# Introduction to ggplot2

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## Goals

### What I will try to do

- ▶ give quick tour of ggplot2
- explain how to think about plots the ggplot2 way
- ► prepare/encourage you to learn more later

#### What I can't do in 30 minutes

- ► show every bell and whistle
- ► make you an expert at using ggplot2

## The Births78 data set – revised edition

```
require(dplyr)
require(mosaic)
require(lubridate)
Births <- Births78 %>%
  mutate( date = mdy(date) - years(100), # y2k fix
          wd = wday(date),
                                          # as a number
          wday = wday(date, label=TRUE, abbr=TRUE)
head(Births, 2)
```

```
## date births dayofyear wd wday
## 1 1978-01-01 7701 1 1 Sun
## 2 1978-01-02 7527 2 2 Mon
```

# The grammar of graphics

geom: the geometric "shape" used to display data (glyph)

▶ bar, point, line, ribbon, text, etc.

aesthetic: an attribute controlling how geom is displayed

► x position, y position, color, fill, shape, size, etc.

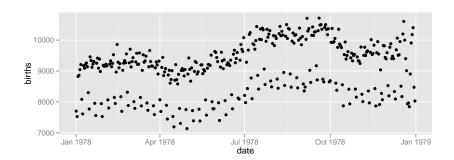
stat: a transformation applied to data before geom gets it

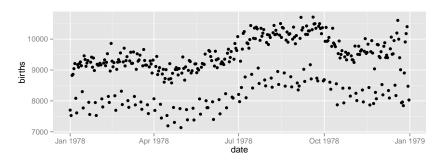
example: histograms work on binned data

scale: conversion of raw data to visual display

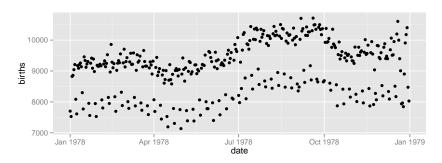
▶ particular assignment of colors, shapes, sizes, etc.

**guide**: helps user convert visual data back into raw data (legends, axes)



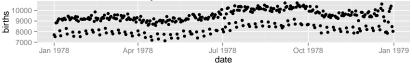


What does R need to know?

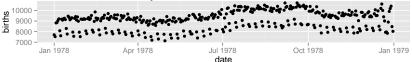


## What does R need to know?

- ► data source
- ▶ aesthetics
- ▶ geom dots



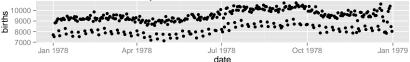
What does R need to know?



What does R need to know?

► data frame containing the data: ggplot(data=)

```
ggplot(data=Births)
```



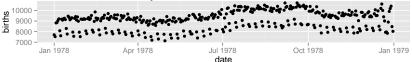
What does R need to know?

▶ data frame containing the data: ggplot(data=)

```
ggplot(data=Births)
```

\* how we want to map our aesthetics: aes()

```
ggplot(data=Births, aes(x=date, y=births))
```



What does R need to know?

► data frame containing the data: ggplot(data=)

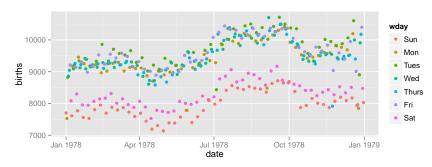
```
ggplot(data=Births)
```

\* how we want to map our aesthetics: aes()

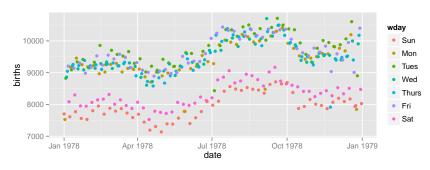
```
ggplot(data=Births, aes(x=date, y=births))
```

▶ what geom we want to use: + geom point()

```
ggplot(data=Births, aes(x=date, y=births)) + geom_point()
```



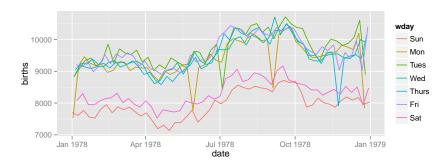
What information has changed?

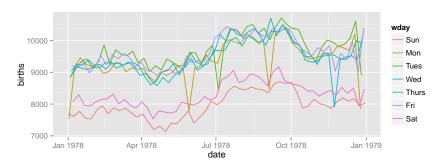


What information has changed?

