

Data Visualization in R

3. Grid & lattice graphics



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SCS Short Course
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<http://www.datavis.ca/courses/RGraphics/>



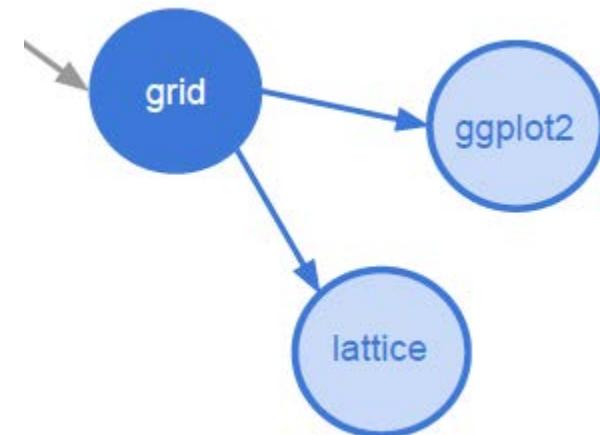
Overview

- Overview of grid-based graphics
 - `grid`: low-level graphics functions
 - `lattice`, `vcd`, `ggplot2`: high-level functions

The `grid` graphics system for R provides an alternative and more powerful way to develop data graphics in R

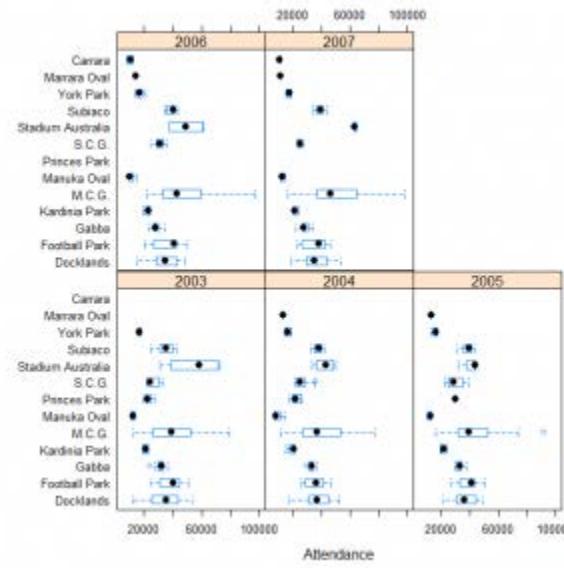
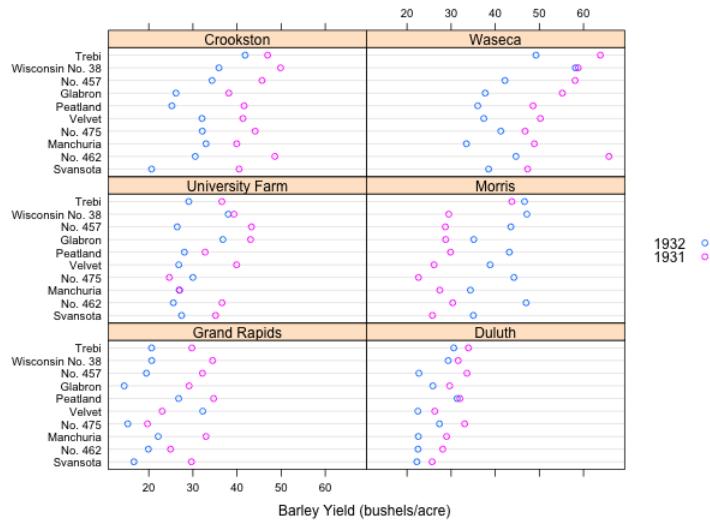
The `lattice` package, provides functions for drawing all standard plots, plus:

- more pleasing defaults
- create and modify graphic “themes”
- collections (“small multiples”) of simpler graphs from subsets of the data.



Lattice, son of Trellis graphics

- Complex multivariate data can often be better visualized by **conditioning & grouping**
 - show how some relationship changes over other variables
 - Tufte: “small multiples”: separate panels, arranged for visual comparison
 - Cleveland et al.: Trellis graphs for S+, ~ 1980
 - Deepayan Sarkar: lattice package, ~ 2000



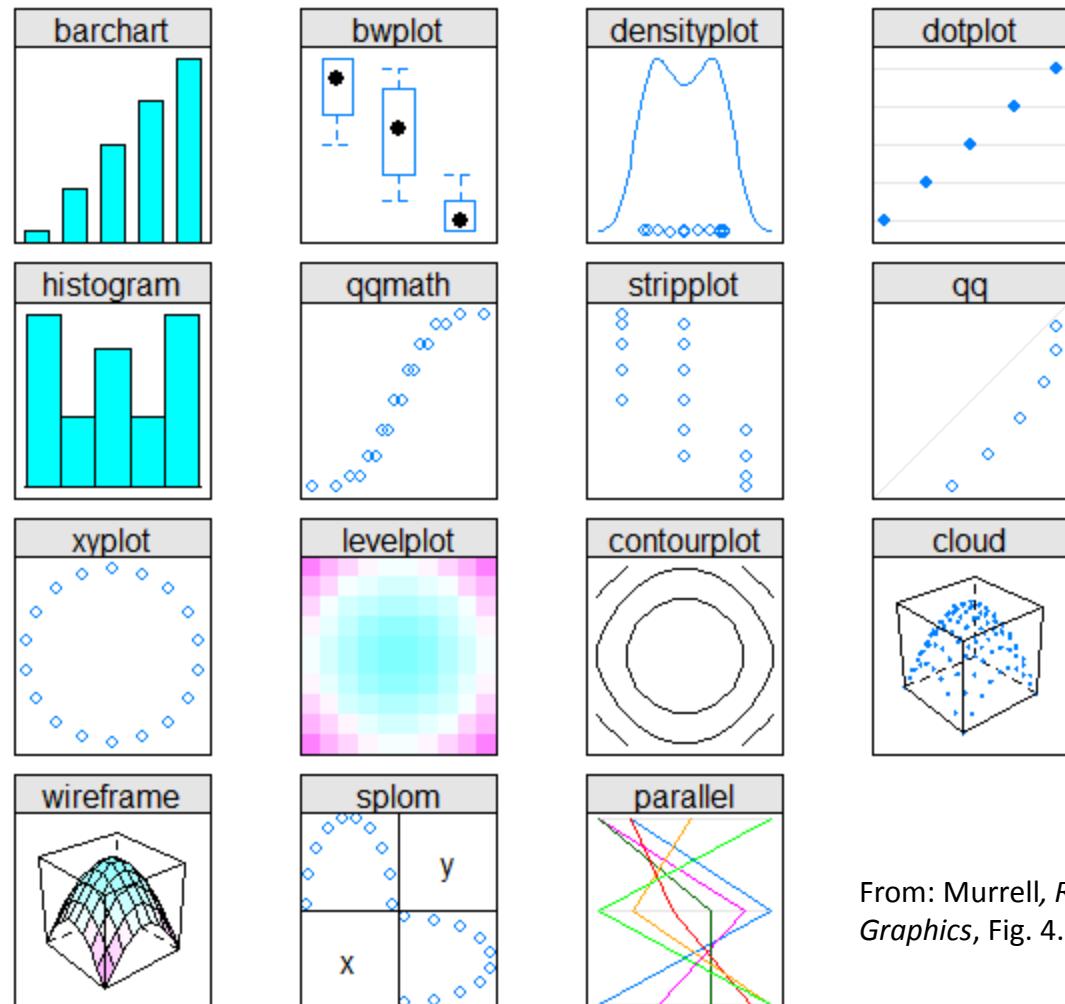
Lattice ideas in a nutshell

- All plots can be described by plot formulas
 - $\sim y$ Some univariate plot (boxplot, histogram, boxplot, ...)
 - $\sim y | A$ Univariate, separate panels for levels of factor A
 - $\sim y | z$ Univariate, cutting z into discrete ranges
 - $y \sim x$ Bivariate
 - $y \sim x | A$ Bivariate, separate panels for levels of A
 - $y \sim x | A + B$ multiple conditioning variables
 - $y_1 + y_2 \sim x_1 + x_2$ multiple Y and X variables
- Conditioning variables define “panels” in a plot
 - These can be laid out on a “page” in various ways
 - **panel functions** get the data for a subset and “render” (plot) it
 - High-level functions handle panel layout, and call panel functions
- Customize
 - graphic “themes” generalize `par()` settings
 - Combine multiple panel functions, write new ones.

