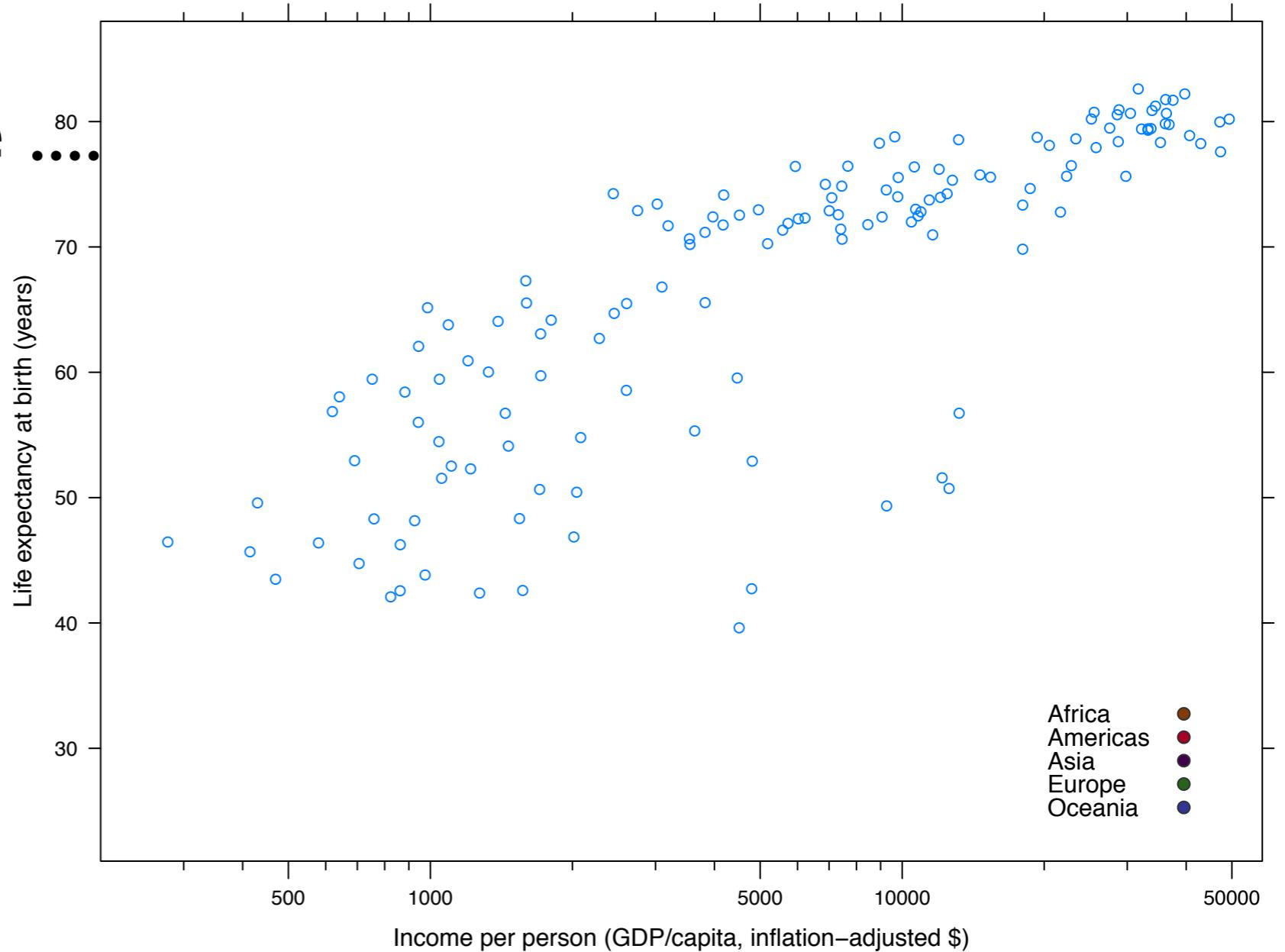
```
yDat <- subset(gDatOrdered, year == jYear)
xyplot(lifeExp ~ gdpPerCap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = panel.xyplot)
```

The points are in the right place. Everything is “right” except size and fill color.....



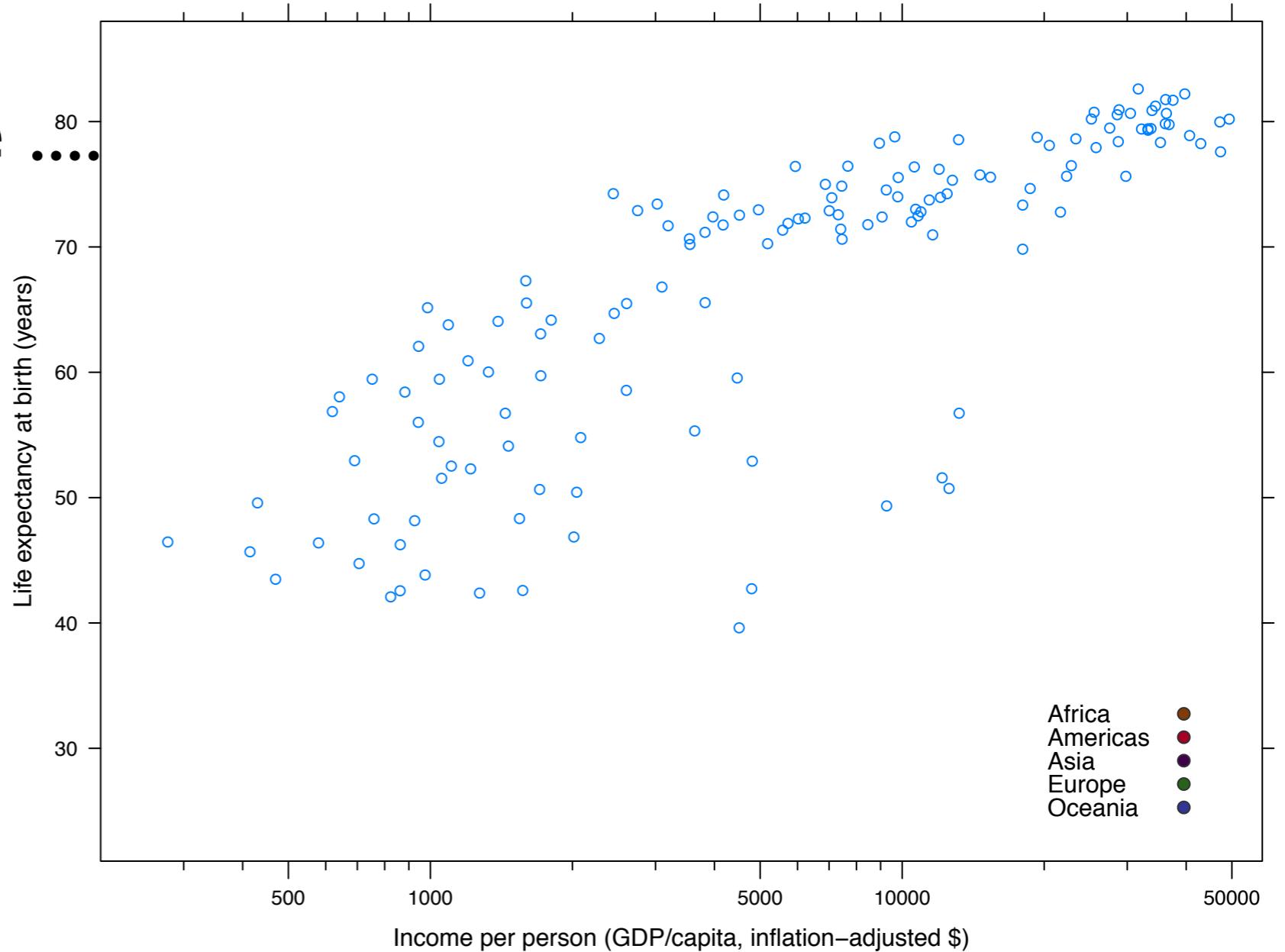
```
yDat <- subset(gDatOrdered, year == jYear)
xyplot(lifeExp ~ gdpPerCap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = panel.xyplot)
```

Same plot as before ....



