

# Introduction to ggplot2

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<http://euclid.psych.yorku.ca/www/psy6135/>

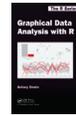
# Resources: Books



Hadley Wickham, *ggplot2: Elegant graphics for data analysis*, 2nd Ed.  
1st Ed: Online, <http://ggplot2.org/book/>  
ggplot2 Quick Reference: <http://sape.inf.usi.ch/quick-reference/ggplot2/>  
Complete ggplot2 documentation: <http://docs.ggplot2.org/current/>



Kieran Healy, *Data Visualization, a Practical Introduction*  
A hands-on introduction to data visualization using ggplot2, with a wide range of topics.  
The online version: <https://socviz.co/> is a great example of R bookdown publishing.



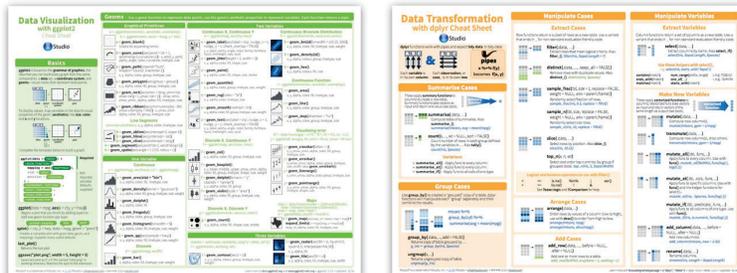
Antony Unwin, *Graphical Data Analysis with R*  
A gentle introduction to doing visual data analysis, mainly with ggplot2.  
R code: <http://www.gradaanwr.net/>



Winston Chang, *R Graphics Cookbook: Practical Recipes for Visualizing Data*  
Cookbook format, covering common graphing tasks; the main focus is on ggplot2  
R code from book: <http://www.cookbook-r.com/Graphs/>  
Download from: <http://ase.tufts.edu/bugs/guide/assets/R%20Graphics%20Cookbook.pdf>

# Resources: Cheat sheets

- R Studio maintains a large number of cheat sheets, <https://www.rstudio.com/resources/cheatsheets/>
- Topics:
  - [R Studio IDE](#), [Data import](#), [Data transformation \(dplyr\)](#), [Data visualization \(ggplot2\)](#), [R Markdown](#), ...
  - My collection: [R Studio Cheat Sheets](#)



# What is ggplot2?

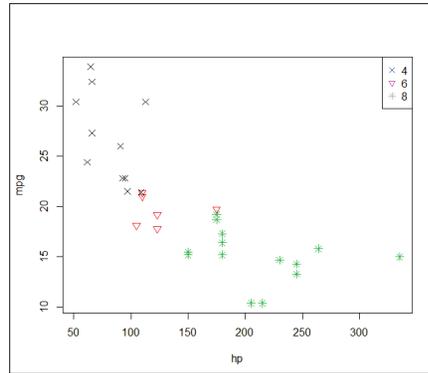
- ggplot2 is Hadley Wickham’s R package for producing “elegant graphics for data analysis”
  - An implementation of the ideas for graphics introduced in Lee Wilkinson’s *Grammar of Graphics*
  - These ideas and the syntax of ggplot2 help to think of graphs in a new and more general way
  - Produces pleasing plots, taking care of many of the fiddly details (legends, axes, colors, ...)
  - It is built upon the “grid” graphics system
  - It is open software, with a large number of gg\_ extensions. See: <https://exts.ggplot2.tidyverse.org/gallery/>

# ggplot2 vs base graphics

Some things that should be simple are harder than you'd like in base graphics

Here, I'm plotting gas mileage (mpg) vs. horsepower and want to use color and shape for different # of cylinders.

But I don't quite get it right!



```
mtcars$cyl <- as.factor(mtcars$cyl)
plot(mpg ~ hp, data=mtcars,
     col=cyl, pch=c(4,6,8)[mtcars$cyl], cex=1.2)
legend("topright", legend=levels(mtcars$cyl),
     pch = c(4,6,8),
     col=levels(mtcars$cyl))
```

colors and point symbols work differently in plot() and legend()

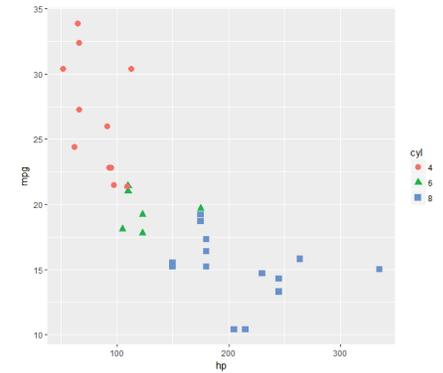
goal of ggplot2: this should "just work"

# ggplot2 vs base graphics

In ggplot2, just map the data variables to aesthetic attributes  
aes(x, y, shape, color, size, ...)

