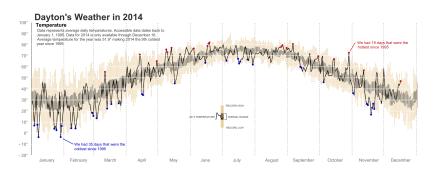
Basics for Enhanced Visualization: 3D/Data Grammar of graphics



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Source: https://rpubs.com/bradleyboehmke/weather graphic

Outline

- 1. Introduction
- 2. A grammar of graphics
- 3. ggplot2
- 4. Conclusions

These slides were partially inspired from

Wickham, H. (2010). A layered grammar of graphics. Journal of Computational and Graphical Statistics, 19(1), 3-28.

How to build a plot

▶ Draw scatter plot of *A vs C* indicating categorical variable *D*:

Data					
Α	В	С	D		
2	3	4	а		
1	2	1	а		
4	5	15	b		
9	10	80	b		

How to build a plot

▶ Draw scatter plot of A vs C indicating categorical variable D:

Data				
_A	В	С	D	
2	3	4	a	
1	2	1	а	
4	5	15	b	
9	10	80	b	

- One solution:
 - Draw a point for each observation.
 - Horizontal position given by A, vertical position by B.
 - D specifies the shape of the point.

How to build a plot

 First step: create a new dataset with the mappings of the visual aesthetics variables and the values of the variables to be displayed.

X	у	Shape
2	4	a
1	1	а
4	15	b
9	80	b

All other variables are deleted.

How to build a plot

- Second step: choose geometries for rendering visually the data.
- In our case the geometry is point. So that we have a scatter plot.
- But we can use lines

bars

texts

paths

ribbons

How to build a plot

 Third step: give scales to the variables from data units to physical units so that the computer know how to display it:

size (in pixels or mm), angles, symbols, colors.

How to build a plot

 Third step: give scales to the variables from data units to physical units so that the computer know how to display it:

size (in pixels or mm), angles, symbols, colors.

- For example: $x := \text{round}\left(\frac{x \min(x)}{\text{range}(x)} \times \text{graphic width}\right)$ pixels
- Scale x to [0, 200], y to [0, 300], "a" to "•" and "b" to "■"

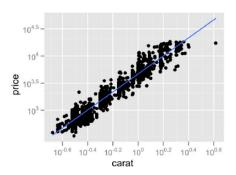
X	уу	Shape
25	11	•
0	0	•
75	53	
200	300	

How to build a plot

Third step: give scales to the variables from data units to physical units so that a computer know how to display it:

length (in pixels or mm), area, angle, symbol, color.

We can also use non linear scales such as log on each variable:

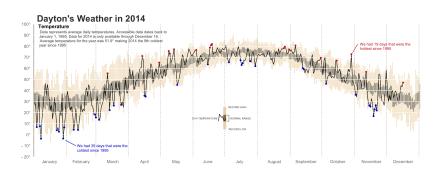


How to build a plot

- Fourth step: use statistical transformations to generate new variables.
- This can be, for example, binning or aggregation: counts, maximum, minimum, average, standard deviation etc.
- In this simple example we will show directly the data: transformation is identity.

How to build a plot

A graphic with both data and statistical transformations:

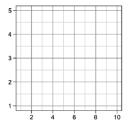


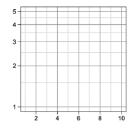
How to build a plot

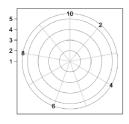
 Fifth step: choose coordinates system of the graphic:

Cartesian, log, semi-log, polar, map projection etc.

Cartesian, semi-log and polar coordinates:

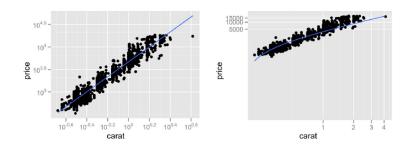






How to build a plot

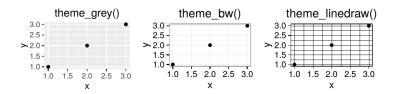
 Note that in some cases changing the scale (left) or the coordinates (right) have the same effect on displaying data.