Lattice plot functions

These are the high-level plot functions in lattice

These schematic examples have all been “rendered” using the default lattice theme settings



From: Murrell, R
Graphics, Fig. 4.3

Lattice plot functions

Lattice plots have analogs in traditional graphics

All use formula-style arguments for what to plot: $y \sim x$, or conditioning: $y \sim x|A$

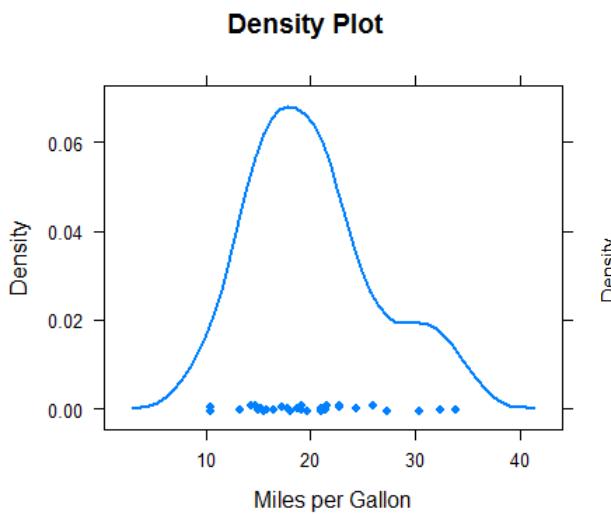
lattice function	description	formula examples	base analog
<code>barchart()</code>	bar chart	$x \sim A$ or $A \sim x$	<code>barplot()</code>
<code>bwplot()</code>	boxplot	$x \sim A$ or $A \sim x$	<code>boxplot()</code>
<code>densityplot()</code>	kernal density plot	$\sim x A * B$	<code>plot.density()</code>
<code>dotplot()</code>	dotplot	$\sim x A$	<code>dotchart()</code>
<code>histogram()</code>	histogram	$\sim x$	<code>hist()</code>
<code>stripplot()</code>	strip plots	$A \sim x$ or $x \sim A$	<code>stripchart()</code>
<code>xyplot()</code>	scatterplot	$y \sim x A$	<code>plot()</code>
<code>contourplot()</code>	3D contour plot	$z \sim x * y$	<code>contour()</code>
<code>cloud()</code>	3D scatterplot	$z \sim x * y A$	NA
<code>levelplot()</code>	3D level plot	$z \sim y * x$	<code>image()</code>
<code>parallel()</code>	parallel coordinates plot	data frame	NA
<code>splom()</code>	scatterplot matrix	data frame	<code>pairs()</code>
<code>wireframe()</code>	3D surface graph	$z \sim y * x$	<code>persp()</code>

Lattice plots: formulas, conditioning & grouping

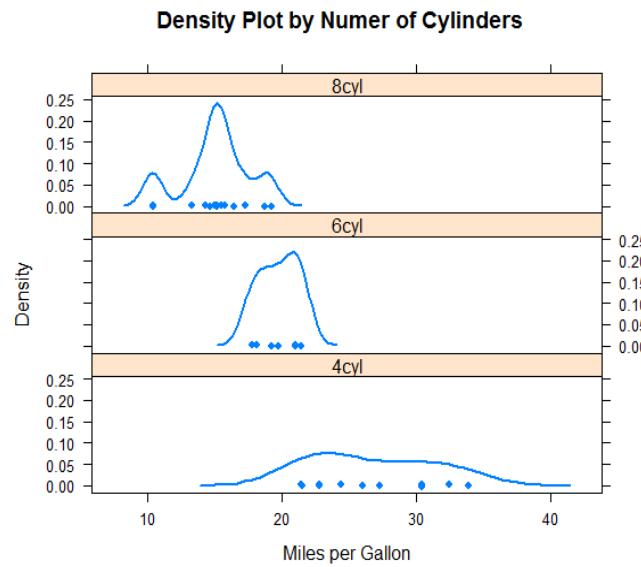
For 1D plots, the formula argument, $\sim y$, specifies the variable to be plotted

- Conditioning: $\sim y \mid \text{group}$ gives **multipanel** plots for the levels of the group factor
- Grouping: $\sim y$, $\text{group} =$ **superposes** plots for the levels of group

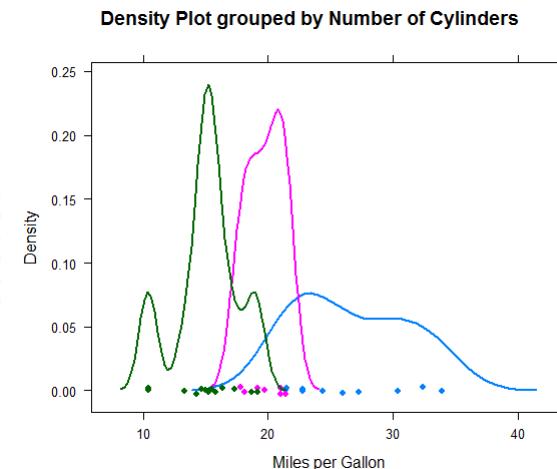
```
densityplot(~mpg,  
           data=mtcars, ...)
```



```
densityplot(~mpg | cyl,  
           data=mtcars, ...)
```



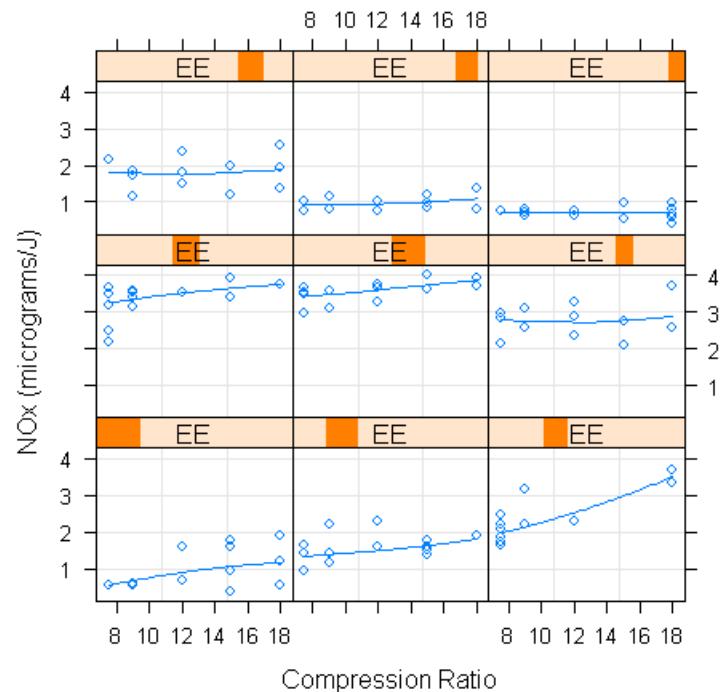
```
densityplot(~mpg, groups=cyl,  
           data=mtcars, ...)
```



Ethanol data: Ethanol fuel was burned in a single-cylinder engine.