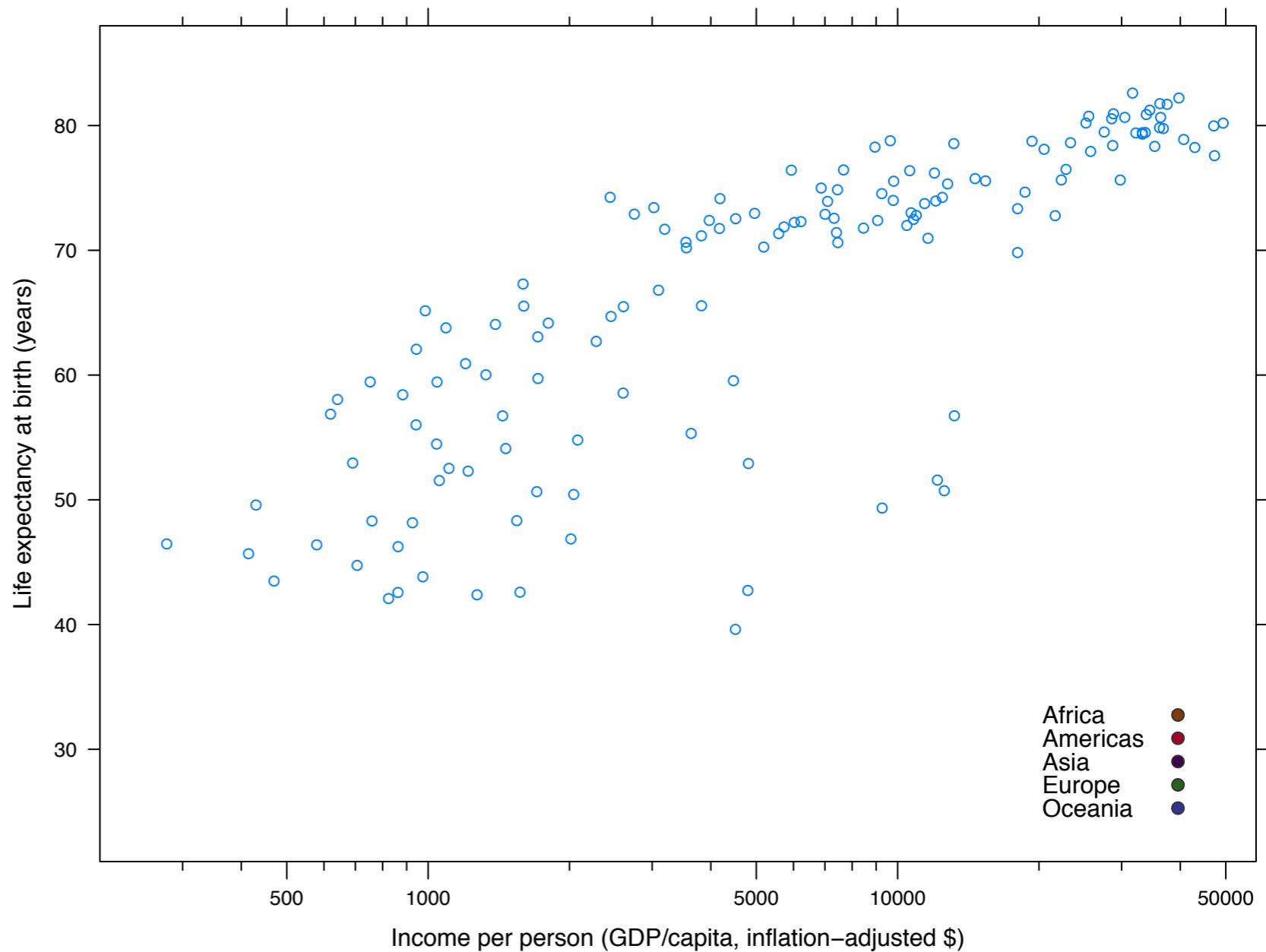
```
xyplot(lifeExp ~ gdpPercap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(...) {
         panel.xyplot(...)
```

# Same plot as before ....



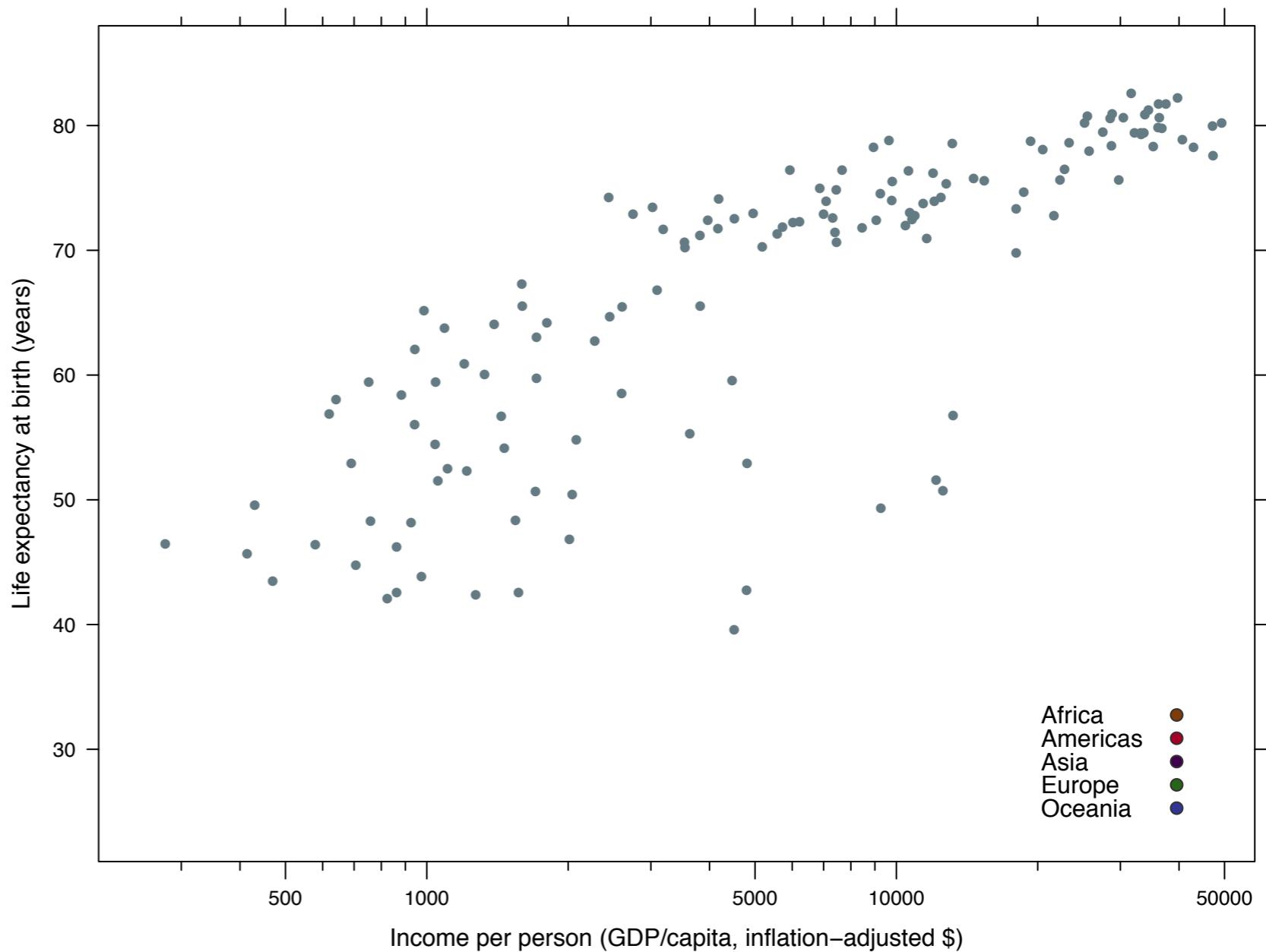
```
xyplot(lifeExp ~ gdpPerCap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(x, y, ...) {
         panel.xyplot(x, y, ...)
       })
```

Same plot as  
before .... but I've  
drawn the points  
myself!



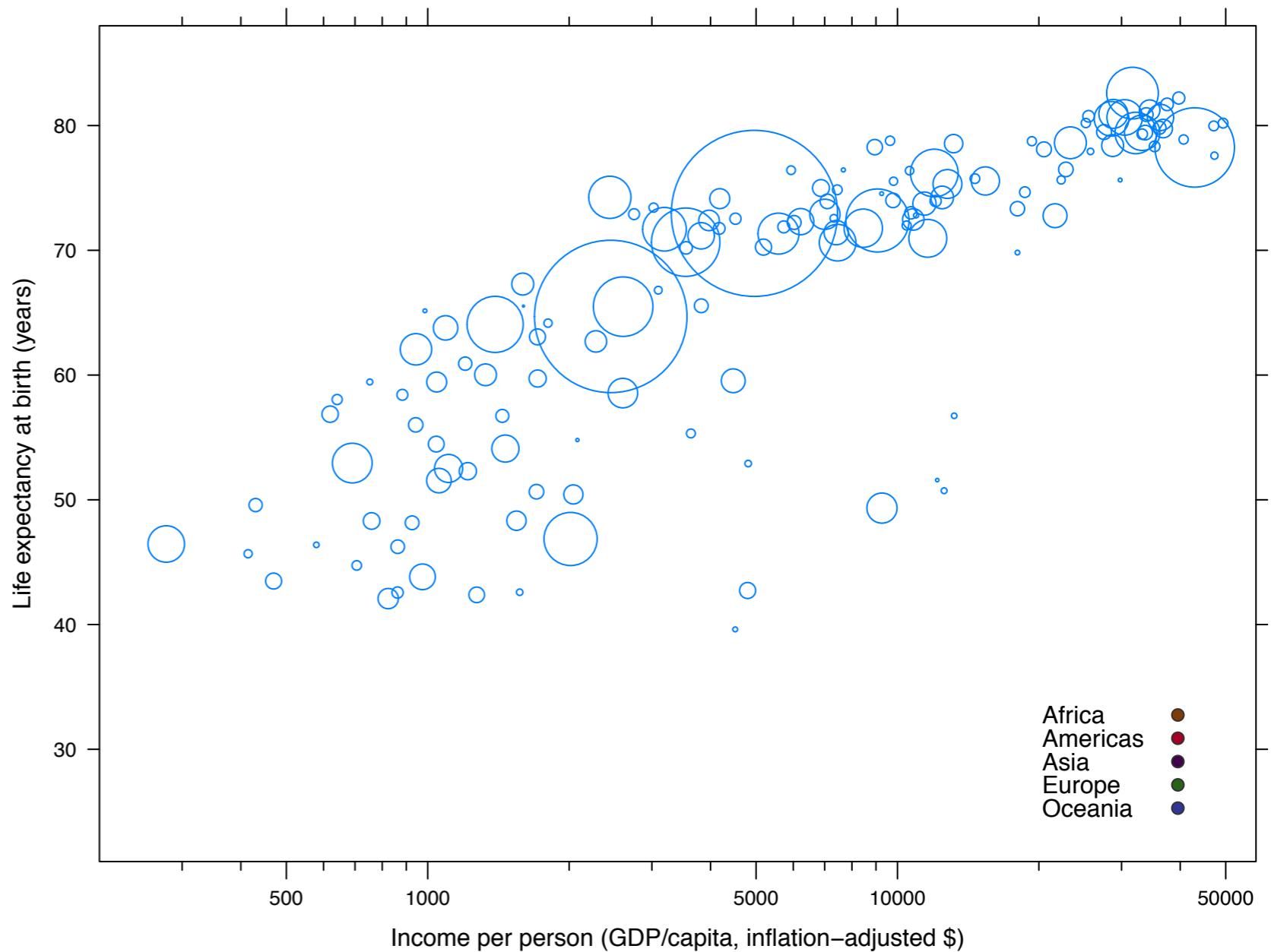
```
xyplot(lifeExp ~ gdpPercap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(x, y, ...) {
         panel.points(x, y)
       })
```

Same plot as  
before .... but I've  
drawn the points  
myself!