▶ new aesthetic: mapping color to day of week

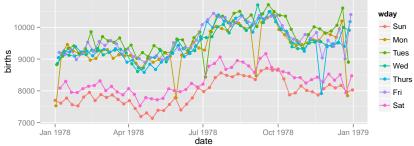
```
ggplot(data=Births, aes(x=date, y=births, color=wday)) +
geom_point()
```





This time we use lines instead of dots

```
ggplot(data=Births, aes(x=date, y=births, color=wday)) +
  geom_line()
```





This time we have two layers, one with points and one with lines

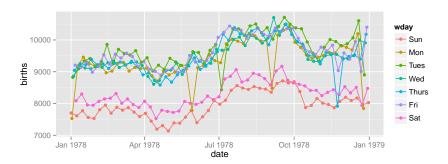


This time we have two layers, one with points and one with lines

► The layers are placed one on top of the other: the points are *below* and the lines are *above*. Sometimes the order of the layers can be important because of overplotting.

## Alternative Syntax

```
Births %>%
  ggplot(aes(x=date, y=births, color=wday)) +
  geom_point() +
  geom_line()
```

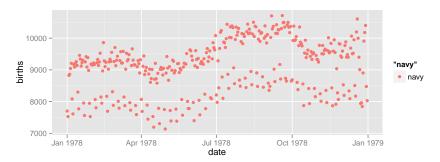


## What does this do?

```
Births %>%
  ggplot(aes(x=date, y=births, color="navy")) +
  geom_point()
```

## What does this do?

```
Births %>%
  ggplot(aes(x=date, y=births, color="navy")) +
  geom_point()
```

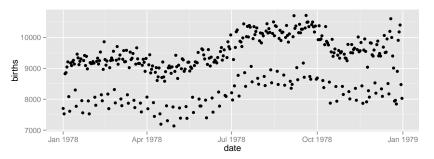


This is *mapping* the color aesthetic to a new variable with only one value ("navy"). So all the dots get set to the same color, but it's not navy.

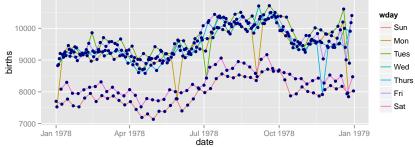
## Setting vs. Mapping

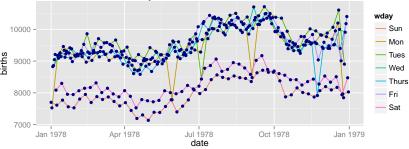
If we want to *set* the color to be navy for all of the dots, we do it this way:

```
Births %>%
  ggplot(aes(x=date, y=births), color="navy") +
  geom_point()
```



► Note that colour = "navy" is now outside of the aesthetics list. That's how ggplot2 distinguishes between mapping and setting.





```
Births %>%
  ggplot(aes(x=date, y=births)) +
  geom_line(aes(color=wday)) +
  geom_point(color="navy")
```

- ► ggplot() establishes the default data and aesthetics for the geoms, but each geom may change these defaults.

## Other geoms

```
apropos("^geom_")
```

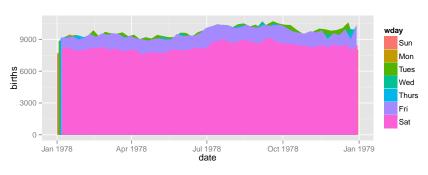
```
[1] "geom_abline"
                          "geom_area"
                                               "geom_bar"
[4] "geom_bin2d"
                          "geom_blank"
                                               "geom_boxplot
[7] "geom_contour"
                                               "geom_density
                          "geom_crossbar"
[10] "geom_density2d"
                          "geom_dotplot"
                                               "geom_errorban
[13] "geom_errorbarh"
                          "geom_freqpoly"
                                               "geom_hex"
[16] "geom_histogram"
                          "geom_hline"
                                               "geom_jitter"
[19] "geom_line"
                          "geom_linerange"
                                               "geom_map"
[22] "geom_path"
                          "geom_point"
                                               "geom_pointra
[25] "geom_polygon"
                          "geom_quantile"
                                               "geom_rangefra
[28] "geom_raster"
                                               "geom_ribbon"
                          "geom_rect"
[31] "geom_rug"
                          "geom_segment"
                                               "geom_smooth"
[34] "geom_step"
                          "geom_text"
                                               "geom_tile"
[37] "geom_tufteboxplot" "geom_violin"
                                               "geom_vline"
```

help pages will tell you their aesthetics and default stats

4 D > 4 A > 4 B > 4 B > 9 Q (

# Let's try geom\_area

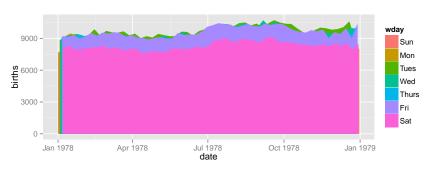
```
Births %>%
  ggplot(aes(x=date, y=births, fill=wday)) +
  geom_area()
```