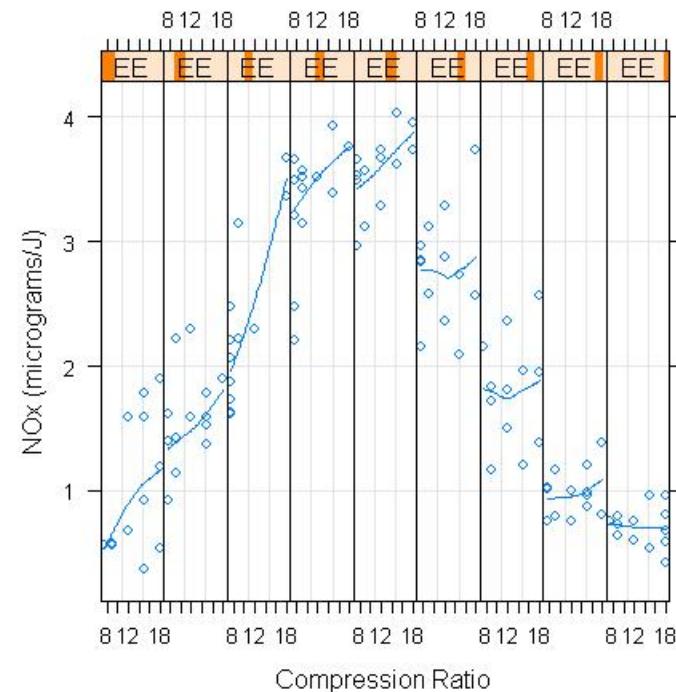
How do emissions of nitrous oxide (NO_x) depend on

- engine compression ratio(C) and
- equivalence ratio (EE), a measure of richness of the air and ethanol fuel mixture



`xyplot()` for lattice scatterplots:

```
xyplot(NOx ~ C | EE, data = ethanol, ...)
```

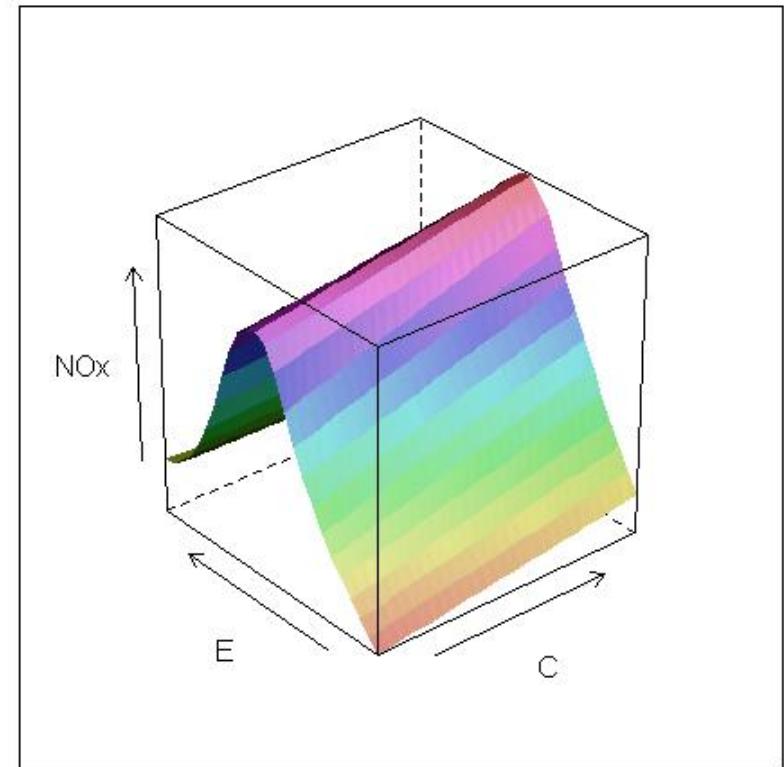


Same plot, with `aspect="xy"`: sets aspect ratio to “bank to 45°”

As in base graphics, some computation is often required to make a simpler or better version of some plot.

- 2D plots of the ethanol data suggest something that might better be seen in 3D
- This requires calculating a fitted response surface, and drawing it
- It doesn't show the data, and uses a non-parametric smoother, not a `lm()` model

```
require(stats)
with(ethanol, {
  eth.lo <- loess(NOx ~ C * E, span = 1/3, parametric = "C",
                  drop.square = "C", family="symmetric")
  eth.marginal <- list(C = seq(min(C), max(C), length.out = 25),
                        E = seq(min(E), max(E), length.out = 25))
  eth.grid <- expand.grid(eth.marginal)
  eth.fit <- predict(eth.lo, eth.grid)
  wireframe(eth.fit ~ eth.grid$C * eth.grid$E,
            shade=TRUE,
            screen = list(z = 40, x = -60, y=0),
            distance = .1,
            xlab = "C", ylab = "E", zlab = "NOx")
})
```



This example is complex. It uses:

- `loess()` to calculate smoothed values of NOx
- `predict()` to evaluate these over ranges of C & E
- `wireframe()` to plot these with nice shading

If this plot is believed, it gives a much simpler description of dependence, $\text{NOx} \sim \text{C} * \text{E}$

Detour: Modeling what we see

- Graphs of the ethanol data suggest a systematic, but complex relationship between $\text{NOx} \sim C + E$
 - Traditional parametric linear models handle this very semi-well
 - E.g., try a model with terms in C, E, E^2 and interactions

```
> eth.mod2 <- lm(NOx ~ (C + poly(E, 2))^2, data=ethanol)
> Anova(eth.mod2)
Anova Table (Type II tests)

Response: NOx
          Sum Sq Df  F value    Pr(>F)
C           5.032  1 25.2282 2.925e-06 ***
poly(E, 2) 91.838  2 230.2103 < 2.2e-16 ***
C:poly(E, 2) 3.322  2    8.3271 0.0005101 ***
Residuals   16.356 82
---
Signif. codes:  0 '****' 0.001 '***' 0.01 '**' 0.05 '*' 0.1 '.' 1
```

Detour: Modeling what we see

- The R model formula, $\text{NOx} \sim (\text{C} + \text{poly}(\text{E}, 2))^2$ is a short-hand notation
- The expanded version is nearly
$$\text{NOx} \sim \text{C} + \text{E} + \text{E}^2 + \text{C:E} + \text{C:E}^2$$
- Interpretation:
 - $\text{C} + \text{E}$: overall linear effects (slopes) of C & E on NOx
 - E^2 : quadratic effect (curvature) of equivalence ratio on NOx
 - C:E – does the slope for E change linearly with C?
 - C:E^2 – does the curvature for E change linearly with C

Detour: Modeling what we see

summary() for a given model gives significance tests of model terms

```
> summary(eth.mod2)

Call:
lm(formula = NOx ~ (C + poly(E, 2))^2, data = ethanol)

Residuals:
    Min      1Q  Median      3Q     Max 
-0.84489 -0.37039 -0.00367  0.39327  0.76796 

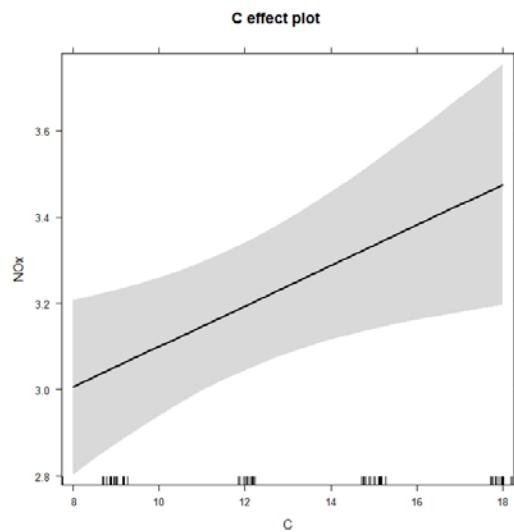
Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept)  1.16206   0.16009   7.259 2.01e-10 ***
C             0.06572   0.01265   5.193 1.48e-06 ***
poly(E, 2)1  4.81844   1.56979   3.069 0.002907 **  
poly(E, 2)2 -12.15328   1.61916  -7.506 6.60e-11 ***
C:poly(E, 2)1 -0.46307   0.11615  -3.987 0.000145 *** 
C:poly(E, 2)2  0.15492   0.11720   1.322 0.189909    
---
Signif. codes:  0 '****' 0.001 '***' 0.01 '**' 0.05 '*' 0.1 '.' 1

Residual standard error: 0.4466 on 82 degrees of freedom
Multiple R-squared:  0.8535,    Adjusted R-squared:  0.8445 
F-statistic: 95.52 on 5 and 82 DF,  p-value: < 2.2e-16
```