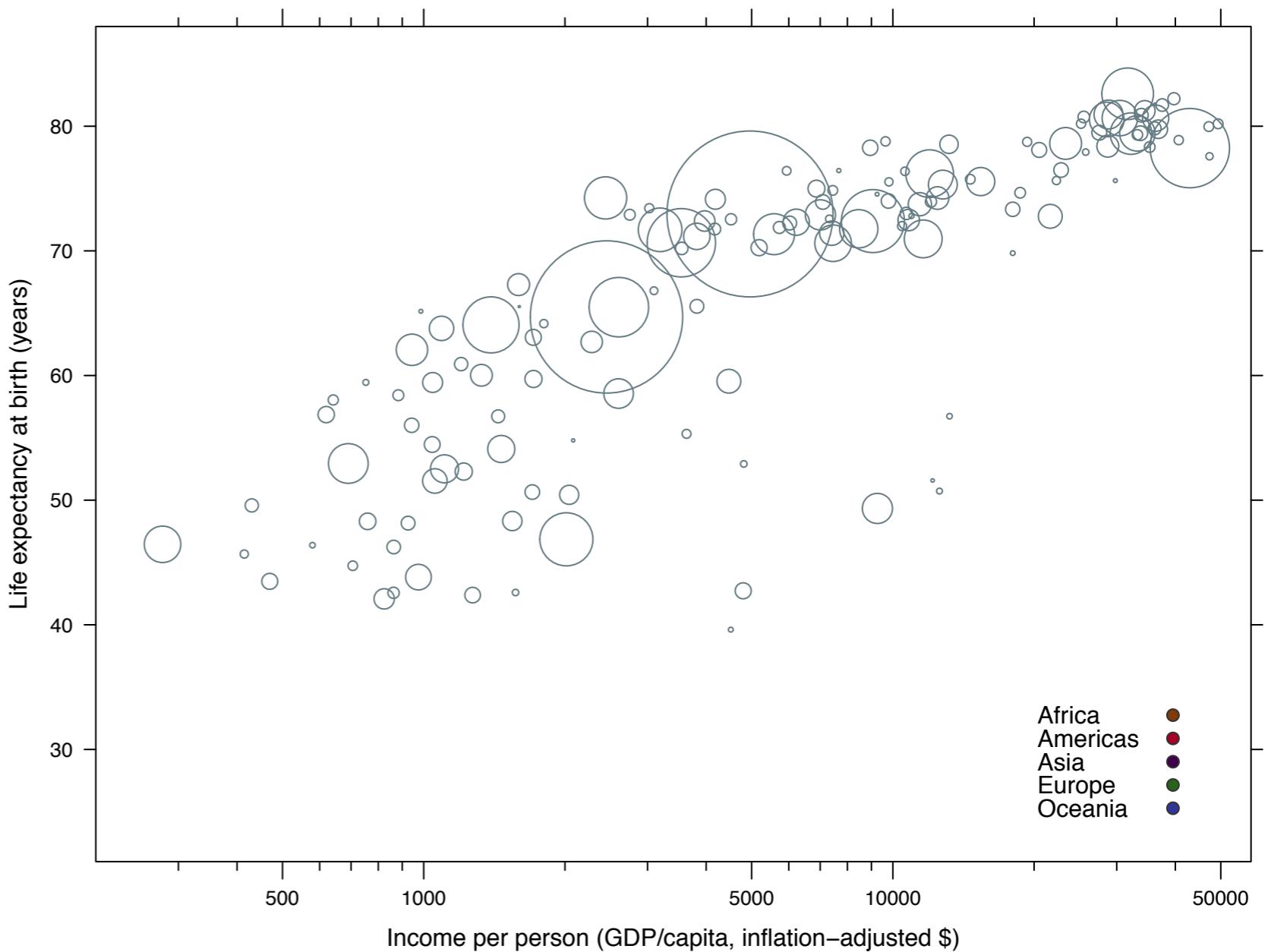
```
xyplot(lifeExp ~ gdpPercap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(x, y, ...) {
         panel.points(x, y)
       })
```

Now I'm  
controlling the  
size! Yay!



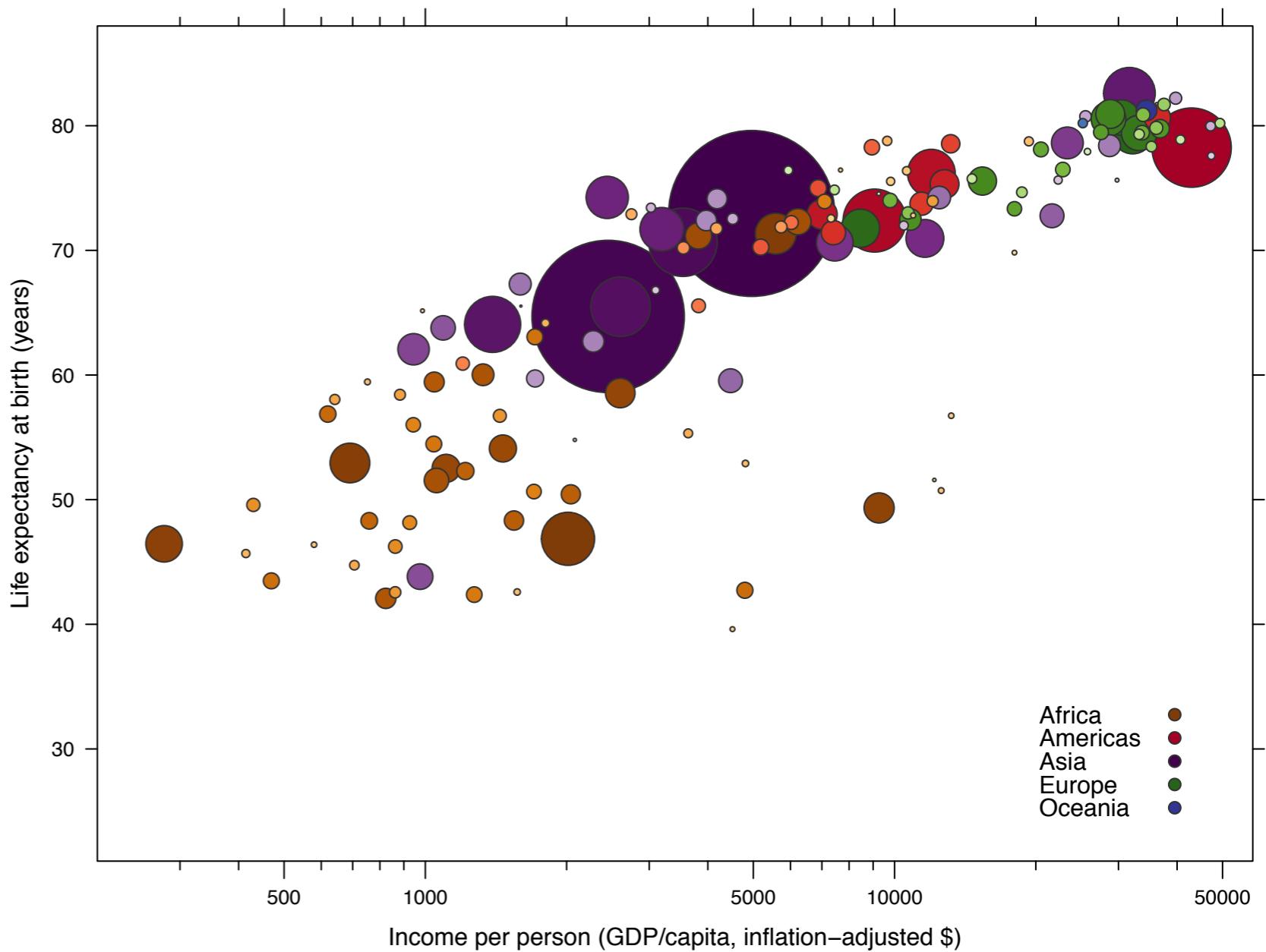
```
xyplot(lifeExp ~ gdpPercap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(x, y, ...) {
         panel.points(x, y, pch = jPch,
                      cex = jPopRadFun(yDat$pop)/jCexDivisor)
       })
```

# Now I'm controlling the size! Yay!



```
xyplot(lifeExp ~ gdpPercap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(x, y, ...) {
         panel.points(x, y, pch = jPch,
                      cex = jPopRadFun(yDat$pop)/jCexDivisor)
       })
```