This is not a good plot

# Let's try geom\_area

```
Births %>%
  ggplot(aes(x=date, y=births, fill=wday)) +
  geom_area()
```



### This is not a good plot

- overplotting is hiding much of the data
- extending y-axis to 0 may or may not be desirable.

Side note: what makes a plot good?

Most (all?) graphics are intended to help us make comparisons

- ► How does something change over time?
- ▶ Do my treatments matter? How much?
- ▶ Do men and women respond the same way?

**Key plot metric:** Does my plot make the comparisions I am interested in

- ► easily, and
- ► accurately?

## Time for some different data

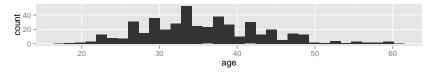
HELPrct: Health Evaluation and Linkage to Primary care randomized clinical trial

?HELPrct

How old are people in the HELP study?

# How old are people in the HELP study?

```
HELPrct %>%
  ggplot(aes(x=age)) +
  geom_histogram()
```

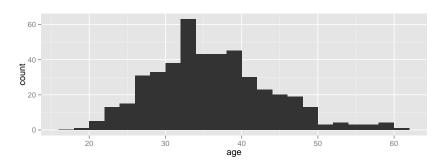


### Notice the messages

- stat\_bin: Histograms are not mapping the raw data but binned data.
  - stat\_bin() performs the data transformation.
- binwidth: a default binwidth has been selected, but we should really choose our own.

## Setting the binwidth manually

```
HELPrct %>%
  ggplot(aes(x=age)) +
  geom_histogram(binwidth=2)
```



## Data Flow

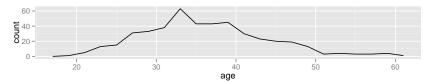
org data  $\xrightarrow{\text{stat}}$  statified  $\xrightarrow{\text{aesthetics}}$  aesthetic data  $\xrightarrow{\text{scales}}$  scaled data

### Simplifications:

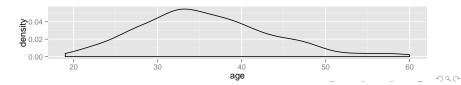
- aesthetics get computed twice, once before the stat and again after. Example: histograms
- ▶ item we need to look at the aesthetics to figure out which variable to bin
  - ▶ then the stat does the binning
  - bin counts become part of the aesthetics for geom: y=..count..
- ► This process happens in each layer

# How old are people in the HELP study? - Other geoms

```
HELPrct %>%
  ggplot(aes(x=age)) +
  geom_freqpoly(binwidth=2)
```



```
HELPrct %>%
  ggplot(aes(x=age)) +
  geom_density()
```

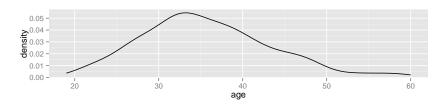


## Selecting stat and geom manually

### Every geom comes with a default stat

- ▶ for simple cases, the stat is stat\_identity() which does nothing
- ▶ we can mix and match geoms and stats however we like

```
HELPrct %>%
  ggplot(aes(x=age)) +
  geom_line(stat="density")
```

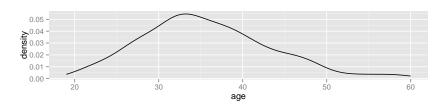


## Selecting stat and geom manually

### Every stat comes with a default geom

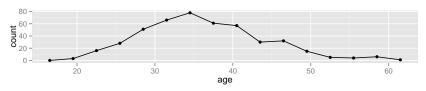
- ▶ we can specify stats instead of geom, if we prefer
- ▶ we can mix and match geoms and stats however we like

```
HELPrct %>%
  ggplot(aes(x=age)) +
  stat_density( geom="line")
```

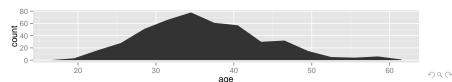


#### More combinations

```
HELPrct %>%
  ggplot(aes(x=age)) +
  geom_point(stat="bin", binwidth=3) +
  geom_line(stat="bin", binwidth=3)
```



```
HELPrct %>%
  ggplot(aes(x=age)) +
  geom_area(stat="bin", binwidth=3)
```



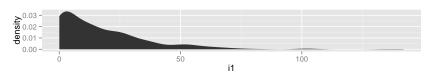
Your turn: How much do they drink? (i1)

Create a plot that shows the distribution of the average daily alcohol consumption in the past 30 days (i2).