ggplot() takes care of the rest

aes() mappings set in the call to ggplot() are passed to geom\_point() here



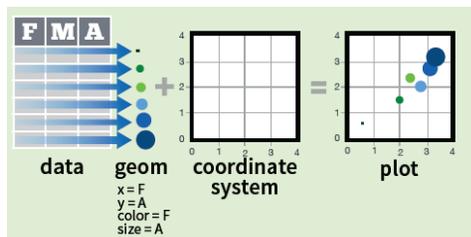
```
library(ggplot2)
ggplot(mtcars, aes(x=hp, y=mpg, color=cyl, shape=cyl)) +
  geom_point(size=3)
```

Follow along: the R script for this example is at: <http://euclid.psych.yorku.ca/www/psy6135/R/gg-cars.R>

# Grammar of Graphics

Every graph can be described as a combination of independent building blocks:

- **data**: a data frame: quantitative, categorical; local or data base query
- **aesthetic** mapping of variables into visual properties: size, color, x, y
- **geometric** objects ("geom"): points, lines, areas, arrows, ...
- **coordinate** system ("coord"): Cartesian, log, polar, map,

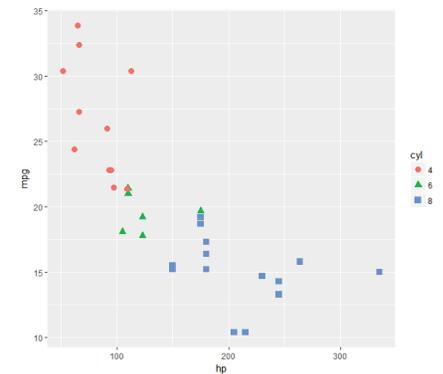


# ggplot2: data + geom -> graph

```
ggplot(data=mtcars,
     aes(x=hp, y=mpg,
         color=cyl, shape=cyl)) +
  geom_point(size=3)
```

- 1
- 2
- 3
- 4

- In this call,
1. data=mtcars: data frame
  2. aes(x=hp, y=mpg): plot variables
  3. aes(color, shape): attributes
  4. geom\_point(): what to plot
- the coordinate system is taken to be the standard Cartesian (x,y)



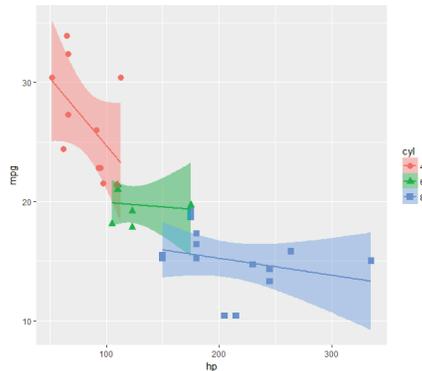
## ggplot2: geoms

Wow! I can really see something there.

How can I enhance this visualization?

Easy: add a `geom_smooth()` to fit linear regressions for each level of `cyl`

It is clear that horsepower and # of cylinders are highly related (Duh!)



```
ggplot(mtcars, aes(x=hp, y=mpg, color=cyl, shape=cyl)) +
  geom_point(size=3) +
  geom_smooth(method="lm", aes(fill=cyl))
```

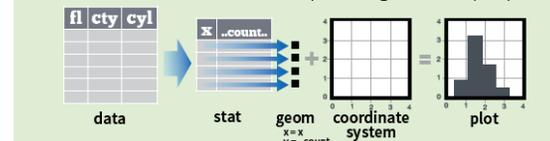
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## Grammar of Graphics

• Other GoG building blocks:

- **statistical transformations** ("stat") -- data summaries: mean, sd, binning & counting, ...
- **scales**: legends, axes to allow reading data from a plot

A stat builds new variables to plot (e.g., count, prop).



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## Grammar of Graphics

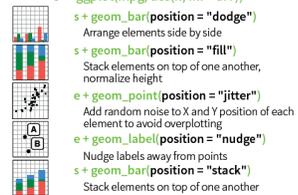
• Other GoG building blocks:

- **position** adjustments: jitter, dodge, stack, ...
- **faceting**: small multiples or conditioning to break a plot into subsets.