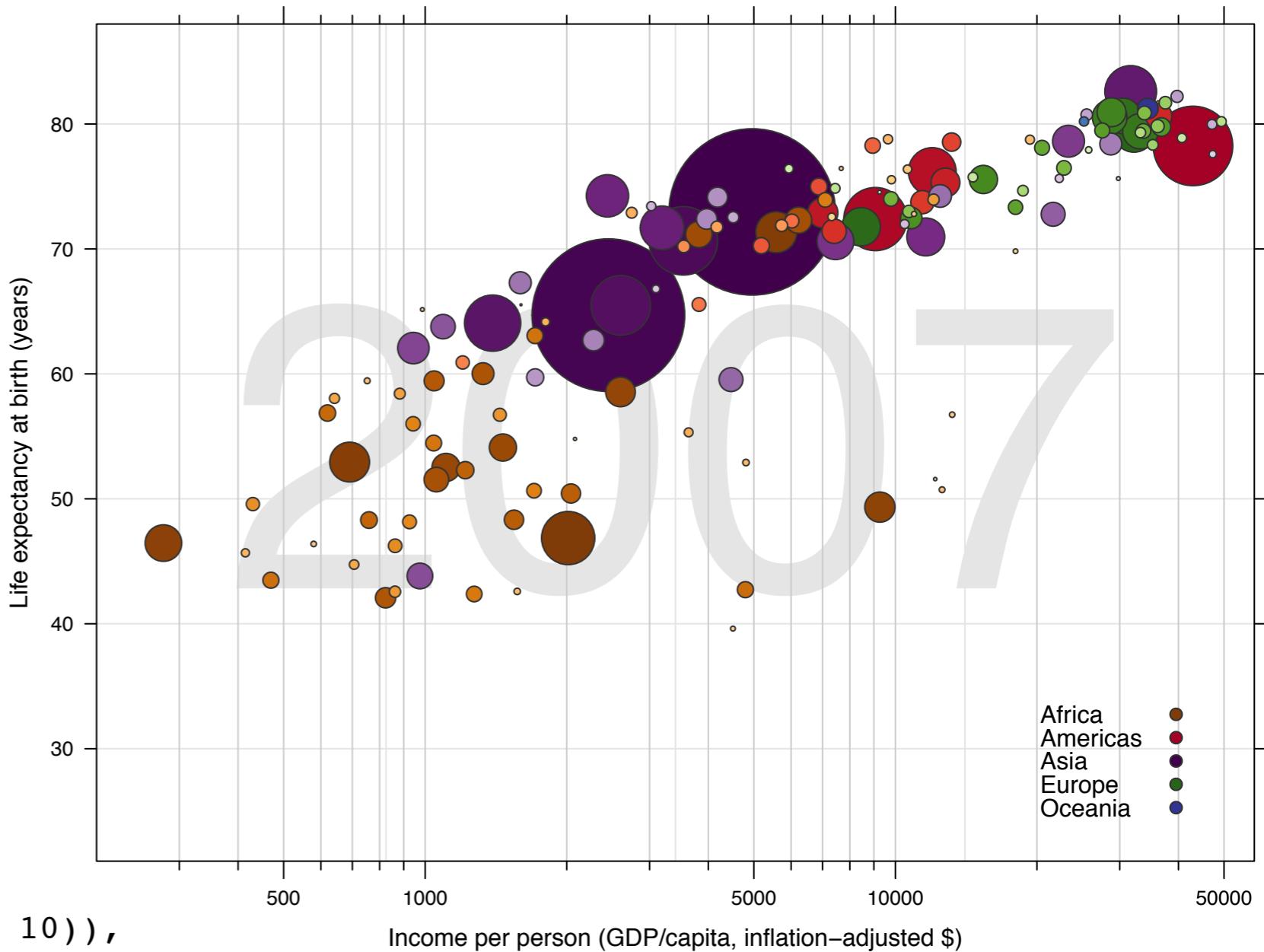
Now I'm  
controlling the  
color! YES!



```
xyplot(lifeExp ~ gdpPercap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(x, y, ...) {
         panel.points(x, y, pch = jPch,
                     cex = jPopRadFun(yDat$pop)/jCexDivisor,
                     col = jDarkGray, fill = yDat$color)
       })
})
```

# Reference grid! Year!

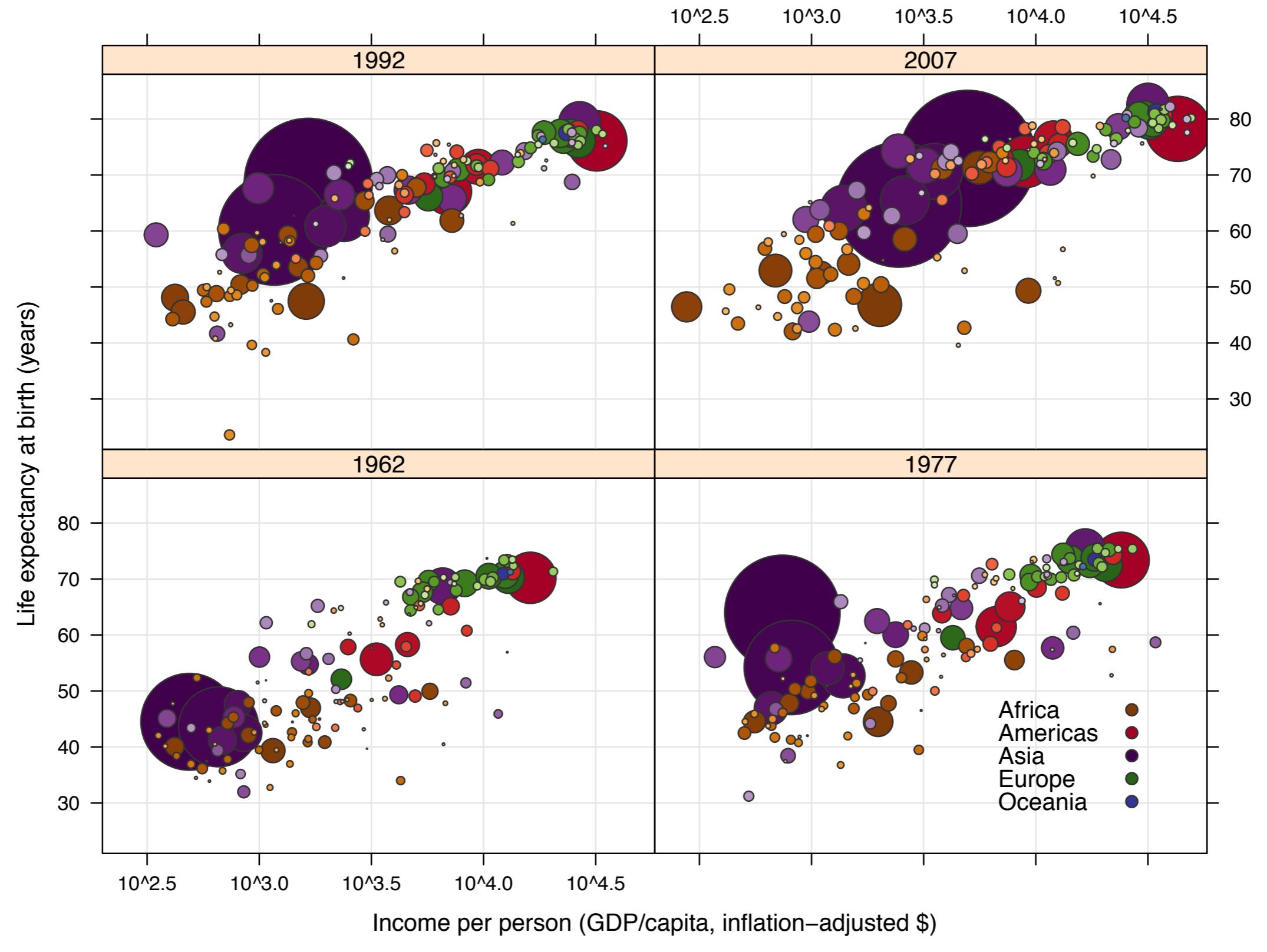
```
xyplot(lifeExp ~ gdpPerCap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(x, y, ...) {
         grid.text(jYear,
                   gp = gpar(fontsize = jFontSize, col = jLightestGray))
         panel.grid(h = -1, v = 0, col.line = jLightGray)
         panel.abline(v = log10(logTicks(c(10, 50000), loc = 1:9)),
                      col.line = jLightGray)
         panel.points(x, y, pch = jPch,
                      cex = jPopRadFun(yDat$pop)/jCexDivisor,
                      col = jDarkGray, fill = yDat$color)
       })
```



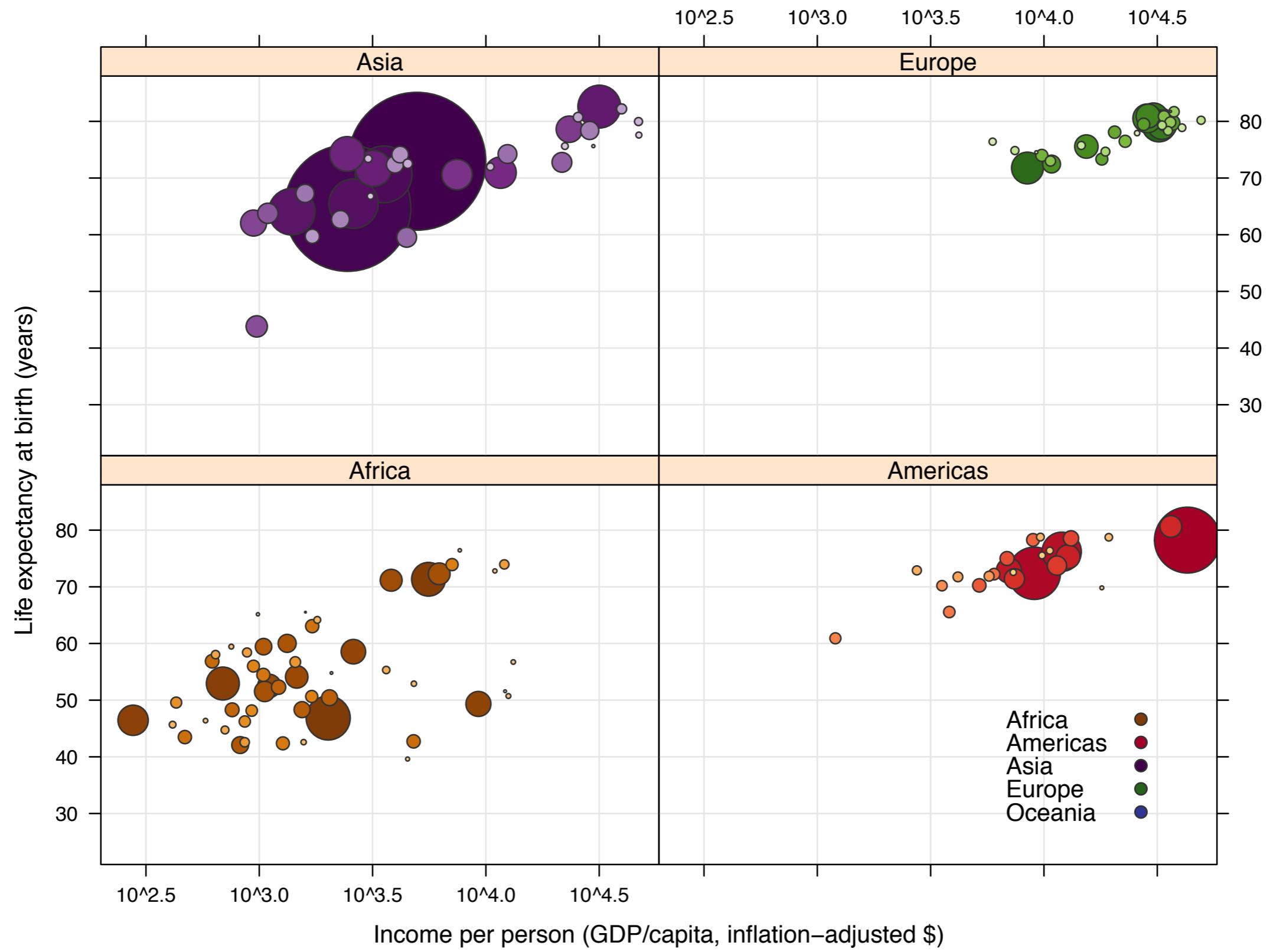
Africa  
Americas  
Asia  
Europe  
Oceania

```
xyplot(lifeExp ~ gdpPercap, yDat,
       xlab = jXlab, ylab = jYlab,
       scales = list(x = list(log = 10)),
       xlim = jXlim, ylim = jYlim,
       xscale.components = xscale.components.log10,
       key = continentKey,
       panel = function(x, y, ...) {
         grid.text(jYear,
                   gp = gpar(fontsize = jFontSize, col = jLightestGray))
         panel.grid(h = -1, v = 0, col.line = jLightGray)
         panel.abline(v = log10(logTicks(c(10, 50000), loc = 1:9)),
                      col.line = jLightGray)
         panel.points(x, y, pch = jPch,
                      cex = jPopRadFun(yDat$pop)/jCexDivisor,
                      col = jDarkGray, fill = yDat$color)
       })
     }
```