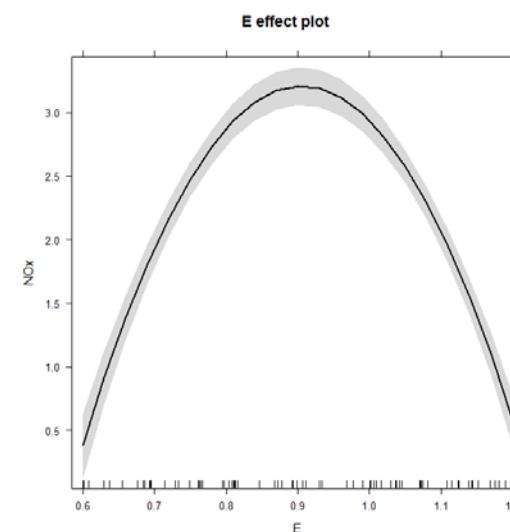
Effect plots: Seeing what we model

In many cases, effect plots help to visualize a fitted model. These all use lattice graphics to render the plot.

```
plot(Effect("C", eth.mod2))
```



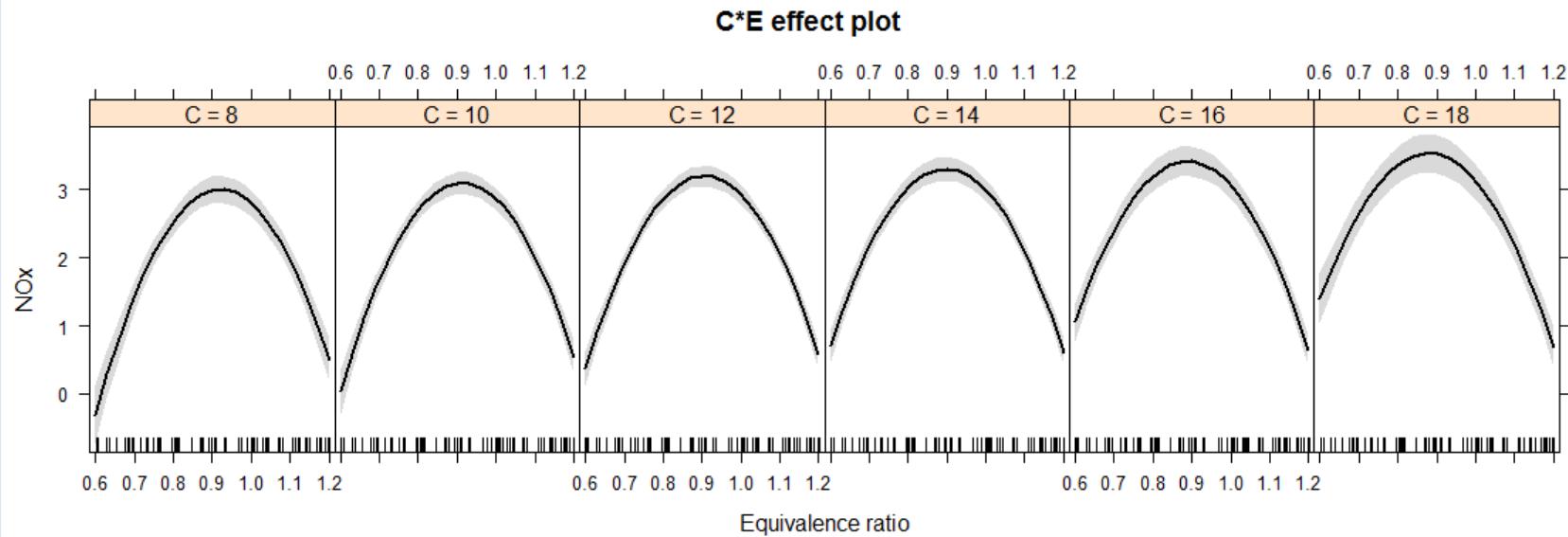
```
plot(Effect("E", eth.mod2))
```



Effect plots: Seeing what we model

The strength of lattice graphics for conditioning is used in multipanel effect plots
Details of the layout and conditioning levels can all be controlled by options.

```
plot(allEffects(eth.mod2), layout=c(6,1), xlab="Equivalence ratio")
```



Detour: gam

- Generalized additive models (gam) are like generalized linear models (glm), but allow non-parametric “smoothed” $s()$ terms
 - degree of smoothing \sim # degrees of freedom
 - models can have linear & smoothed $s()$ terms
 - approx. significance tests are available for smooth terms

```
> library(mgcv)
> eth.gam1 <- gam(NOx ~ C + s(E), data=ethanol)
> summary(eth.gam1)

Parametric coefficients:
              Estimate Std. Error t value Pr(>|t|)    
(Intercept) 1.291342   0.088898 14.526 < 2e-16 ***
C            0.055345   0.007062  7.837 1.88e-11 ***
---
Approximate significance of smooth terms:
        edf Ref.df      F p-value    
s(E) 7.553 8.469 208.8 <2e-16 ***
---
R-sq.(adj) =  0.953  Deviance explained = 95.8%
GCV = 0.067206  Scale est. = 0.05991 n = 88
```

This is sometimes called “semi-parametric regression”

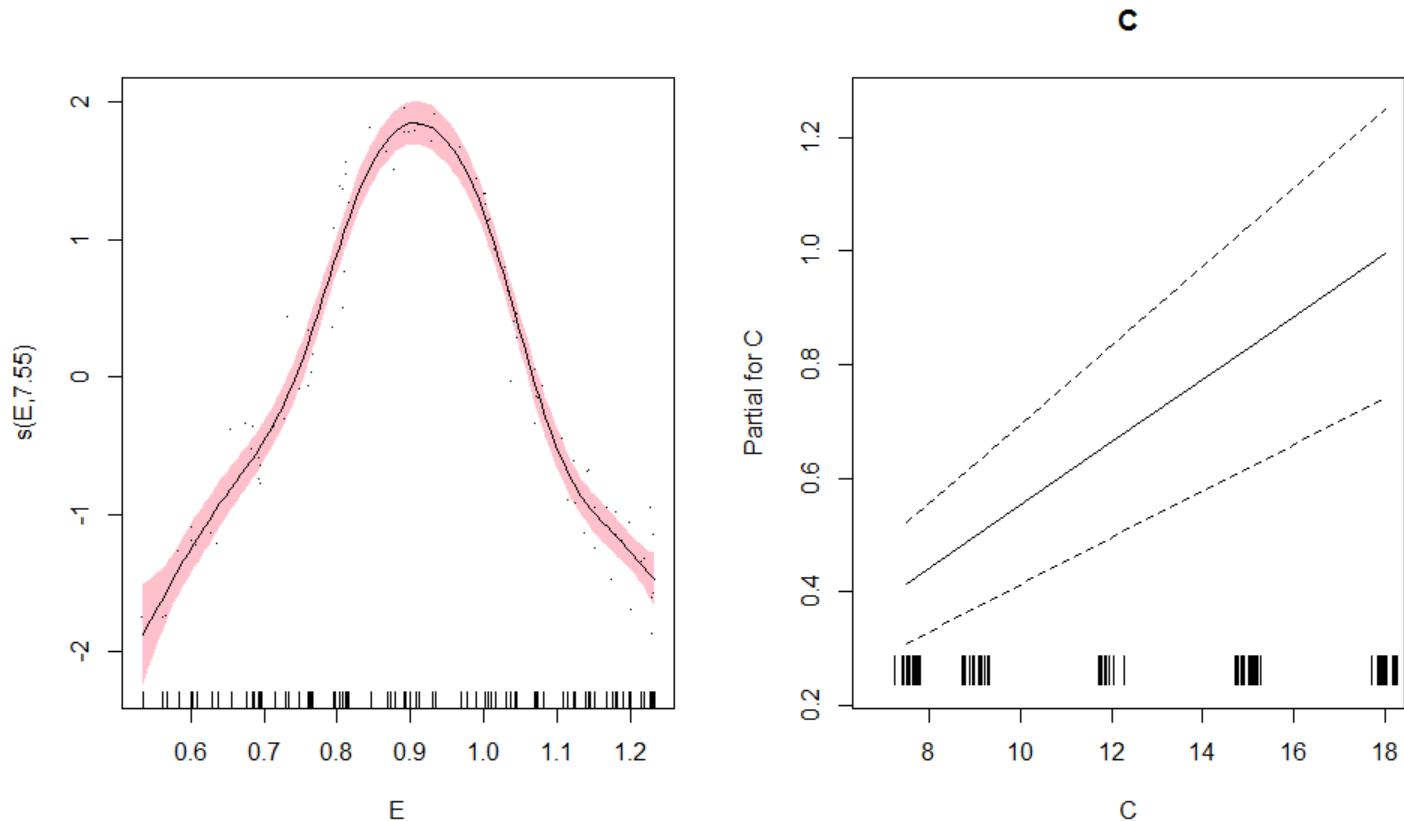
The edf for the smoothed term is found using cross-validation