The difference appears on the way the scales are shown.

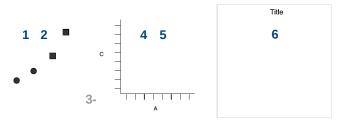
How to build a plot

- Sixth step: indicate additional non-data information title, small description of the context, names of the axes, units etc.
- ▶ The non-data part of the graphic is called a **theme**.



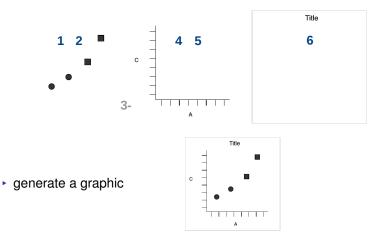
How to build a plot

All six steps together: 1- aesthetics, 2- geometries, 3- statistics,
 4- scales, 5- coordinates, 6- themes



How to build a plot

All six steps together: 1- aesthetics, 2- geometries, 3- statistics,
 4- scales, 5- coordinates, 6- themes



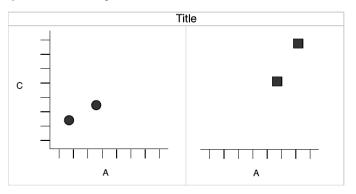
A more complex plot

- Consider that we want to generate different plots of the previous data conditioned on the variable D.
- This will lead us to create two facets of the data, i.e. two plots using subsets of the data set. See the table:

X	у	Shape	
Facet 1	25	11	•
raceti	0	0	•
Facet 2	75	53	
racel 2	200	300	

A more complex plot

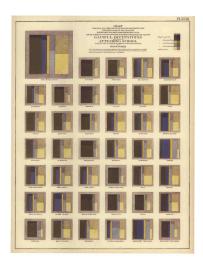
The plot with faceting:



- Note that all previous specifications can remain unchanged.
- Scales are normally specified globally, so that the facets can be compared.

A more complex plot

- Scales are normally specified globally, so that the facets can be compared.
- This is a way to describe small multiples: the same graphic pattern is used to different data.



The words of a graphic

Joining data, we can design a plethora of graphics.

aesthetics geometries statistics scales coordinates facets themes

These 8 elements can be changed almost independently one from the other. Similarly to a sentence construction with words:

The red bird is sitting in a tree.
The blue bird is sitting on a branch of the tree.
The yellow bird is singing on the top the tree.

The grammar of graphics

- These 8 elements can be changed almost independently one from the other.
- This is the reason this approach to construct graphics is called grammar of graphics.
- It was proposed originally in The Grammar of Graphics. L. Wilkinson. Springer Science & Business Media, 1999.

Source: books.google.com

The Grammar of Graphics

The grammar of graphics

 Further developed as the 8 elements approach presented here in the book

ggplot2 : Elegant Graphics for Data Analysis. H. Wickham. Springer, 2009.



- With the corresponding R package.
- It is also the base for *Tableau*, *IBM SPSS GPL* and Python's *Bokeh* package.