# How much do they drink? (i1)

```
HELPrct %>%
    ggplot(aes(x=i1)) + geom_histogram()
```





## Covariates: Adding in more variables

Q. How does alcohol consumption (or age, your choice) differ by sex and substance (alcohol, cocaine, heroin)?

#### Decisions:

- How will we display the variables: i1 (or age), sex, substance
- ► What comparisons are we most interested in?

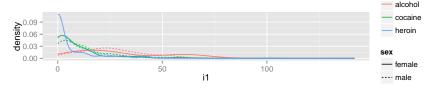
#### Give it a try.

► Note: I'm cheating a bit. You may want to do some things I haven't shown you yet. (Feel free to ask.)

## Covariates: Adding in more variables

Using color and linetype:

```
HELPrct %>%
  ggplot(aes(x=i1, color=substance, linetype=sex)) +
  geom_line(stat="density")
```



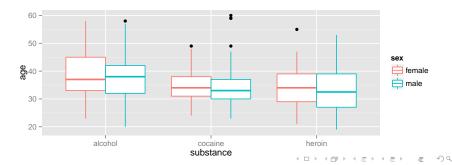
Using color and facets

```
HELPrct %>%
   ggplot(aes(x=i1, color=substance)) +
   geom_line(stat="density") + facet_grid( . ~ sex )
```

### **Boxplots**

Boxplots use stat\_quantile() which computes a five-number summary (roughly the five quartiles of the data) and uses them to define a "box" and "whiskers". The quantitative variable must be y, and there must be an additional x variable.

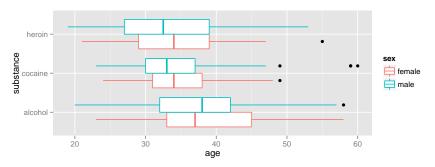
```
HELPrct %>%
   ggplot(aes(x=substance, y=age, color=sex)) +
   geom_boxplot()
```



#### Horizontal boxplots

Horizontal boxplots are obtained by flipping the coordinate system:

```
HELPrct %>%
  ggplot(aes(x=substance, y=age, color=sex)) +
  geom_boxplot() +
  coord_flip()
```

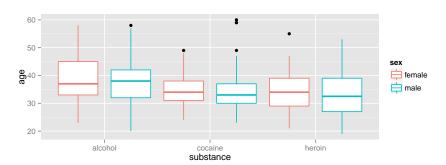


► coord\_flip() may be used with other plots as well to reverse the roles of x and y on the plot.

### Give me some space

We've triggered a new feature: dodge (for dodging things left/right). We can control how much if we set the dodge manually.

```
HELPrct %>%
   ggplot(aes(x=substance, y=age, color=sex)) +
   geom_boxplot(position=position_dodge(width=1))
```

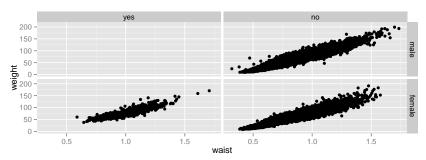


# Issues with bigger data

```
dim(NHANES)
```

```
## [1] 31126 53
```

```
NHANES %>% ggplot(aes(x=waist, y=weight)) +
geom_point() + facet_grid( sex ~ pregnant )
```



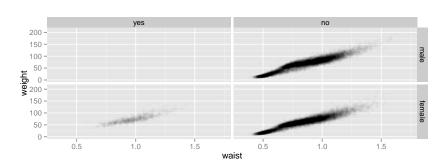
► Although we can see a generally positive association (as we would expect), the overplotting may be hiding information.



## Using alpha (opacity)

One way to deal with overplotting is to set the opacity low.