There are other kinds of smoothing models

Plotting “gam” objects

The mgcv package contains a number of plot methods for “gam” objects

```
plot(eth.gam1, shade=TRUE, shade.col="pink", all.terms=TRUE, residuals=TRUE, pages=1)
```

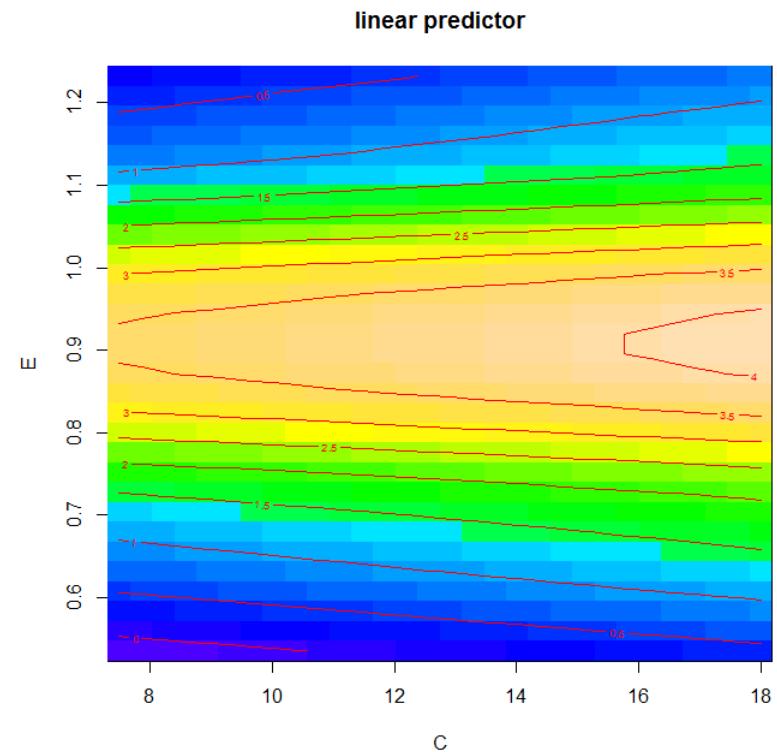
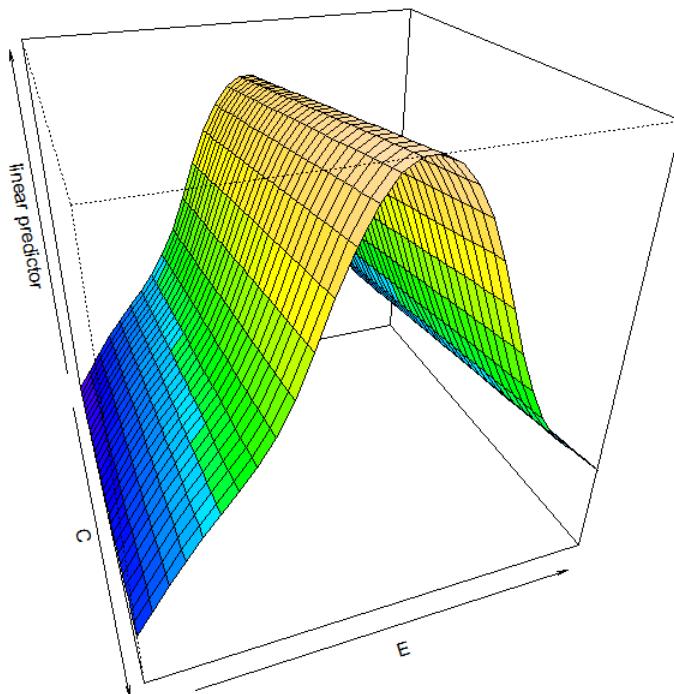


Plotting “gam” objects

`vis.gam()` is like a 3D version of an effect plot. It shows the fitted values for two predictors, holding constant all others.

```
vis.gam(eth.gam1, color="topo", theta=70, phi=30)  
vis.gam(eth.gam1, color="topo", plot.type="contour")
```

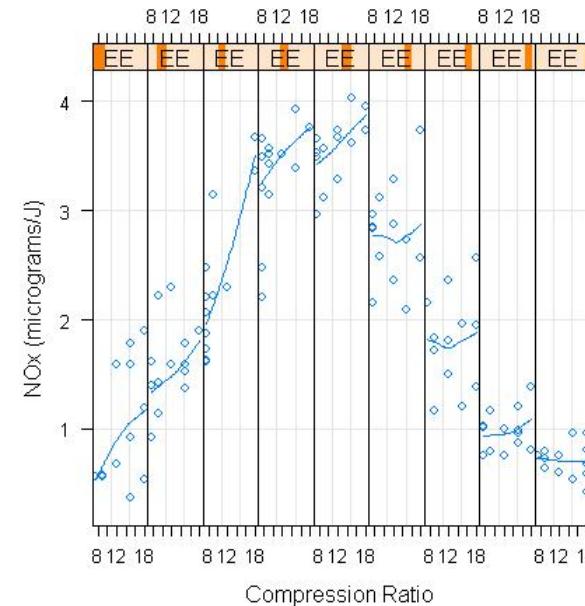
NB: The result is similar to what we got using `loess()`. However, this is a full-fledged statistical model, so we can find confidence intervals, prediction intervals, etc.



Lattice panel functions

- Lattice plots use **panel functions** to add info to a plot
 - `panel.grid()` – grid lines
 - `panel.xyplot(x, y, type=, ...)` – various types of (x, y) plots
 - `panel.lmline()` – add regression line
 - `panel.loess()` – smoothed loess curve
 - many others ...

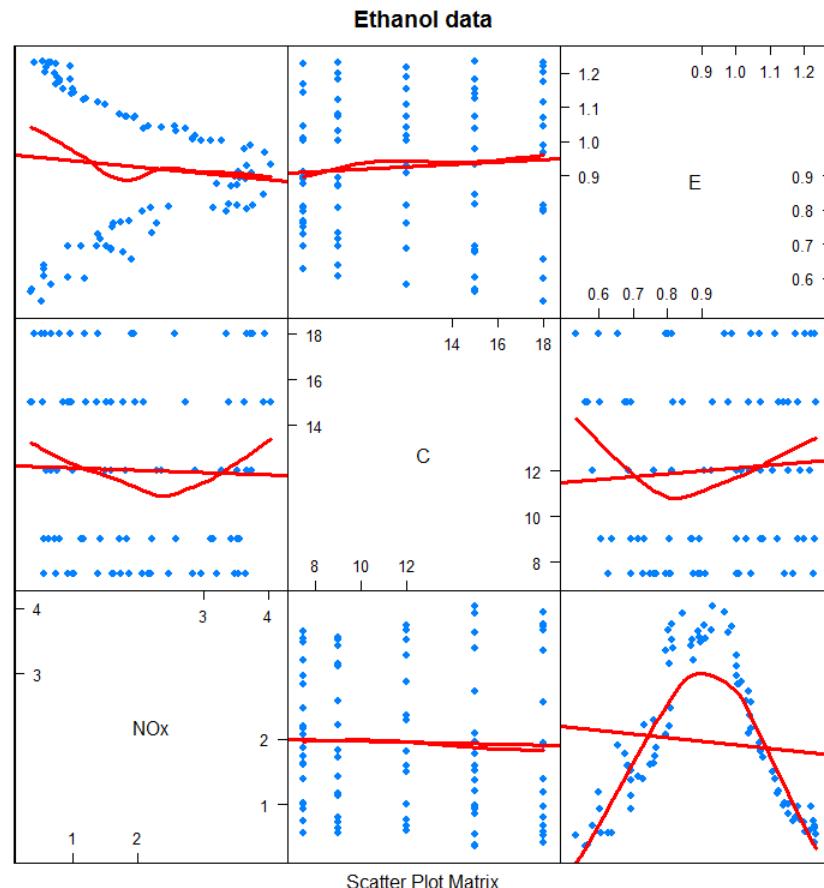
```
EE <- equal.count(ethanol$E, number=9, overlap=1/4)
xyplot(NOx ~ C | EE, data = ethanol,
       prepanel = function(x, y) prepanel.loess(x, y, span = 1),
       xlab = "Compression ratio", ylab = "NOx (micrograms/J)",
       panel = function(x, y) {
         panel.grid(h=-1, v= 2)
         panel.xyplot(x, y)
         panel.loess(x,y, span=1)
       },
       aspect = "xy")
```