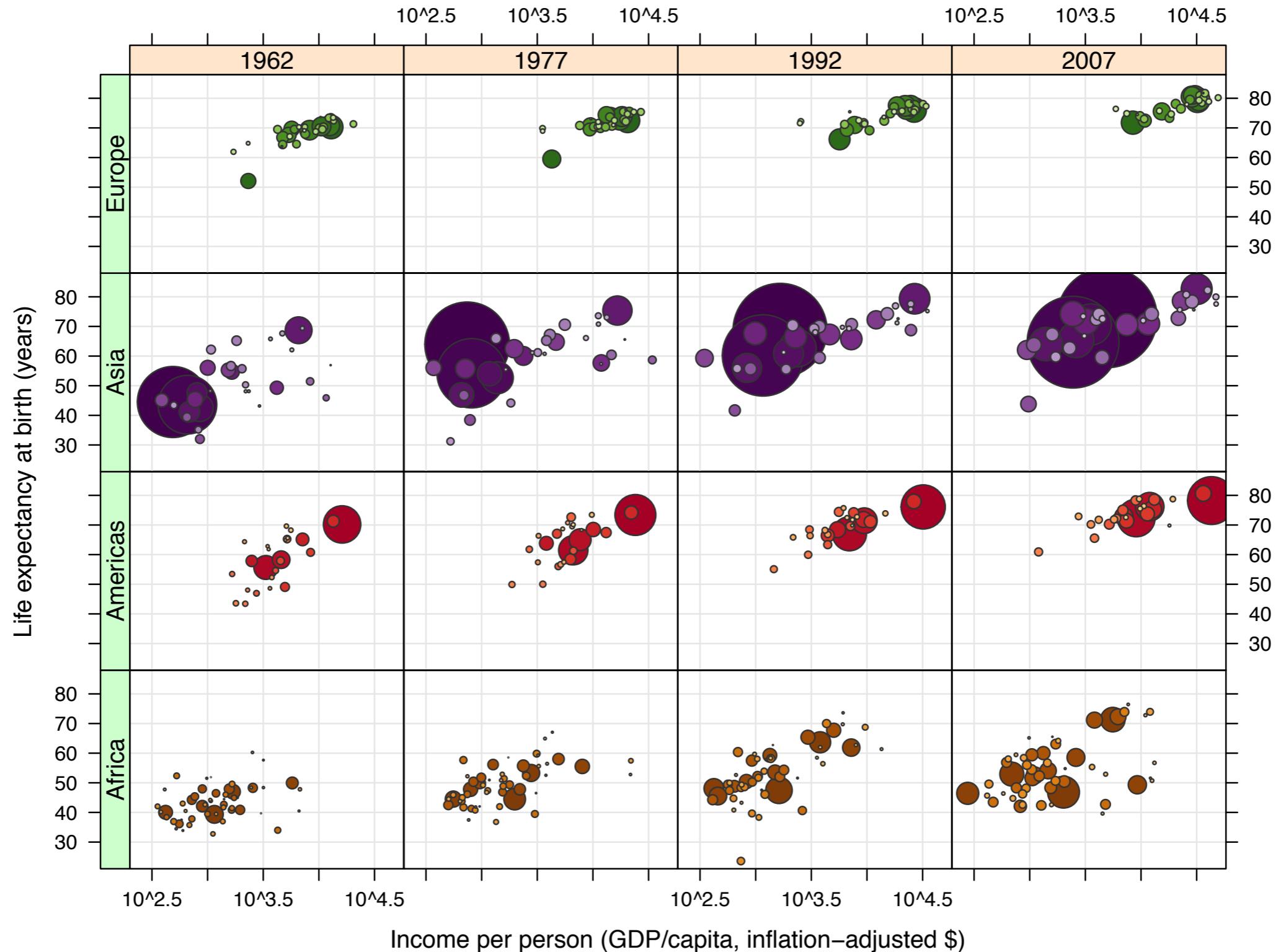
**grid.text()** puts the year in the background  
**panel.grid()** adds reference lines (used here for horizontal)  
**panel.abline** adds arbitrary lines (used here for custom vertical reflines)  
**panel.points()** draws the filled circles



```
xyplot(lifeExp ~ gdpPerCap | factor(year), ...)
```

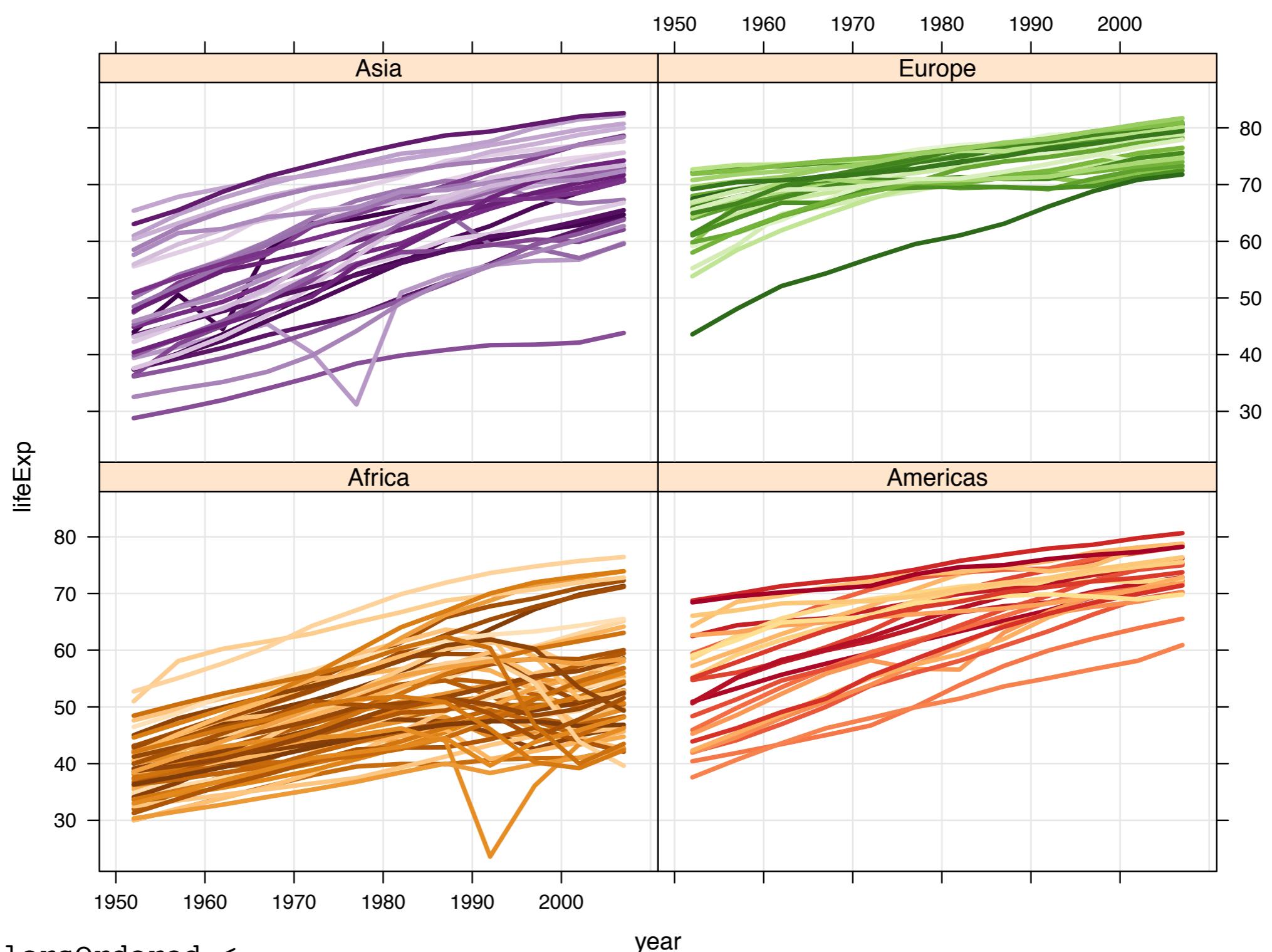


```
xyplot(lifeExp ~ gdpPercap | continent, ...)
```



```

library(latticeExtra)
jPlot <-
  xyplot(lifeExp ~ gdpPerCap | factor(year) * continent, ...)
useOuterStrips(jPlot)
  