Examples

Pie chart

data object type × quantities

aesthetics object type - color

quantities - y

geometries stacked bar

vertical

statistics identity scales y - linear

color - integers

coordinates polar

y - θ

Examples

Pie chart

data object type × quantities

aesthetics object type - color

quantities - y

geometries stacked bar

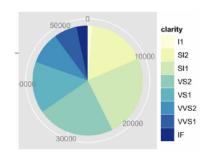
vertical

statistics identity scales y - linear

color - integers

coordinates polar

y - θ



Examples

Bullseye chart

data object type × quantities

aesthetics object type - color

quantities - y

geometries stacked bar

vertical

statistics identity scales v - linear

color - integers

coordinates polar

y - *r*

Examples

Bullseye chart

data object type × quantities

aesthetics object type - color

quantities - y

geometries stacked bar

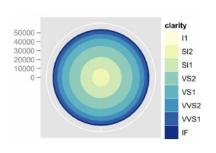
vertical

statistics identity scales y - linear

color - integers

coordinates polar

y - r



Examples

Bar chart

data object type × quantities

aesthetics object type - x quantities - y

object type - color

geometries bars

vertical

statistics identity

scales x - integers

y - linear

color - integers

coordinates Cartesian

Examples

Bar chart

data object type × quantities

aesthetics object type - x

quantities - y object type - color

geometries bars

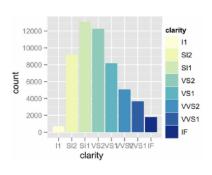
vertical

statistics identity

scales x - integers y - linear

color - integers

coordinates Cartesian



Examples

Coxcomb chart

data object type × quantities

aesthetics object type - x

quantities - y

object type - color

geometries bars

vertical

statistics identity

scales x - integers

y - linear

color - integers

coordinates Polar

y - *r*

Examples

Coxcomb chart

data object type × quantities

aesthetics object type - x

quantities - y

object type - color

geometries bars

vertical

statistics identity

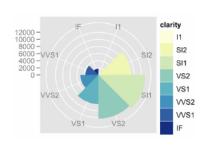
scales x - integers

y - linear

color - integers

coordinates Polar

y - *r*



Examples

Histogram

data object × price

aesthetics price - x

statistic - y

geometries bars

vertical

statistics count

30 uniform bins

scales x - linear

y - linear

coordinates Cartesian

Examples

Histogram

data object × price

aesthetics price - x

statistic - y

geometries bars

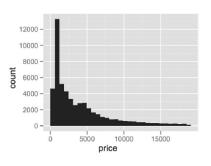
vertical

statistics count

30 uniform bins

scales x - linear y - linear

coordinates Cartesian



Examples

Density plot

data object × price

aesthetics price - x

statistic - y

geometries points statistics density

19 points

scales x - linear

y - linear

coordinates Cartesian

Examples

Density plot

data object × price

aesthetics price - x

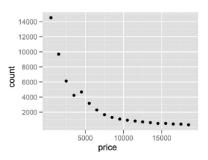
statistic - y

geometries points statistics density 19 points

scales x - linear

y - linear

coordinates Cartesian



Examples

Filled density plot

data object × price

aesthetics price - x

statistic - y

geometries ribbon

vertical starting at 0

statistics density

19 points

scales x - linear

y - linear

coordinates Cartesian

A grammar of graphics

Examples

Filled density plot

data object × price

aesthetics price - x

statistic - y

geometries ribbon

vertical starting at 0

statistics density

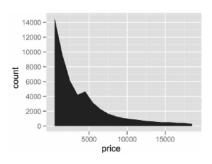
19 points

scales x - linear

v - linear

coordinates Cartesian

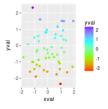
facets none themes none



A grammar of graphics

Axes and legends

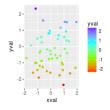
 Axes: guides indicating the inverse functions of scales of continuous variables or the spatial coordinates.



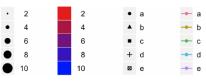
A grammar of graphics

Axes and legends

 Axes: guides indicating the inverse functions of scales of continuous variables or the spatial coordinates.

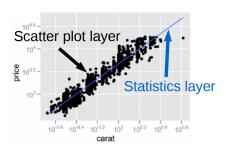


 Legends: guides indicating the inverse of scales of discrete variables.



A layered grammar of graphics

- Implementation of the grammar of graphics in R.
- It is a layered implementation: we can add different plots to the same graphic as layers:



R code: ggplot and layer

We start by defining a ggplot object:

```
ggplot()+...
```

then we add layers

- A layer has standard parameters:
 - geom: geometries
 - stat: statistics
 - position: positioning specification for geometric objects

R code: ggplot specification

 Data (data) and aesthetics mapping (mapping) can be either defined on the layer or passed directly to the ggplot object at the beginning.

ggplot(data=diamonds,mapping=aes(x=carat,y=price))+...