### Position Adjustments

Position adjustments determine how to arrange geoms that would otherwise occupy the same space.

```
s <- ggplot(mpg, aes(fl, fill = drv))
```



Each position adjustment can be recast as a function with manual **width** and **height** arguments

```
s + geom_bar(position = position_dodge(width = 1))
```

### Faceting

Facets divide a plot into subplots based on the values of one or more discrete variables.

```
t <- ggplot(mpg, aes(cty, hwy)) + geom_point()
```



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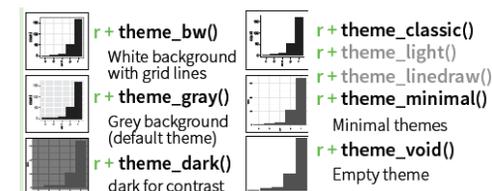
## ggplot2: GoG -> graphic language

• The implementation of GoG ideas in ggplot2 for R created a more expressive language for data graphs

- **layers**: graph elements combined with "+" (read: "and")

```
ggplot(mtcars, aes(x=hp, y=mpg)) +
  geom_point(aes(color = cyl)) +
  geom_smooth(method = "lm") +
```

- **themes**: change graphic elements consistently



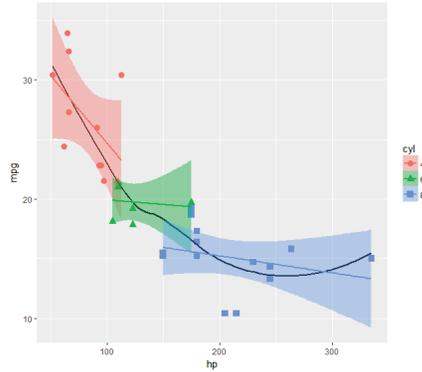
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## ggplot2: layers & aes()

Aesthetic attributes in the ggplot() call are passed to geom\_() layers

Other attributes can be passed as **constants** (size=3, color="black") or with aes(color=, ...) in different layers

This plot adds an overall loess smooth to the previous plot.  
color="black" overrides the aes(color=cyl)



```
ggplot(mtcars, aes(x=hp, y=mpg)) +
  geom_point(size=3, aes(color=cyl, shape=cyl)) +
  geom_smooth(method="lm", aes(color=cyl, fill=cyl)) +
  geom_smooth(method="loess", color="black", se=FALSE)
```

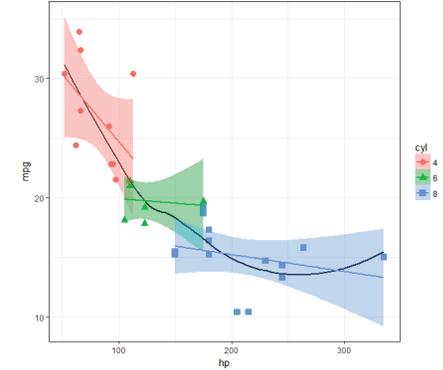
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## ggplot2: themes

All the graphical attributes of ggplot2 are governed by themes – settings for all aspects of a plot

A given plot can be rendered quite differently just by changing the theme

If you haven't saved the ggplot object, last\_plot() gives you something to work with further



```
last_plot() + theme_bw()
```

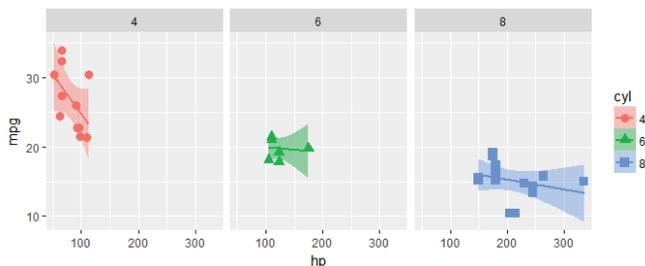
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## ggplot2: facets

Facets divide a plot into separate subplots based on one or more discrete variables

```
plt <-
  ggplot(mtcars, aes(x=hp, y=mpg, color=cyl, shape=cyl)) +
    geom_point(size=3) +
    geom_smooth(method="lm", aes(fill=cyl))
```

```
plt + facet_wrap(~cyl)
```



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## labeling points: geom\_text()

```
plt2 <- ggplot(mtcars, aes(x=wt, y=mpg)) +
  geom_point(color = 'red', size=2) +
  geom_smooth(method="loess") +
  labs(y="Miles per gallon", x="Weight (1000 lbs.)") +
  theme_classic(base_size = 16)
```

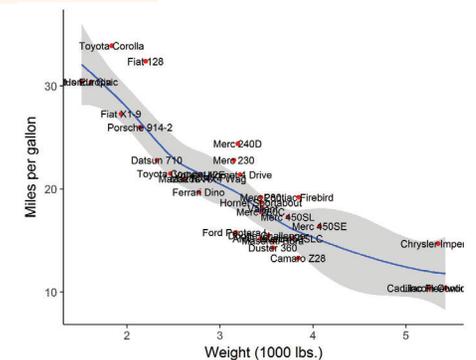
```
plt2 + geom_text(aes(label = rownames(mtcars)))
```

Sometimes it is useful to label points to show their identities.

geom\_text() usually gives messy, overlapping text

Note the use of theme\_classic() and better axis labels

But this is still messy: wouldn't want to publish this.