splom()

splom() draws a scatterplot matrix. As with other lattice functions, a type= argument can be used to invoke several panel functions.

```
splom(ethanol,
  type=c("p", "r", "smooth"),
  col.line = "red",
  pch=16, lwd=3,
  main="Ethanol data")
```



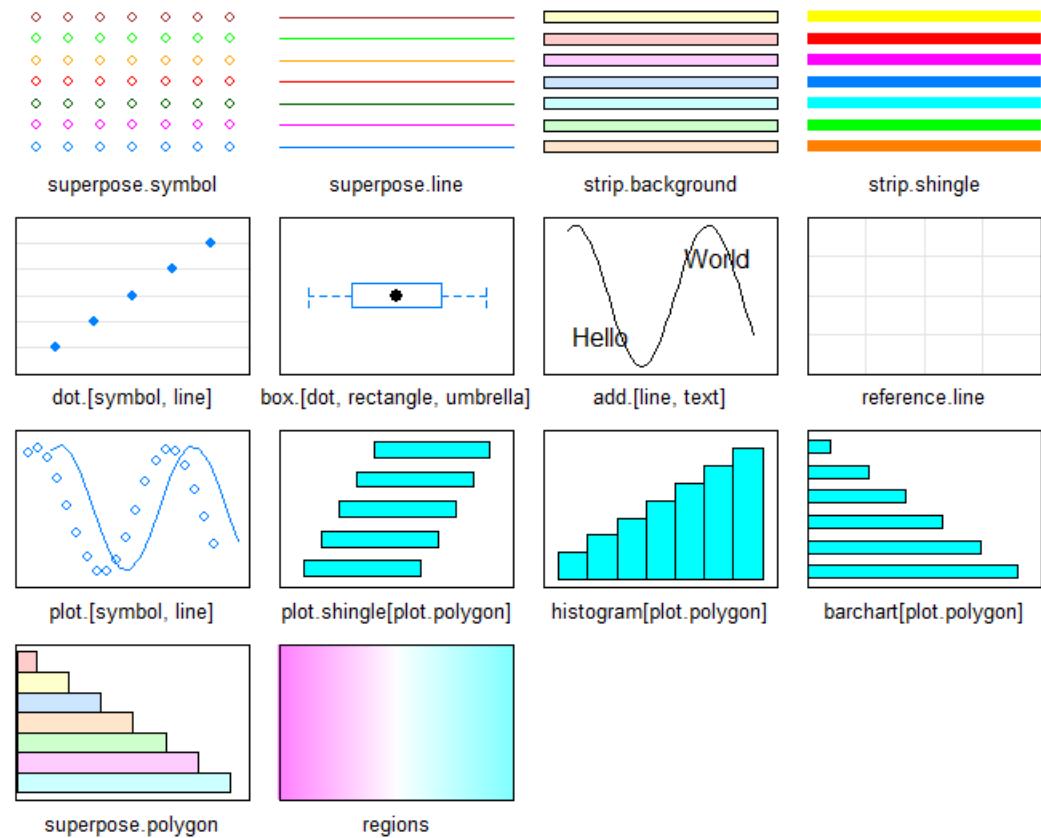
Lattice themes and settings

The Trellis approach allows creating effective graphs with a consistent look and feel. It uses “themes” to define colour, size and other features of components of a graph.

A theme consists of settings for the attributes of various graphical elements.

The current settings are displayed with `show.settings()`

This differs from base graphics, where `par()` settings are used inconsistently across different graph types



Lattice themes and settings

- Get theme settings with `trellis.par.get()`
- Set new ones with `trellis.par.set()`

```
> my.theme <- trellis.par.get()
> names(my.theme)
[1] "grid.pars"          "fontsize"           "background"        "panel.background"
[5] "clip"                "add.line"            "add.text"          "plot.polygon"
[9] "box.dot"             "box.rectangle"      "box.umbrella"     "dot.line"
[13] "dot.symbol"         "plot.line"          "plot.symbol"      "reference.line"
[17] "strip.background"   "strip.shingle"      "strip.border"    "superpose.line"
[21] "superpose.symbol"   "superpose.polygon" "regions"          "shade.colors"
[25] "axis.line"           "axis.text"          "axis.components" "layout.heights"
[29] "layout.widths"       "box.3d"              "par.xlab.text"   "par.ylab.text"
[33] "par.zlab.text"       "par.main.text"      "par.sub.text"
```

There are 35 different attributes, each of which is a list of more basic settings

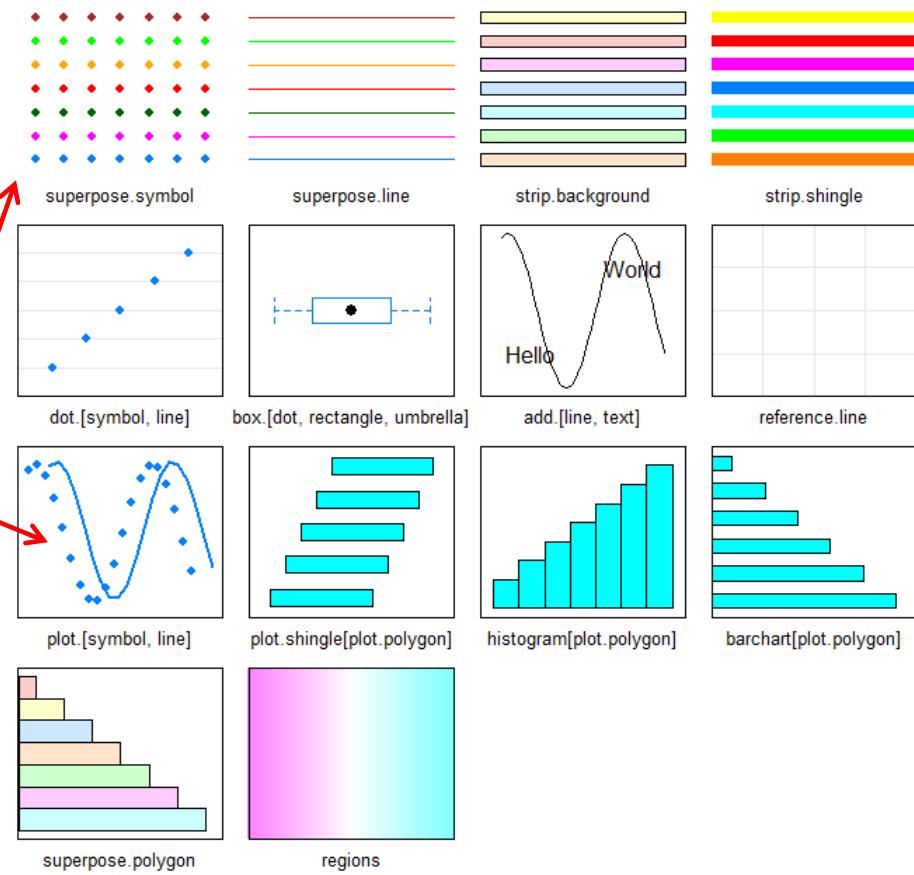
```
> names(my.theme$plot.symbol)
[1] "alpha" "cex"   "col"   "font"  "pch"   "fill"
> names(my.theme$plot.line)
[1] "alpha" "col"   "lty"   "lwd"
```

I like to use filled point symbols (pch=16) and make lines thicker

```
my.theme$plot.line$lwd <- 2  
my.theme$plot.symbol$pch <- 16  
my.theme$superpose.symbol$pch <- rep(16, 7)
```

```
#establish my.theme  
trellis.par.set(my.theme)  
show.settings()
```

points are now filled
circles & lines are
thicker



NB: This is tedious, but useful if you
are writing a paper or a book. Do it
ONCE, for all figures!