R code: ggplot specification

 Data (data) and aesthetics mapping (mapping) can be either defined on the layer or passed directly to the ggplot object at the beginning.

ggplot(data=diamonds,mapping=aes(x=carat,y=price))+...

- All subsequent layers will inherit the data and aesthetics used in ggplot.
- We can overwrite the default for a layer by respecifying data and aesthetics.

R code: ggplot and layer

We finish by adding scales and coordinates:

```
scale_y_continuous() +
```

scale_x_continuous() +

coord_cartesian()

R code: ggplot and layer

We finish by adding scales and coordinates:

```
scale_y_continuous() +
scale_x_continuous() +
coord_cartesian()
```

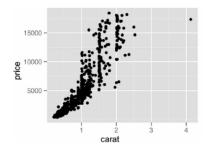
- ggplot2 has many default options and shortcuts to simplify plotting.
- All previous code can be rewritten using defaults and shortcuts:

```
ggplot(diamonds, aes(carat, price)) + geom_point()
```

Examples

Scatter plot

```
ggplot() +
layer(data = diamonds,
mapping = aes(x = carat, y = price),
geom = "point", stat = "identity",
position = "identity" ) +
scale_y_continuous() +
scale_x_continuous() +
coord_cartesian()
```



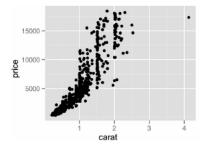
Examples

Scatter plot

```
ggplot() +
layer(data = diamonds,
mapping = aes(x = carat, y = price),
geom = "point", stat = "identity",
position = "identity" ) +
scale_y_continuous() +
scale_x_continuous() +
coord_cartesian()
```

or

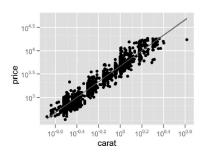
ggplot(diamonds, aes(carat, price)) +
geom_point()



Examples

Loglog scatter plot with linear regression

```
ggplot() +
layer(data = diamonds,
mapping = aes(x = carat, y = price),
geom = "point", stat = "identity",
position = "identity") +
layer(data = diamonds,
mapping = aes(x = carat, y = price),
geom = "smooth", position = "identity",
stat = "smooth", method = lm) +
scale_y_log10() + scale_x_log10() +
coord_cartesian()
```

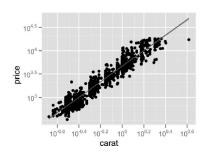


ggplot() +

Examples

Loglog scatter plot with linear regression

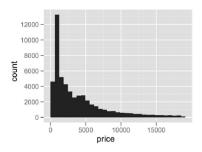
```
layer(data = diamonds,
mapping = aes(x = carat, y = price),
geom = "point", stat = "identity",
position = "identity") +
layer(data = diamonds,
mapping = aes(x = carat, y = price),
geom = "smooth", position = "identity",
stat = "smooth", method = lm) +
scale y log10() + scale x log10() +
coord cartesian()
                   or
ggplot(diamonds, aes(carat, price)) +
qeom point() +
stat smooth(method = Im) +
scale x log10() + scale y log10()
```



Examples

Histogram

```
ggplot(data = diamonds, mapping =
aes(price)) +
layer(geom = "bar", stat = "bin",
mapping = aes(y = ..count..))
```



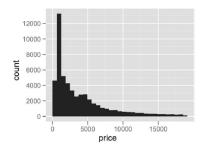
Examples

Histogram

```
ggplot(data = diamonds, mapping =
aes(price)) +
layer(geom = "bar", stat = "bin",
mapping = aes(y = ..count..))
```

or

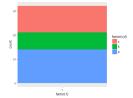
ggplot(diamonds, aes(x = price)) +
geom_histogram()



Examples

Stacked bar chart

```
sbar <- ggplot(mtcars,
aes(x = factor(1),
fill = factor(cyl))) +
geom_bar(width = 1)</pre>
```



Examples

Stacked bar chart

```
sbar <- ggplot(mtcars,
aes(x = factor(1),
fill = factor(cyl))) +
geom_bar(width = 1)</pre>
```