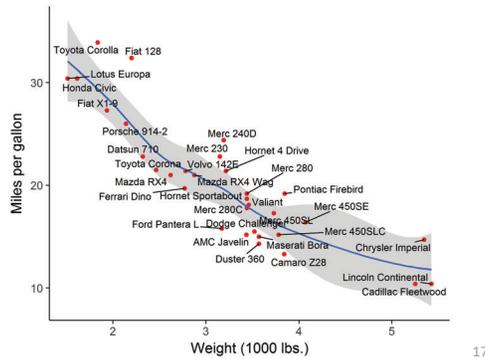
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# labeling points: geom\_text\_repel()

```
library(ggrepel)
plt2 +
  geom_text_repel(aes(label = rownames(mtcars)))
```

geom\_text\_repel() in the ggrepel package assigns repulsive forces among points and labels to assure no overlap

Some lines are drawn to make the assignment clearer



# labeling points: selection

It is easy to label points selectively, using some criterion to assign labels to points

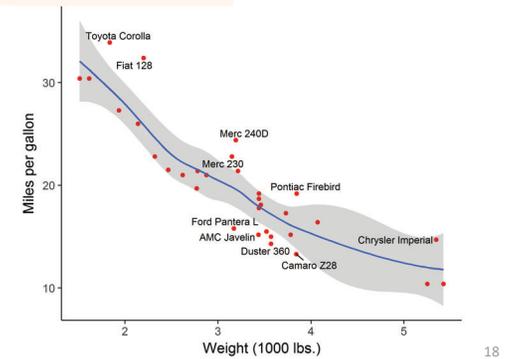
```
mod <- loess( mpg ~ wt, data=mtcars)
resids <- residuals(mod)
mtcars$label <- ifelse(abs(resids) > 2.5,
  rownames(mtcars), "")
```

```
plt2 + geom_text_repel(aes(label = mtcars$label))
```

- 1
- 2
- 3
- 4

Here, I:

1. fit the smoothed loess curve,
2. extract residuals,  $r_i$
3. assign labels where  $|r_i| > 2.5$
4. add the text layer

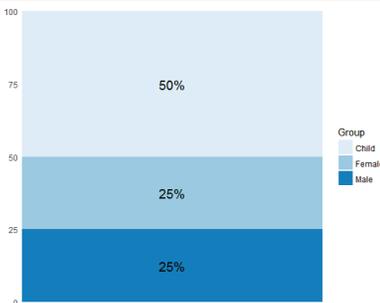


# ggplot2: coords

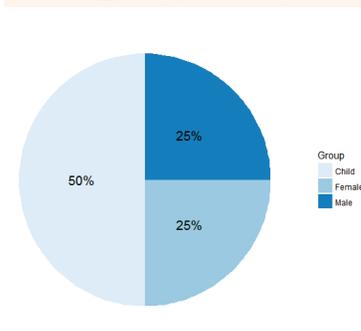
Coordinate systems, coord\_\*( ) functions, handle conversion from geometric objects to what you see on a 2D plot.

- A simple bar chart, standard coordinates
- A pie chart is just a bar chart in polar coordinates!

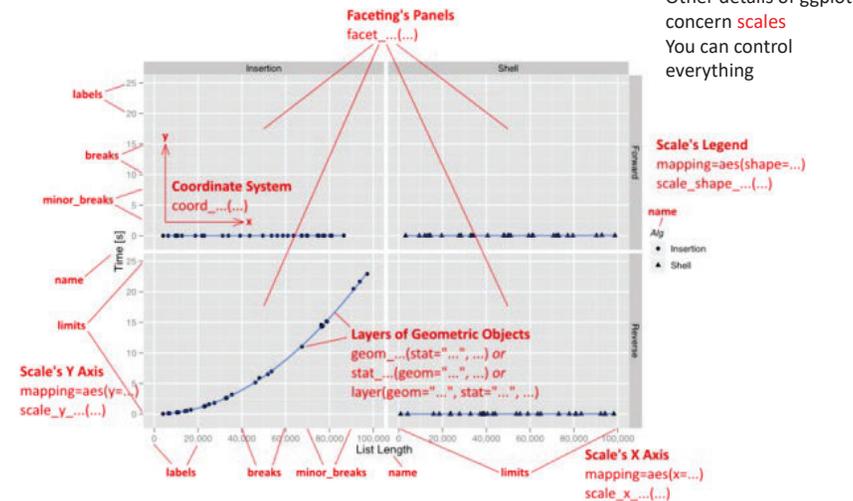
```
p <- ggplot(df, aes(x = "", y = value, fill = group)) +
  geom_bar( stat = "identity")
```



```
p + coord_polar("y", start = 0)
```