Lattice themes: color to BW

Lattice plots are “trellis” objects. They can be printed with different themes w/o changing your code

```
plt <- barchart(Class ~ Freq | Sex + Age,  
  data = as.data.frame(Titanic),  
  groups = Survived, stack = TRUE,  
  layout = c(4, 1),  
  auto.key = list(title = "Survived", columns = 2),  
  scales = list(x = "free"))  
print(plt)
```

Figure 2.9

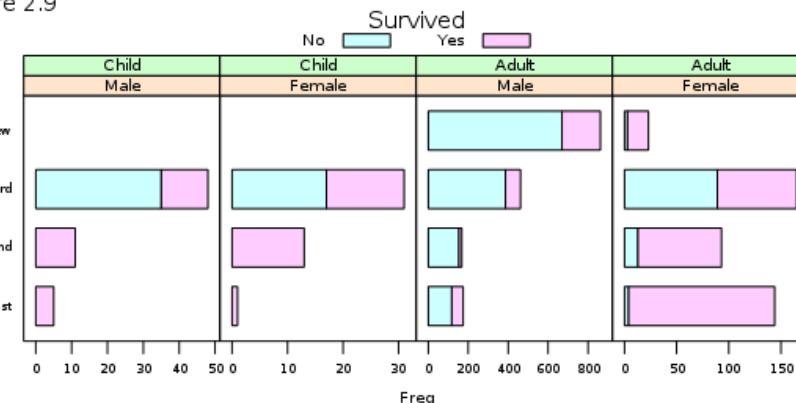
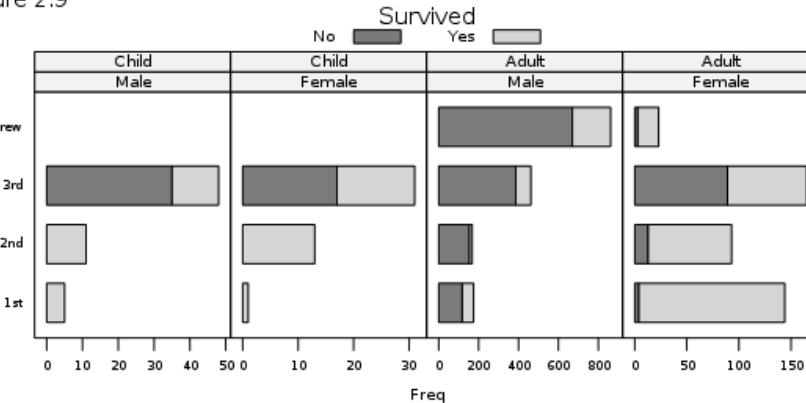


Figure 2.9



```
trellis.device(color = FALSE)  
print(plt)
```

As this example demonstrates, lattice themes are generally well-designed to handle color vs. B/W

Boxplots -> Violin plots

Boxplots show some aspects of the shape of distributions: median, IQR, outliers, ...

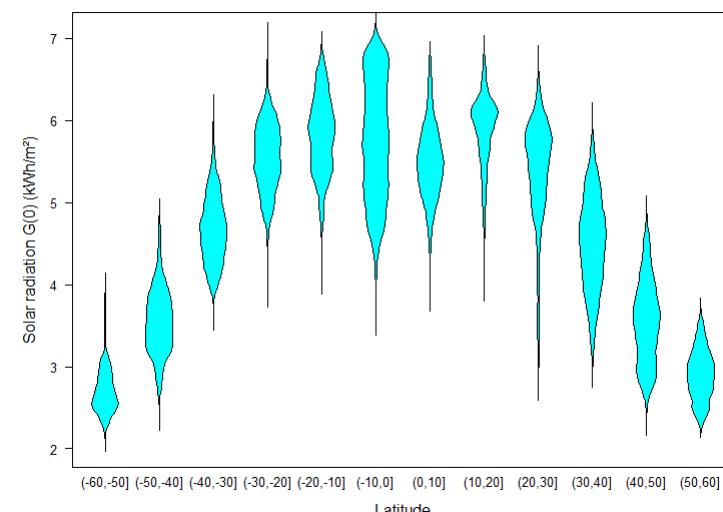
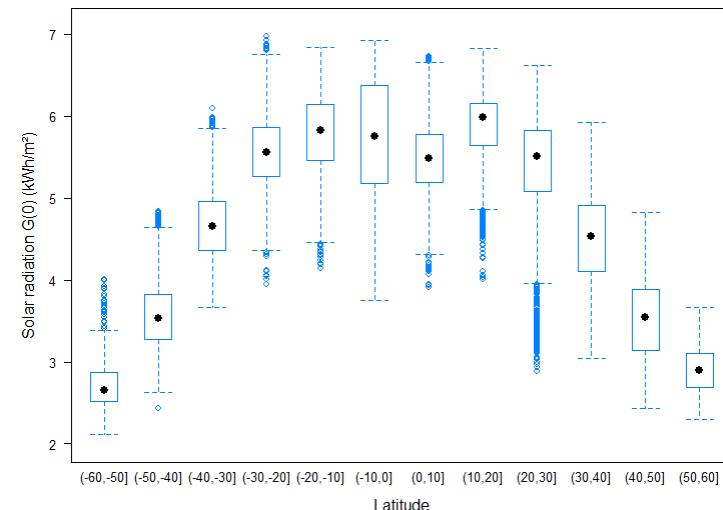
Violin plots use a mirrored kernel density plot instead

```
bwplot(Ann ~ cut(Lat, pretty(Lat, 20)),  
       data=nasa, subset=(abs(Lat)<60),  
       xlab='Latitude', ylab='Solar radiation G(0) (kWh/m2)')
```

For lattice, this is just a boxplot using a different panel function: `panel.violin()`

```
bwplot(Ann ~ cut(Lat, pretty(Lat, 20)),  
       data=nasa, subset=(abs(Lat)<60),  
       xlab='Latitude', ylab='Solar radiation G(0) (kWh/m2)',  
       panel = panel.violin)
```

NASA data on average solar radiation by latitude



Example from: <https://www.r-bloggers.com/violin-and-boxplots-with-lattice-and-r/> by Oscar Perpiñán Lamigueiro