```



```

countryColorsOrdered <-
  countryColors$color[match(levels(gDatOrdered$country),
                           countryColors$country)]

jGapminderPars <-
  list(superpose.line = list(alpha = 0.1, lwd = 3, col = countryColorsOrdered))

xyplot(lifeExp ~ year | continent, gDatOrdered,
       subset = continent != "Oceania", ylim = jYlim, type = c("l", "g"),
       groups = country, par.settings = jGapminderPars)

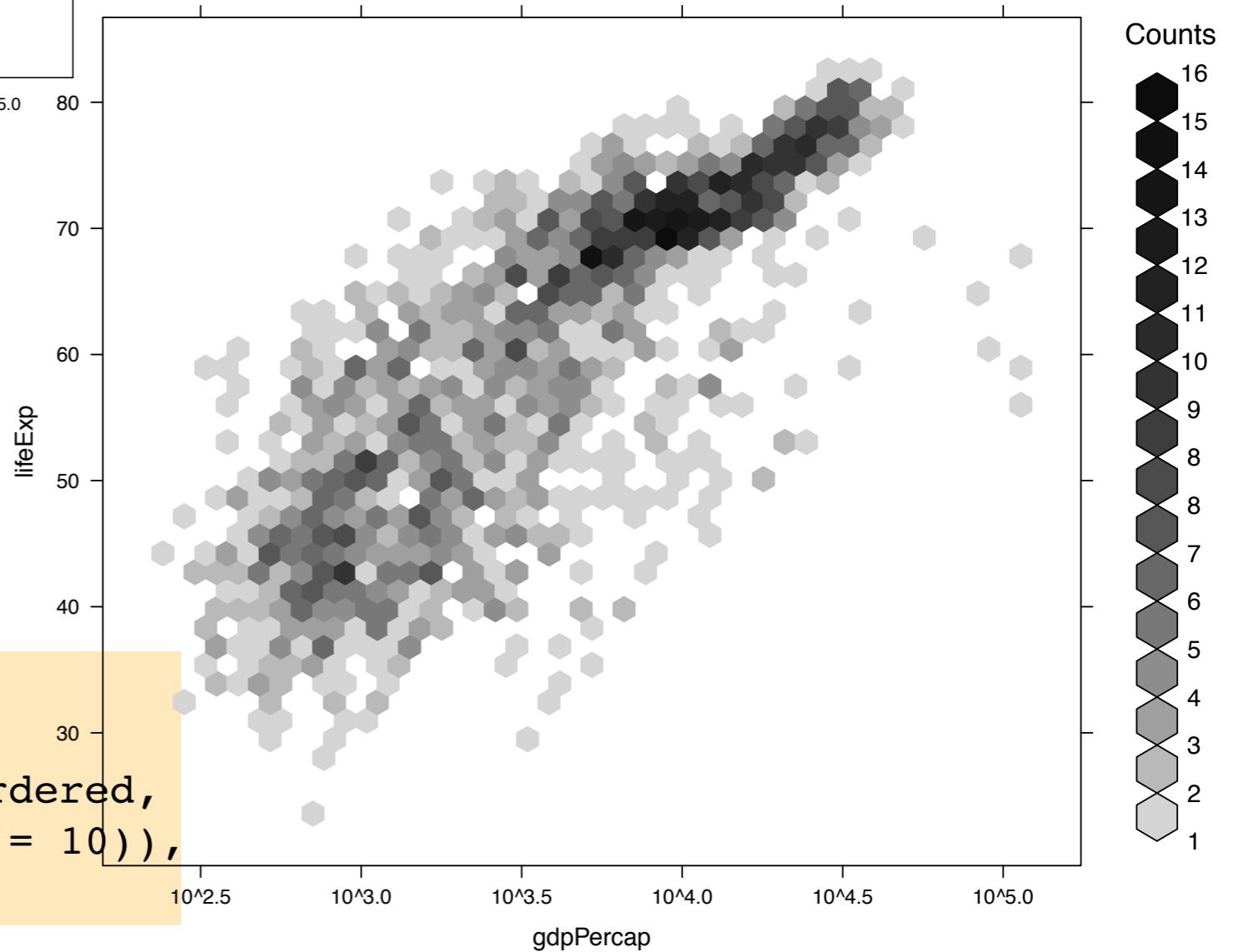
```

```
xyplot(lifeExp ~ gdpPerCap, gDatOrdered,  
scales = list(x = list(log = 10)))
```

For high volume  
scatterplots, consider  
hexagonal binning.

```
library(hexbin)
```

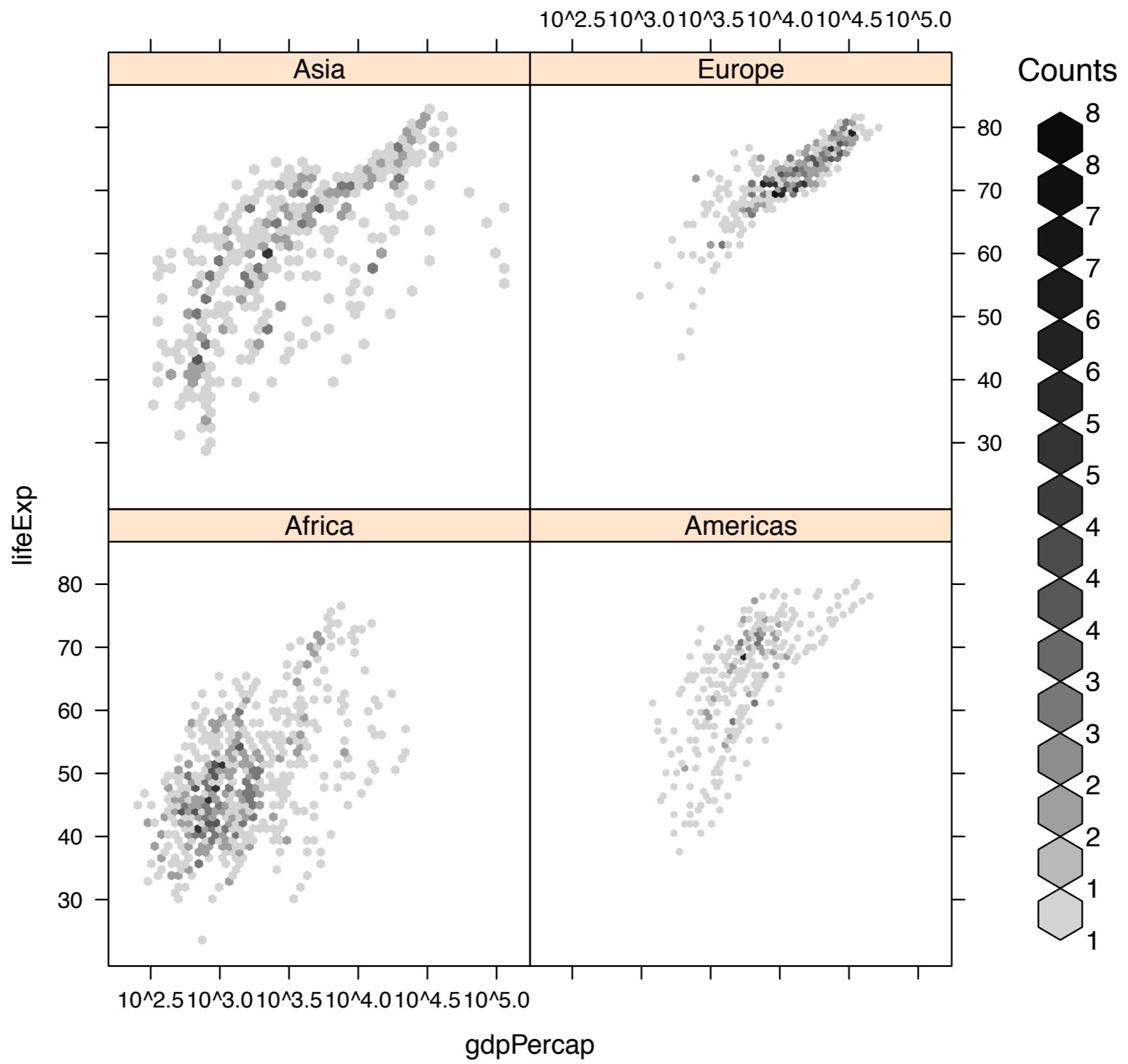
```
hexbinplot(lifeExp ~ gdpPerCap, gDatOrdered,  
scales = list(x = list(log = 10))),  
xbins = 40)
```



“Hexagonal binning”