# Anatomy of a ggplot



Other details of ggplot concern scales. You can control everything

# ggplot objects

Traditional R graphics just produce graphical output on a device  
However, ggplot() produces a "ggplot" object, a list of elements

```
> names(plt)
[1] "data" "layers" "scales" "mapping" "theme" "coordinates"
[7] "facet" "plot_env" "labels"
> class(plt)
[1] "gg" "ggplot"
```

What methods are available?

```
> methods(class="gg")
[1] +
> methods(class="ggplot")
[1] grid.draw plot print summary
```

The "gg" class provides the "+" method

The "ggplot" class provides other, standard methods

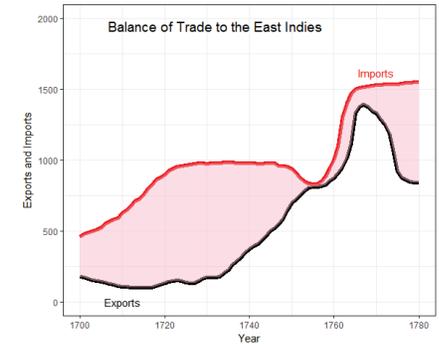
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# Playfair: Balance of trade charts

In the *Commercial and Political Atlas*, William Playfair used charts of imports and exports from England to its trading partners to ask "How are we doing"?

Here is a re-creation of one example, using ggplot2. How was it done?

```
> data(EastIndiesTrade, package="GDAdata")
> head(EastIndiesTrade)
  Year Exports Imports
1 1700  180   460
2 1701  170   480
3 1702  160   490
4 1703  150   500
5 1704  145   510
6 1705  140   525
...   ...   ...
```

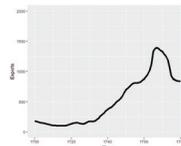


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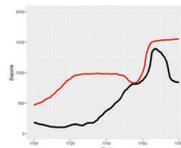
# ggplot thinking

I want to plot two time series, & fill the area between them

Start with a line plot of Exports vs. Year: **geom\_line()**  
Add a layer for the line plot of Imports vs. Year

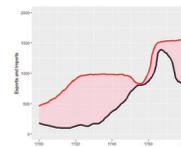


```
c1 <-
ggplot(EastIndiesTrade, aes(x=Year, y=Exports)) +
  ylim(0,2000) +
  geom_line(colour="black", size=2) +
  geom_line(aes(x=Year, y=Imports), colour="red", size=2)
```



Fill the area between the curves: **geom\_ribbon()**  
change the Y label

```
c1 <- c1 +
  geom_ribbon(aes(ymin=Exports, ymax=Imports), fill="pink") +
  ylab("Exports and Imports")
```



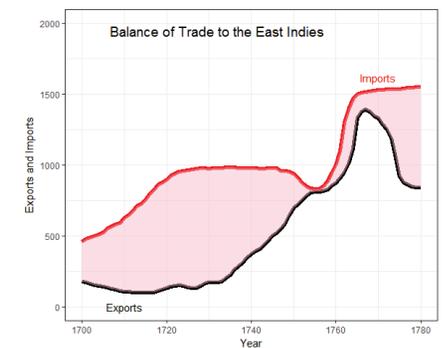
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That looks pretty good. Add some text labels using **annotate()**

```
c1 <- c1 +
  annotate("text", x = 1710, y = 0, label = "Exports", size=4) +
  annotate("text", x = 1770, y = 1620, label = "Imports", color="red", size=4) +
  annotate("text", x = 1732, y = 1950, label = "Balance of Trade to the East Indies", color="black", size=5)
```

Finally, change the theme to b/w

```
c1 <- c1 + theme_bw()
```



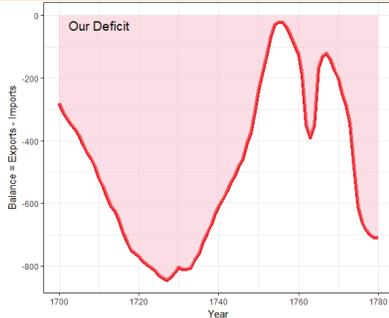
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## Plot what you want to show

Playfair's goal was to show the balance of trade with different countries.  
Why not plot Exports – Imports directly?

```
c2 <-
ggplot(EastIndiesTrade, aes(x = Year, y = Exports - Imports)) +
  geom_line(colour="red", size=2) +
  ylab("Balance = Exports - Imports") +
  geom_ribbon(aes(ymin=Exports-Imports, ymax=0), fill="pink", alpha=0.5) +
  annotate("text", x = 1710, y = -30, label = "Our Deficit", color="black", size=5) +
  theme_bw()
```