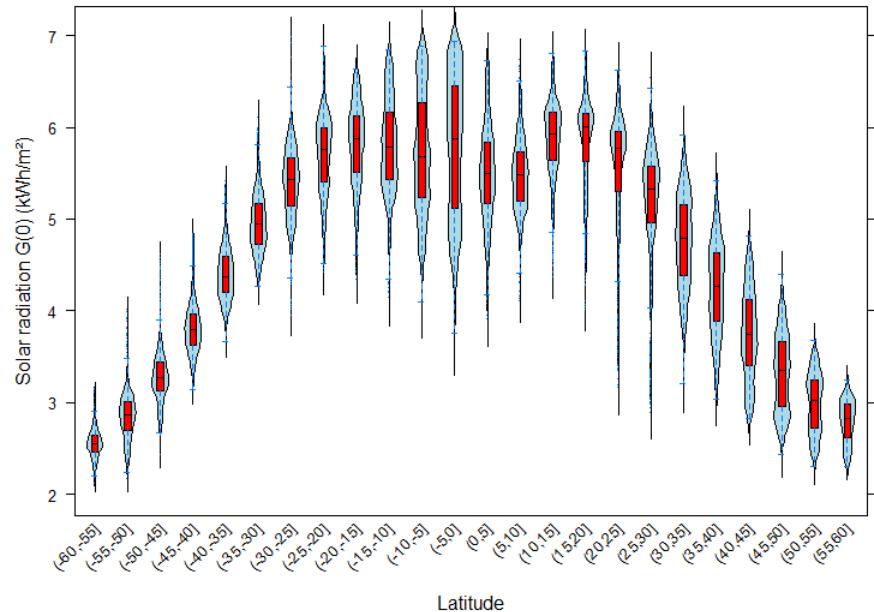
Custom panel functions

You can combine these using a custom panel function that calls both

```
my.panel <- function(..., box.ratio) {  
  panel.violin(..., col = "lightblue",  
    varwidth = FALSE, box.ratio = box.ratio)  
  panel.bwplot(..., col='black',  
    cex=0.9, pch='|', fill='red', box.ratio = .25)  
}
```

Use it:

```
bwplot(Ann ~ cut(Lat, pretty(Lat, 40)),  
  data=nasa, subset=(abs(Lat)<60),  
  xlab='Latitude', ylab='Solar radiation G(0) (kWh/m2)',  
  horizontal=FALSE,  
  panel = my.panel,  
  par.settings = list(box.rectangle=list(col='black'),  
    plot.symbol = list(pch='.', cex = 0.1)),  
  scales=list(x=list(rot=45, cex=0.5))  
)
```



Notes:

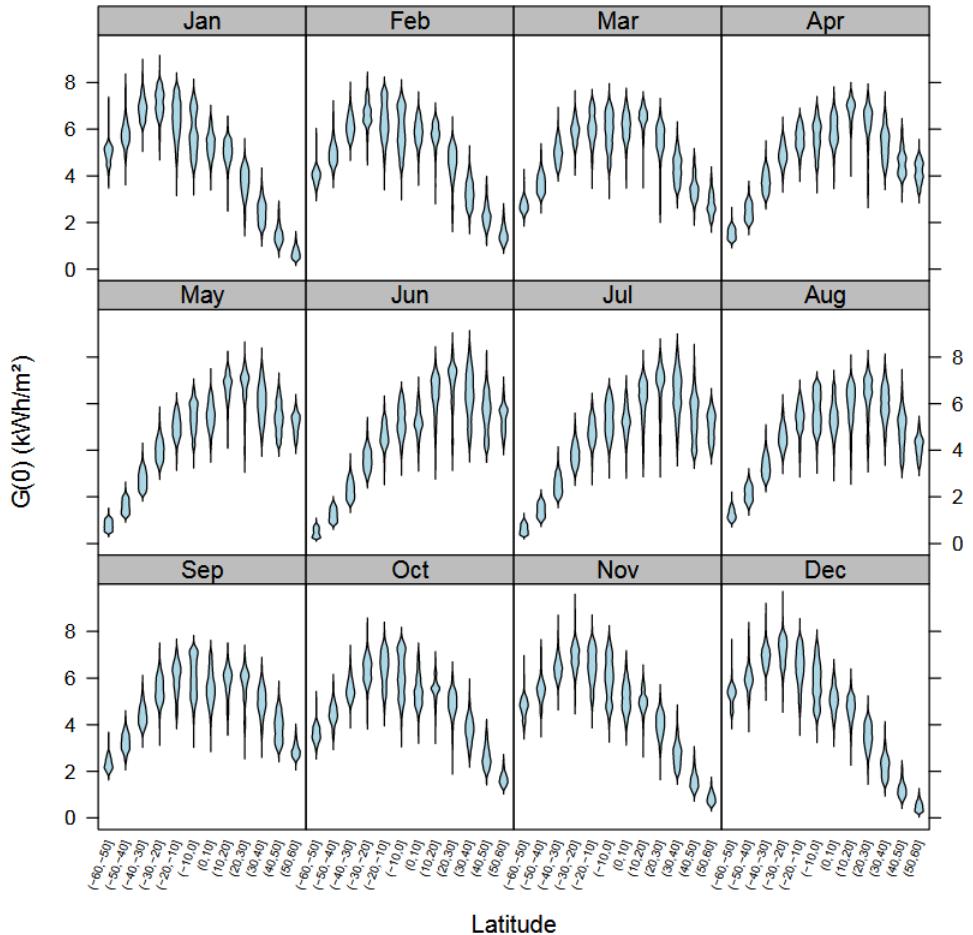
`cut()`: breaks a quantitative variable to a factor
`subset`: use only $-60 < \text{Lat} < 60$
`par.settings`: set some plot attributes
`scales`: tweak labeling of x axis, rotating labels

How does solar radiation vary with latitude, over months of the year?

- The result of this plot suggests some sort of scientific explanation
- Models to confirm/reject any of these would have to take the distributions into account

How was this graph produced?

- What was the plot formula?
- What was the panel function?
- What plot attributes where modified?



Data munging for plots & models

- Very often, the difficult problems in data analysis and graphics concern:
 - How to get my data into a format required for analysis?
 - How to get my data into a format for plotting?
 - How to get my model results into a table or plot?
- The first step is to understand the structure of your data

```
> str(nasa)
'data.frame': 64800 obs. of 15 variables:
 $ Lat: int -90 -90 -90 -90 -90 -90 -90 -90 -90 ...
 $ Lon: int -180 -179 -178 -177 -176 -175 -174 -173 -172 -171 ...
 $ Jan: num 9.63 9.63 9.63 9.63 9.63 9.63 9.63 9.63 9.63 ...
 $ Feb: num 5.28 5.28 5.28 5.28 5.28 5.28 5.28 5.28 5.28 ...
 $ Mar: num 0.75 0.75 0.75 0.75 0.75 0.75 0.75 0.75 0.75 ...
 $ Apr: num 0 0 0 0 0 0 0 0 0 ...
 $ May: num 0 0 0 0 0 0 0 0 0 ...
 $ Jun: num 0 0 0 0 0 0 0 0 0 ...
 $ Jul: num 0 0 0 0 0 0 0 0 0 ...
 $ Aug: num 0 0 0 0 0 0 0 0 0 ...
 $ Sep: num 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 ...
 $ Oct: num 3.24 3.24 3.24 3.24 3.24 3.24 3.24 3.24 3.24 ...
 $ Nov: num 8.28 8.28 8.28 8.28 8.28 8.28 8.28 8.28 8.28 ...
 $ Dec: num 11 11 11 11 11 ...
 $ Ann: num 3.19 3.19 3.19 3.19 3.19 3.19 3.19 3.19 3.19 ...
```

Previous plots used the annual average (Ann) against Latitude (Lat), with a plot formula:

$\text{Ann} \sim \text{cut}(\text{Lat}, \text{pretty}(\text{Lat}, 40))$

But now, we want to plot monthly values, Jan:Dec

Data munging

The solution used here works, but it is opaque, in that it tries to coerce the data into what is required for plot formulas for lattice

```
> (x <- paste(names(nasa)[3:14], collapse='+'))  
[1] "Jan+Feb+Mar+Apr+May+Jun+Jul+Aug+Sep+Oct+Nov+Dec"  
> (formula <- as.formula(paste(x, '~cut(Lat, pretty(Lat, 20))', sep="")))  
Jan + Feb + Mar + Apr + May + Jun + Jul + Aug + Sep + Oct + Nov +  
Dec ~ cut(Lat, pretty(Lat, 20))
```

With this, the monthly plot can be produced by:

```
bwplot(formula, data=nasa, subset=(abs(Lat)<60),  
       xlab='Latitude', ylab='G(0) (kWh/m2)',  
       outer=TRUE, as.table=TRUE, horizontal=FALSE,  
       col='lightblue',  
       panel=panel.violin,  
       scales=list(x=list(rot=70, cex=0.5)))
```