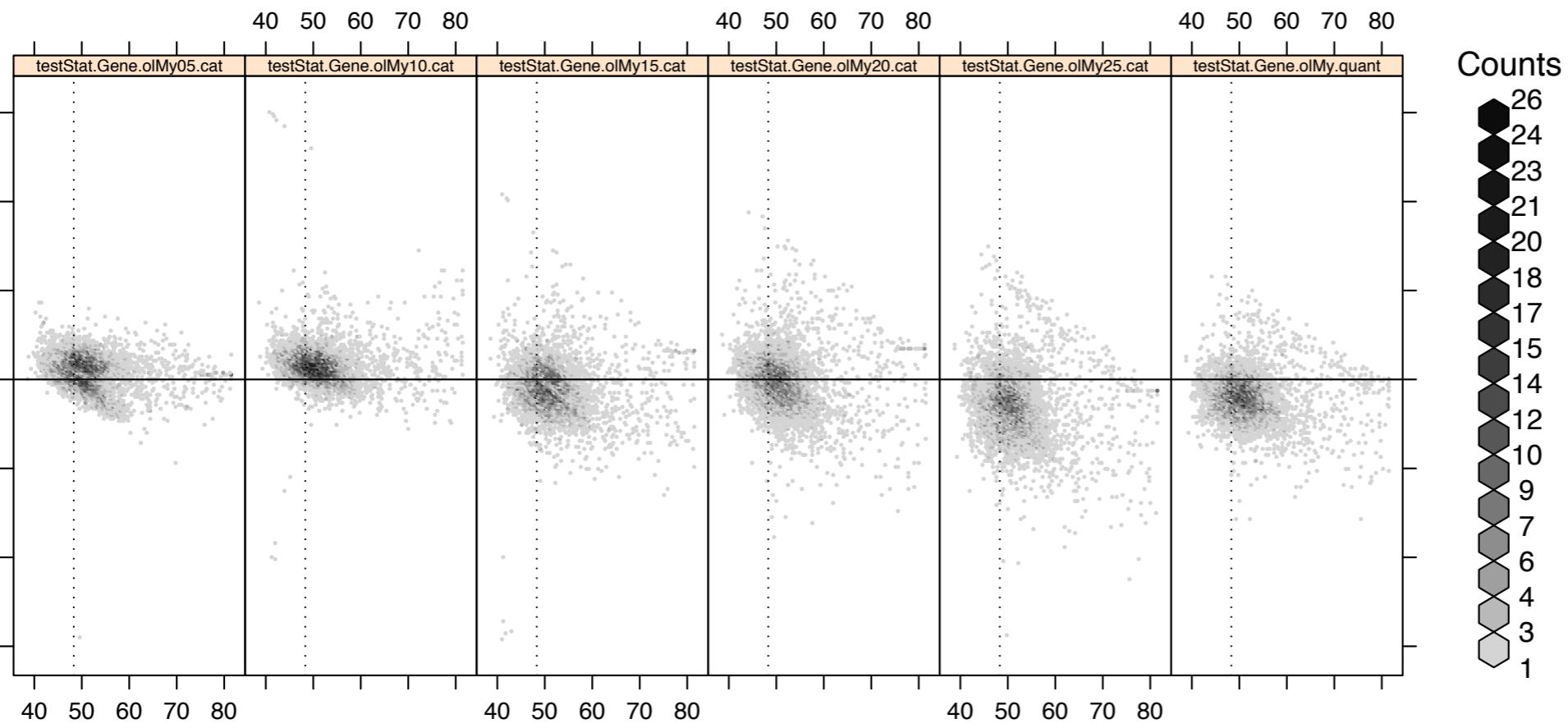
Superior to scatterplot, when number of points creates overplotting problem.

Idea: tile the plane with hexagons, shade according to relative frequency of points.

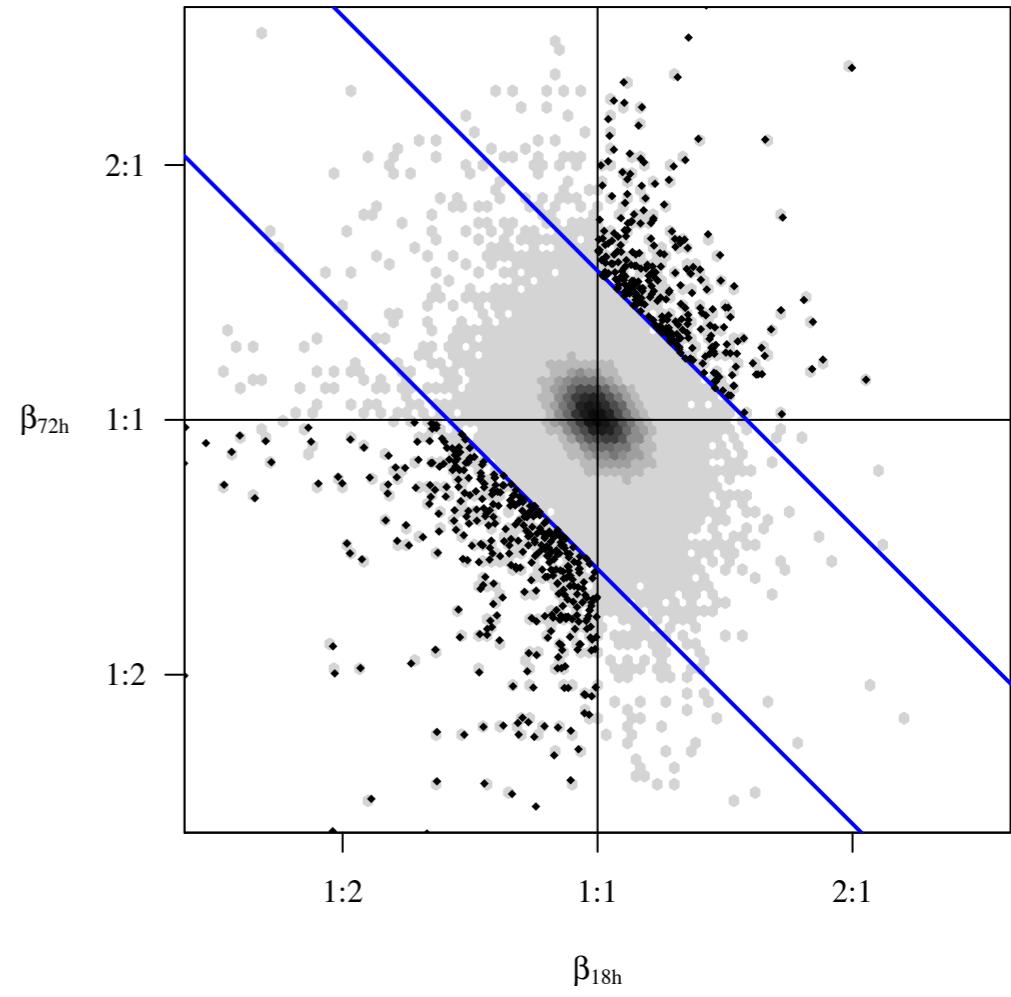


```
hexbinplot(lifeExp ~ gdpPercap | continent, gDatOrdered,  
subset = continent != "Oceania",  
scales = list(x = list(log = 10)),  
xbins = 40)
```

Test stat of an interaction term for deletion and oligomycin



L in condition conc0.0



Relevant R functions:

`hexbinplot()` and `hexplom()` from `hexbin` package

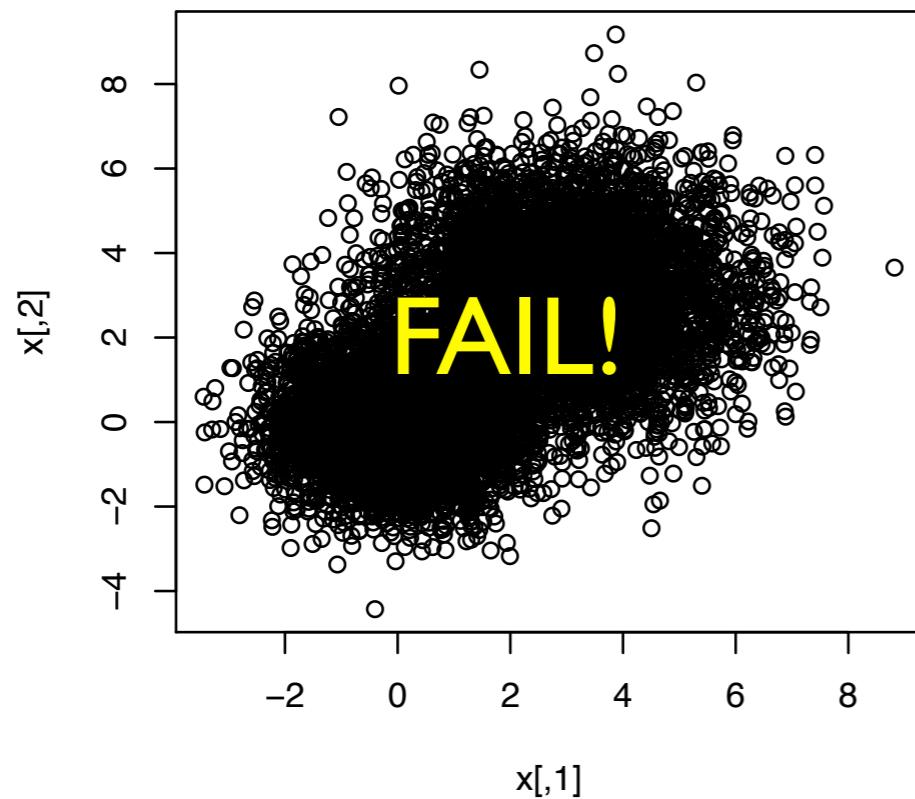
`smoothScatter()` -- also works well with `pairs()`

try running this:

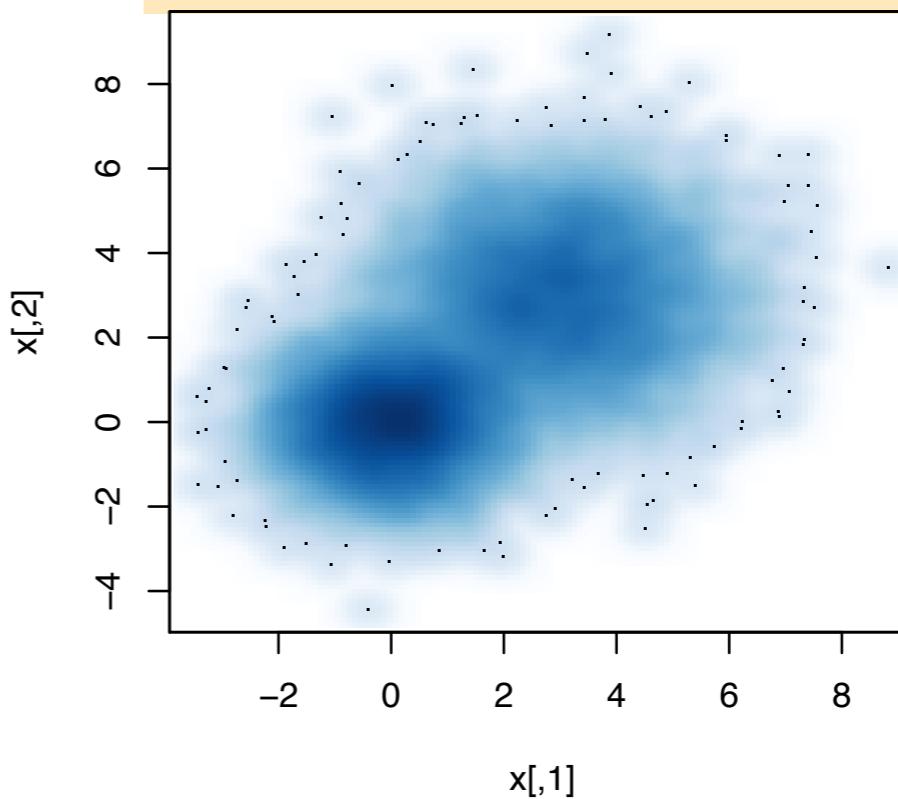
```
example(smoothScatter)
```

```
library(hexbin)  
example(hexbin)
```