aes(x=, y=) can use expressions  
calculated from data variables



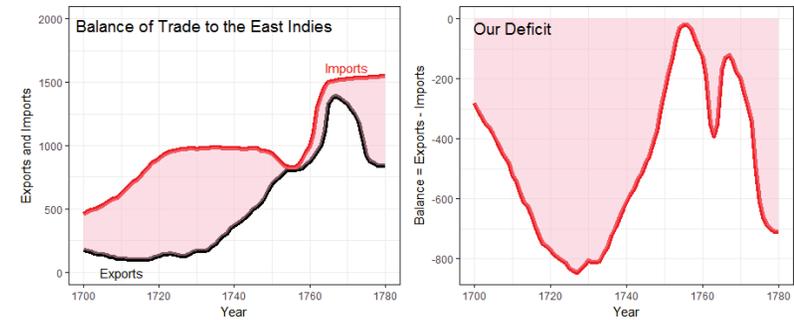
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## Composing several plots

ggplot objects use grid graphics for rendering

The `gridExtra` package has functions for combining or manipulating grid-based graphs

```
library(gridExtra)
grid.arrange(c1, c2, nrow=1)
```



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## Saving plots: ggsave()

- If the plot is on the screen

```
ggsave("path/filename.png") # height=, width=
```

- If you have a plot object

```
ggsave(myplot, file="path/filename.png")
```

- Specify size:

```
ggsave(myplot, "path/filename.png", width=6, height=4)
```

- any plot format (pdf, png, eps, svg, jpg, ...)

```
ggsave(myplot, file="path/filename.jpg")
```

```
ggsave(myplot, file="path/filename.pdf")
```

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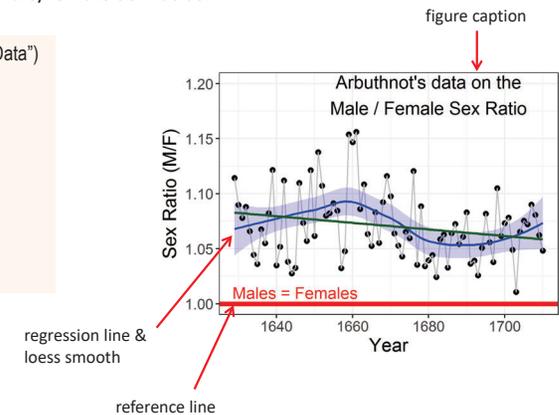
## Building a custom graph

Custom graphs can be constructed by adding graphical elements (points, lines, text, arrows, etc.) to a basic ggplot()

John Arbuthnot: data on male/female sex ratios:

```
> data(Arbuthnot, package="HistData")
> head(Arbuthnot[,c(1:3,6,7)])
  Year Males Females Ratio Total
1 1629 5218  4683 1.114 9.901
2 1630 4858  4457 1.090 9.315
3 1631 4422  4102 1.078 8.524
4 1632 4994  4590 1.088 9.584
5 1633 5158  4839 1.066 9.997
6 1634 5035  4820 1.045 9.855
... ..
```

Arbuthnot didn't make a graph. He simply calculated the probability that in 81 years from 1629–1710, the sex ratio would always be > 1. The first significance test!

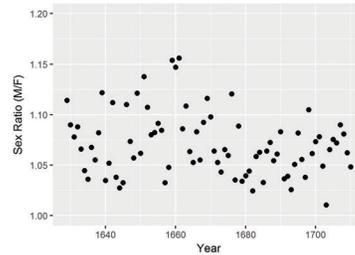


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## Building a custom graph

```
ggplot(Arbuthnot, aes(x=Year, y=Ratio)) +
  ylim(1, 1.20) +
  ylab("Sex Ratio (M/F)") +
  geom_point(pch=16, size=2)
```

Start with a basic scatterplot,  
Ratio vs. Year



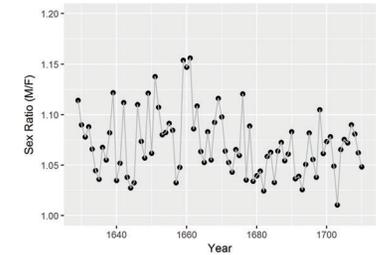
An R script for this example is available at:  
<http://euclid.psych.yorku.ca/www/psy6135/R/arbuthnot-gg.R>

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## Building a custom graph

```
ggplot(Arbuthnot, aes(x=Year, y=Ratio)) +
  ylim(1, 1.20) +
  ylab("Sex Ratio (M/F)") +
  geom_point(pch=16, size=2) +
  geom_line(color="gray")
```

Connect points with a line



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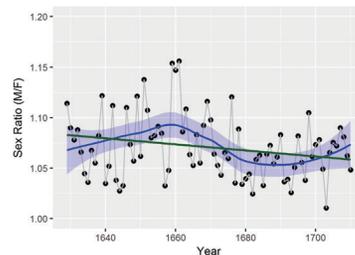
## Building a custom graph

```
ggplot(Arbuthnot, aes(x=Year, y=Ratio)) +
  ylim(1, 1.20) +
  ylab("Sex Ratio (M/F)") +
  geom_point(pch=16, size=2) +
  geom_line(color="gray") +
  geom_smooth(method="loess", color="blue",
             fill="blue", alpha=0.2) +
  geom_smooth(method="lm", color="darkgreen",
             se=FALSE)
```