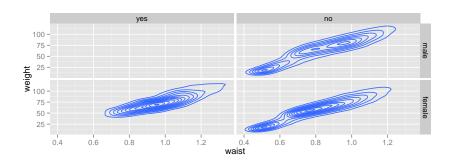
```
NHANES %>%
  ggplot(aes(x=waist, y=weight)) +
  geom_point(alpha=0.01) + facet_grid( sex ~ pregnant )
```



## geom\_density2d

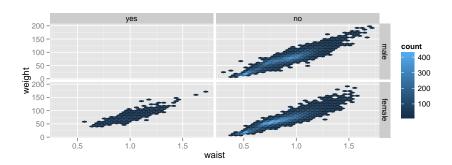
Alternatively (or simultaneously) we might prefere a different geom altogether.

```
NHANES %>%
  ggplot(aes(x=waist, y=weight)) +
  geom_density2d() + facet_grid( sex ~ pregnant )
```



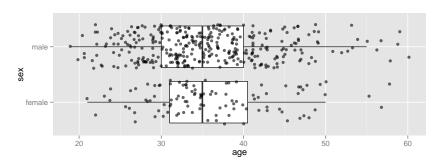
## geom\_hex

```
NHANES %>%
  ggplot(aes(x=waist, y=weight)) +
  geom_hex() + facet_grid( sex ~ pregnant )
```



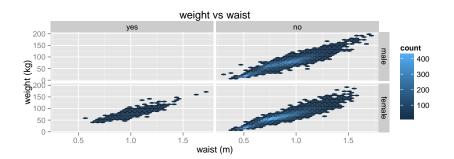
### Multiple layers

```
ggplot( data=HELPrct, aes(x=sex, y=age)) +
  geom_boxplot(outlier.size=0) +
  geom_jitter(alpha=.6) +
  coord_flip()
```



## Labeling

```
NHANES %>%
  ggplot(aes(x=waist, y=weight)) +
  geom_hex() + facet_grid( sex ~ pregnant ) +
  labs(x="waist (m)", y="weight (kg)", title="weight vs was
```

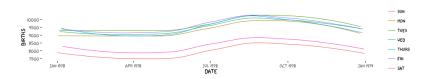


- scales (fine tuning mapping from data to plot)
- ▶ guides (so reader can map from plot to data)
- ▶ coords (coord\_flip() is good to know about)
- ► themes (for customizing appearance)

```
require(ggthemes)
qplot( x=date, y=births, data=Births) + theme_wsj()
```

```
10000
9000
8000
7000 Jan 1978 Apr 1978 Jul 1978 Oct 1978 Jan 1979
```

- scales (fine tuning mapping from data to plot)
- ▶ guides (so reader can map from plot to data)
- ► coords (coord\_flip() is good to know about)
- ▶ themes (for customizing appearance)



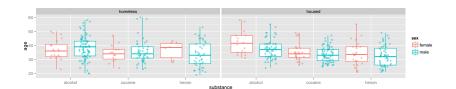
- scales (fine tuning mapping from data to plot)
- ▶ guides (so reader can map from plot to data)
- ► coords (coord\_flip() is good to know about)
- ▶ themes (for customizing appearance)
- position (position\_dodge() can be used for side by side bars)

```
ggplot( data=HELPrct, aes(x=substance, y=age, color=sex)) -
geom_violin(coef = 10, position=position_dodge()) +
geom_point(aes(color=sex, fill=sex), position=position_j
```



- scales (fine tuning mapping from data to plot)
- ▶ guides (so reader can map from plot to data)
- ► themes (for customizing appearance)
- position (position\_dodge(), position\_jitterdodge(), position\_stack(), etc.)

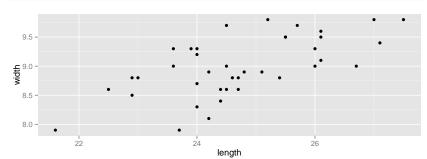
### A little bit of everything



#### Some short cuts

1. qplot() provides "quick plots" for ggplot2

```
qplot(length, width, data=KidsFeet)
```



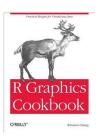
2. mplot(dataframe) provides an interactive plotting tool for both ggplot2 and lattice.

#### mplot(HELPrct)

▶ quickly make several plots from a data frame ( □ > ( ≥ > ( ≥ > ) ≥ ) < ○

#### Want to learn more?

- ► docs.ggplot2.org/
- ► Winston Chang's: *R Graphics Cookbook*



#### What's around the corner?

#### ggvis

- ► dynamic graphics (brushing, sliders, tooltips, etc.)
- ▶ uses Vega (D3) to animate plots in a browser
- ► similar structure to ggplot2 but different syntax and names
- ▶ version 0.3 just released to github

#### Dynamic documents

- ► combination of RMarkdown, ggvis, and shiny
- ▶ beta testing now