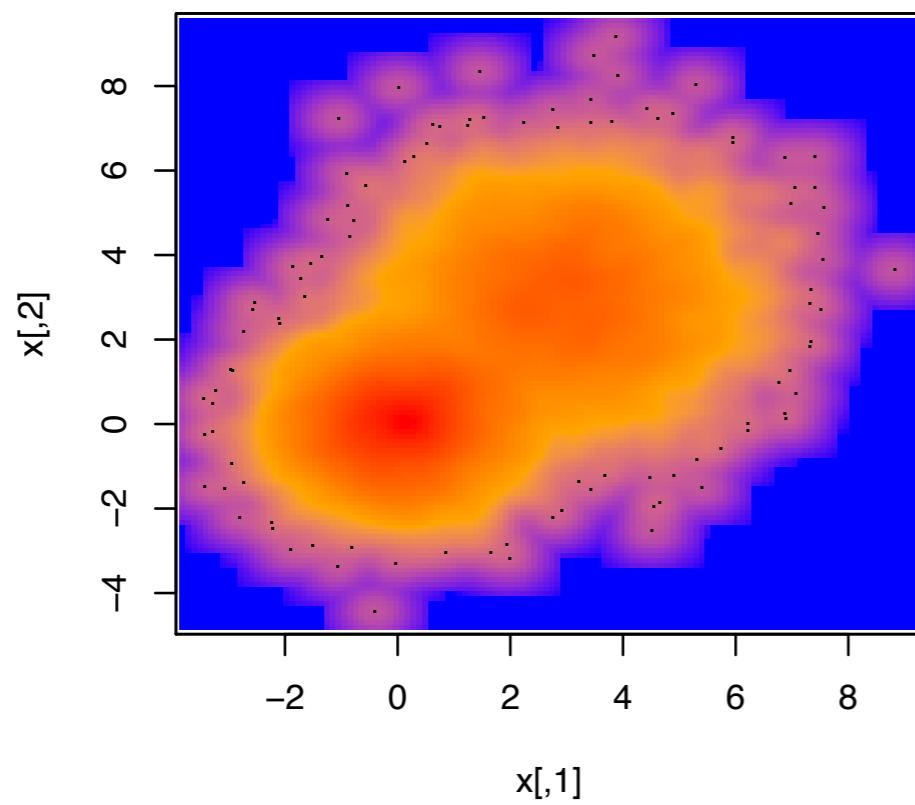
`plot(x)`



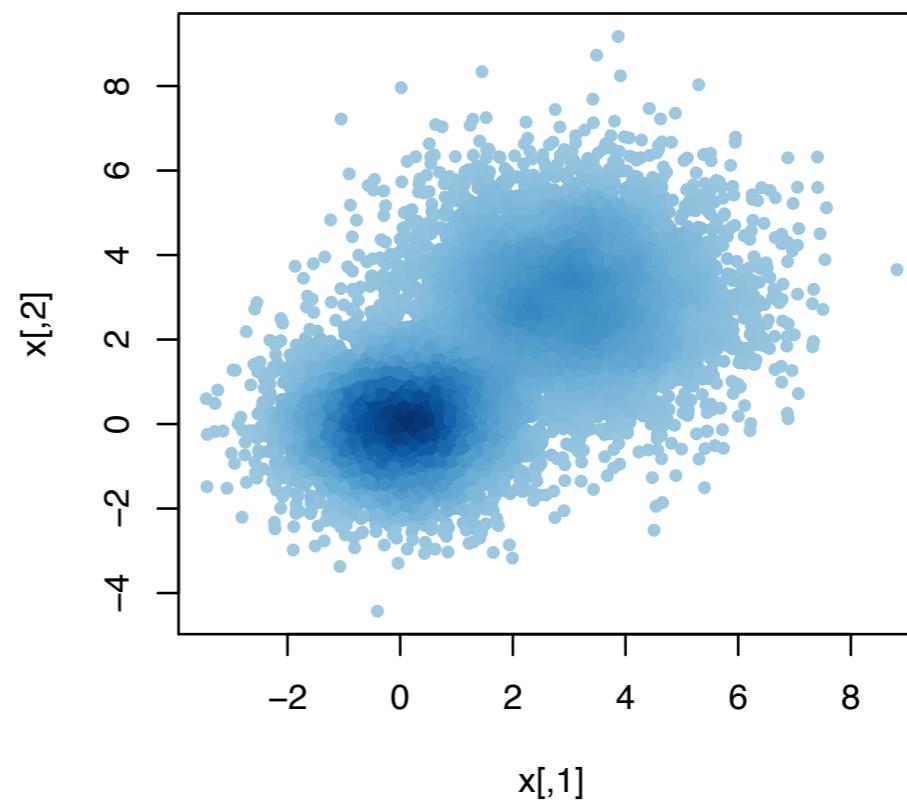
`smoothScatter(x)`



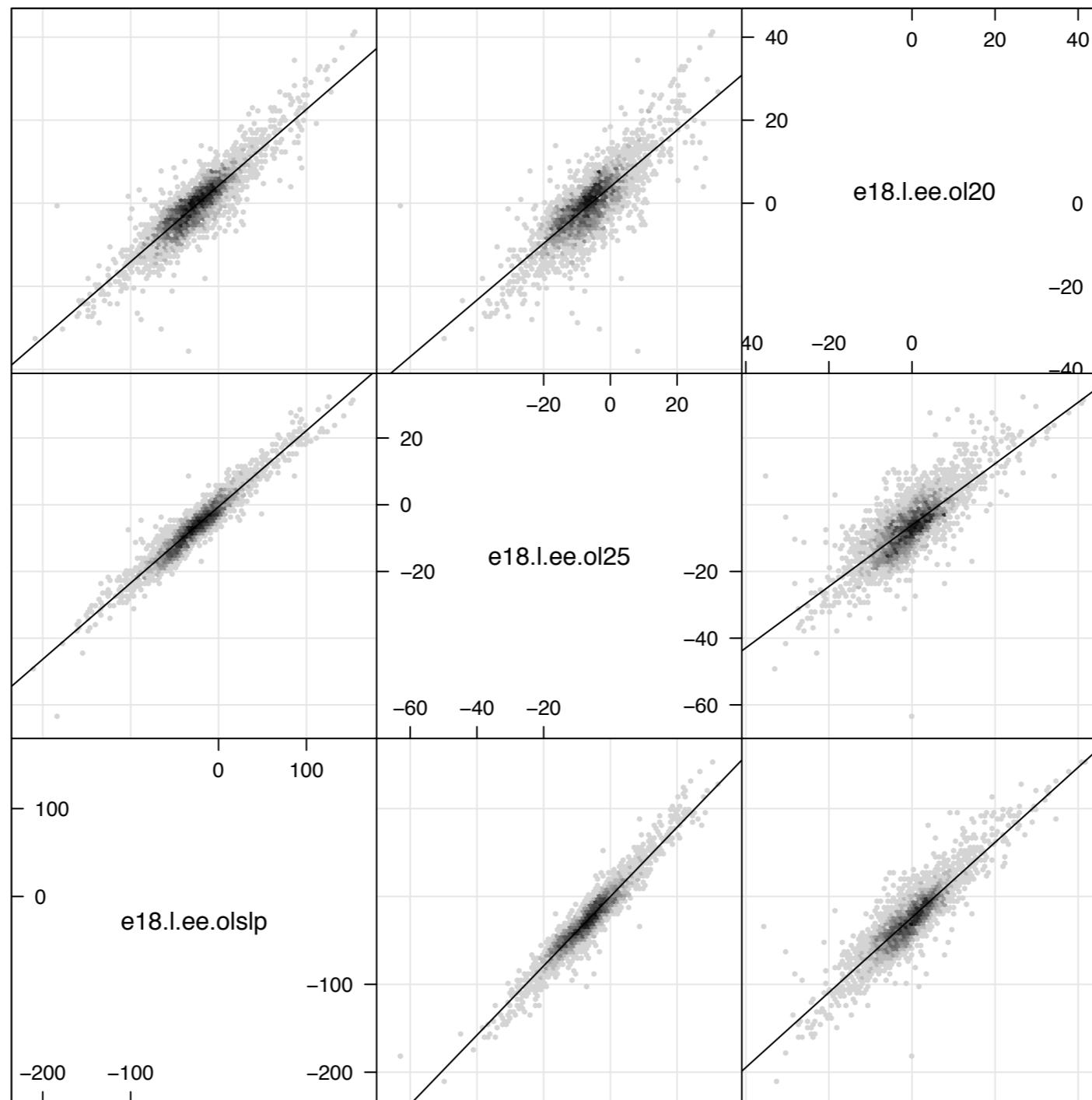
```
Lab.palette <- colorRampPalette(c("blue", "orange", "red"), space = "Lab")
smoothScatter(x, colramp = Lab.palette)
```



`plot(x, col = densCols(x), pch=20)`



### Exp 18, L: comparing effect estimates



Hexagonal binning, again, but in context of a scatterplot matrix. Read about `pairs()`, `splom()` and `hexsplom()`.

```
pairs(y, panel = function(...) smoothScatter(..., nrpoints=0, add=TRUE))
```