Add smooths:

- loess curve
- linear regression line

```
# save what we have so far
arbuth <- last_plot()
```

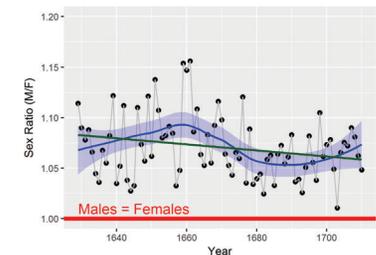


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## Building a custom graph

```
arbuth +
  geom_hline(yintercept=1, color="red", size=2) +
  annotate("text", x=1645, y=1.01, label="Males = Females", color="red", size=5)
```

Add horizontal reference line  
& text label



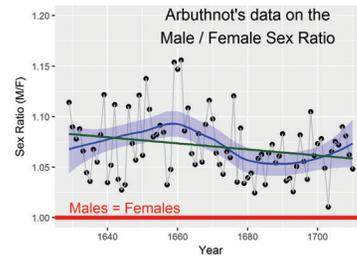
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## Building a custom graph

```

arbuth +
  geom_hline(yintercept=1, color="red", size=2) +
  annotate("text", x=1645, y=1.01, label="Males = Females", color="red", size=5) +
  annotate("text", x=1680, y=1.19,
    label="Arbutnot's data on the\nMale / Female Sex Ratio", size=5.5)
  
```

Add figure title



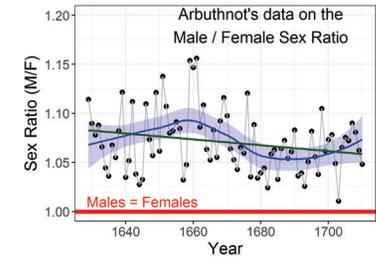
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## Building a custom graph

```

arbuth +
  geom_hline(yintercept=1, color="red", size=2) +
  annotate("text", x=1645, y=1.01, label="Males = Females", color="red", size=5) +
  annotate("text", x=1680, y=1.19,
    label="Arbutnot's data on the\nMale / Female Sex Ratio", size=5.5) +
  theme_bw() + theme(text = element_text(size = 16))
  
```

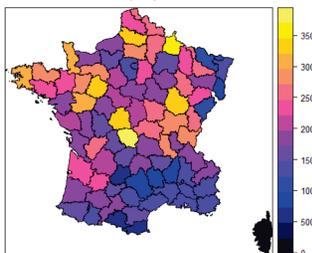
Change the theme and font size



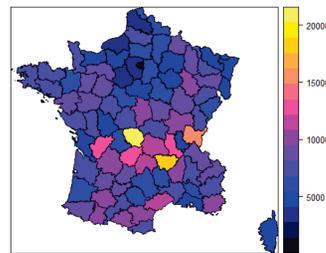
34

## Guerry: Moral statistics of France

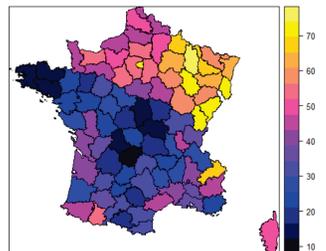
Persons per personal crime



Persons per property crime



Literacy (% Read & Write)



Guerry (1833) made shaded maps of France to determine if crime was related to literacy & other factors

```

library(Guerry)
library(sp)
spplot(gfrance, "Crime_pers")
spplot(gfrance, "Crime_prop")
spplot(gfrance, "Literacy")
  
```

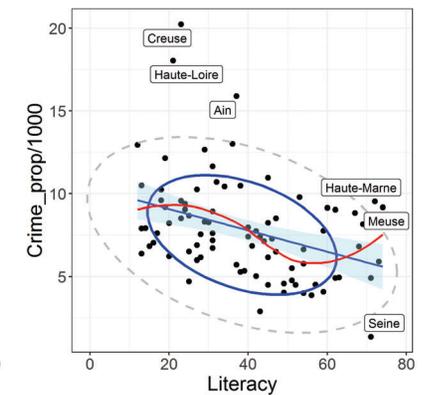
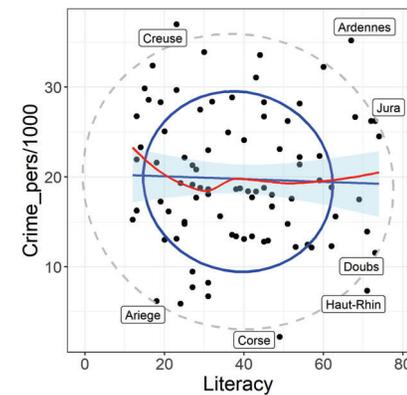
35

## Consulting for Guerry

Guerry: Mes cartes sont très jolies, non? But how can I go further?