Pie chart

pie <- sbar + coord_polar("y")



Examples

Stacked bar chart

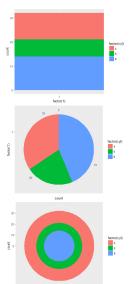
```
sbar <- ggplot(mtcars,
aes(x = factor(1),
fill = factor(cyl))) +
geom bar(width = 1)
```

Pie chart

pie <- sbar + coord polar("y")

Bullseye chart

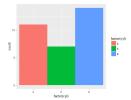
bull <- sbar + coord polar("x")



Examples

Bar chart

```
bar <- ggplot(mtcars,
aes(x = factor(cyl),
fill = factor(cyl))) +
geom_bar(width = 1)</pre>
```



Examples

Bar chart

```
bar <- ggplot(mtcars,
aes(x = factor(cyl),
fill = factor(cyl))) +
geom_bar(width = 1)
```

Coxcomb chart

cxc <- bar + coord_polar("x")</pre>



Examples

Bar chart

bar <- ggplot(mtcars,
aes(x = factor(cyl),
fill = factor(cyl))) +
geom_bar(width = 1)</pre>

Coxcomb chart

cxc <- bar + coord_polar("x")</pre>

?????? chart

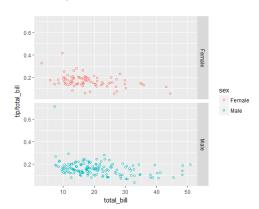
xxc <- bar + coord polar("y")



Examples

Small multiples

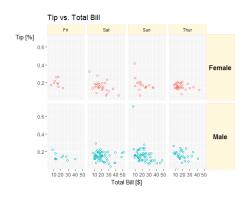
tips <- ggplot(tips, aes(x=total_bill, y=tip/total_bill,color=sex))=+ geom_point(shape=1) + facet_grid(sex ~ .)



Examples

Theme

```
tips + facet grid(sex ~ day) +
ggtitle("Tip vs. Total Bill") +
labs(x="Total Bill [$]",y="Tip [%]") +
theme(strip.text.x =
   element text(size=8, angle=0).
strip.text.y =
   element_text(size=12, face="bold",
angle=0).
strip.background =
   element rect(colour="antiquewhite",
      fill="cornsilk"),
panel.background =
   element rect(fill = 'gray96',
      colour = 'gray100'),
axis.title.y =
   element text(angle = 0).
legend.position="none")
```



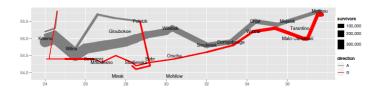
Examples

Napoleon's march

```
plot_troops <- ggplot(troops, aes(long, lat)) + geom_path(aes(size = survivors, color = direction, group = group))

plot_both <- troops_plot + geom_text(aes(label = city), size = 4, data = cities)

plot_polished <- both + scale_size(to = c(1, 10), breaks = c(1, 2, 3) * 10<sup>5</sup>, labels = comma(c(1, 2, 3) * 10<sup>5</sup>))+ scale_color_manual(values = c("grey50","red")) + xlab(NULL) + ylab(NULL)
```



Other resources

R studio cheat sheet (available on Jalon):

```
https://www.rstudio.com/wp-content/uploads/2016/11/ggplot2-cheatsheet-2.1.pdf
```

R cookbook (Graphs):

```
http://www.cookbook-r.com/Graphs/
```

ggplot2 book website:

```
http://ggplot2.org/book/
```

Conclusions

Conclusions

- Grammar of graphics ≠"zoo of charts" which is often proposed in any data processing software (Excel charts).
- Grammar approach: extend the effectiveness of our written communication to graphics.
- Fundamental idea: richness of the communication system comes from the structure, not only from the quantity of different symbols.

Conclusions

- First popular implementation: ggplot2, but others are appearing (Tableau).
- Still in its infancy: ggplot2 still has some "zoo of charts" and it does not take into account interaction.
- As any grammar it does not prevent us from constructing uggly phrases (graphics).