MF: Make scatterplots! Add smooths & data ellipses.

Guerry: Les boissons sont sur moi!

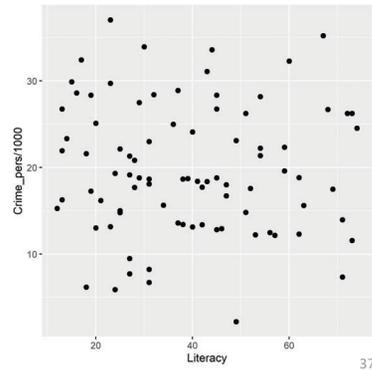


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## Building Guerry's plots

```
ggplot(aes(x=Literacy, y=Crime_pers/1000), data=Guerry) +  
  geom_point(size=2)
```

Start with a basic scatterplot



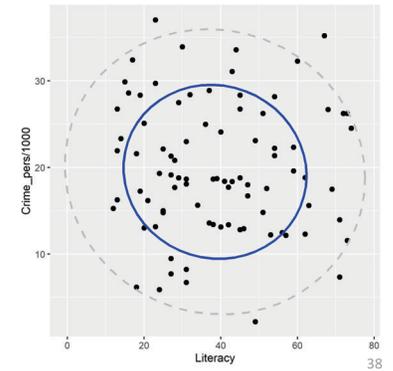
37

## Building Guerry's plots

```
ggplot(aes(x=Literacy, y=Crime_pers/1000), data=Guerry) +  
  geom_point(size=2) +  
  stat_ellipse(level=0.68, color="blue", size=1.2) +  
  stat_ellipse(level=0.95, color="gray", size=1, linetype=2)
```

Add data ellipses to show correlation

- 68% ~ mean  $\pm$  1 sd
- 95% ~ mean  $\pm$  2 sd



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## Guerry's plots: Add smooths

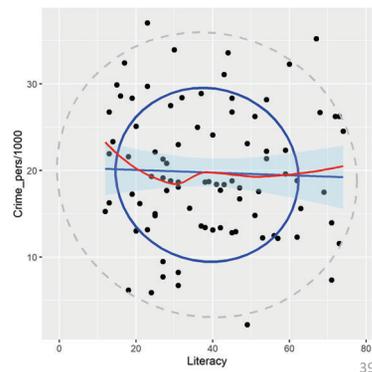
```
ggplot(aes(x=Literacy, y=Crime_pers/1000), data=Guerry) +  
  geom_point(size=2) +  
  stat_ellipse(level=0.68, color="blue", size=1.2) +  
  stat_ellipse(level=0.95, color="gray", size=1, linetype=2) +  
  geom_smooth(method="lm", formula=y~x, fill="lightblue") +  
  geom_smooth(method="loess", formula=y~x, color="red", se=FALSE)
```

Add lm() and loess() smooths

- lm shows regression slope
- loess diagnoses possible non-linearity

Coffee break: save the current plot object

```
gplot <- last.plot()
```



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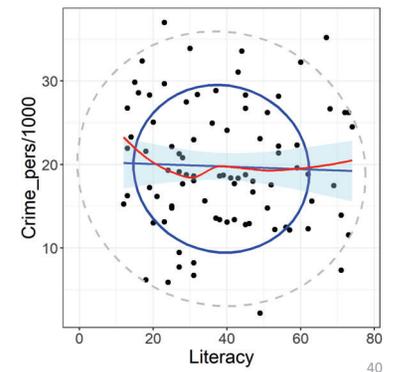
## Guerry's plots: Styling

Guerry: I want to publish this! But need to make axis labels larger

```
gplot <- last.plot()  
gplot + theme_bw() +  
  theme(text = element_text(size=18))
```

MF:

- Change the basic theme to theme\_bw()
- Increase the font size for all text
- You can change the style of anything you want



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# Guerry's plots: Labeling

Guerry: OK, but I see some unusual points. What are they?

MF: Need to calculate "unusualness" – Mahalanobis  $D^2$  squared distance from centroid

```
gdf <- Guerry[, c("Literacy", "Crime_pers", "Department")]  
gdf$dsq <- mahalanobis(gdf[,1:2], colMeans(gdf[,1:2]), cov(gdf[,1:2]))
```

$$D^2 = (x - \bar{x})' S^{-1} (x - \bar{x})'$$

```
library(ggrepel)  
gplot +  
  theme_bw() +  
  theme(text = element_text(size=18)) +  
  geom_label_repel(aes(label=Department),  
    data = gdf[gdf$dsq > 4.6,])
```

