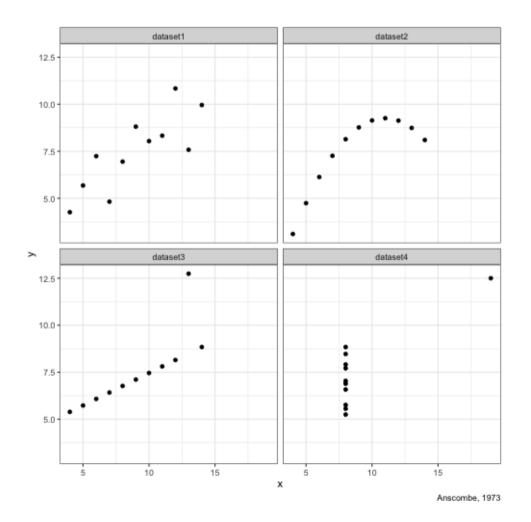
# Visualizing data

Abhijit Dasgupta

Fall, 2019

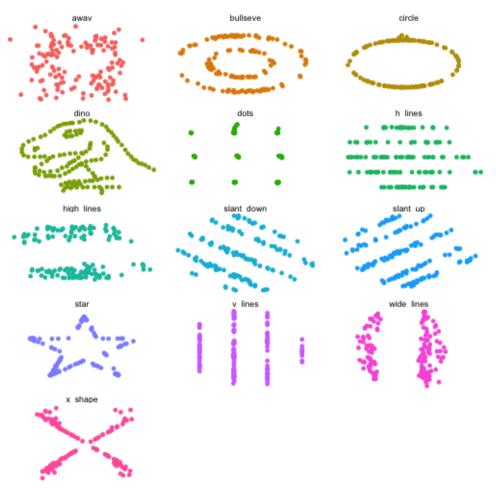
# Why visualize data?

## **Anscombe's data**



Value
9
7.5
11
4.13
0.82

## The DataSaurus dozen



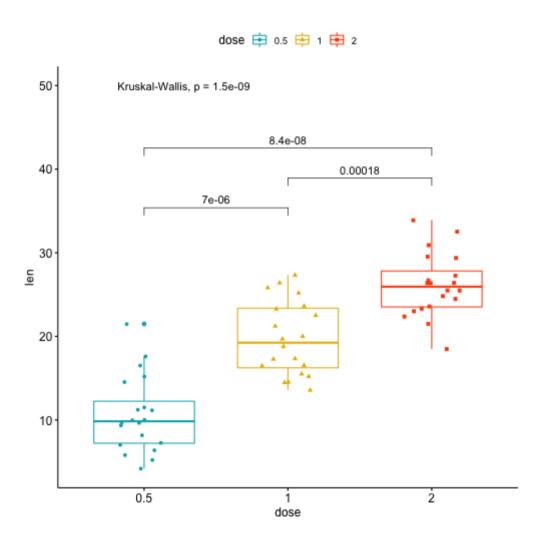
Statistic	Value
mean(x)	54.3
mean(y)	47.8
var(x)	281
<pre>var(y)</pre>	725
cor(x,y)	-0.07

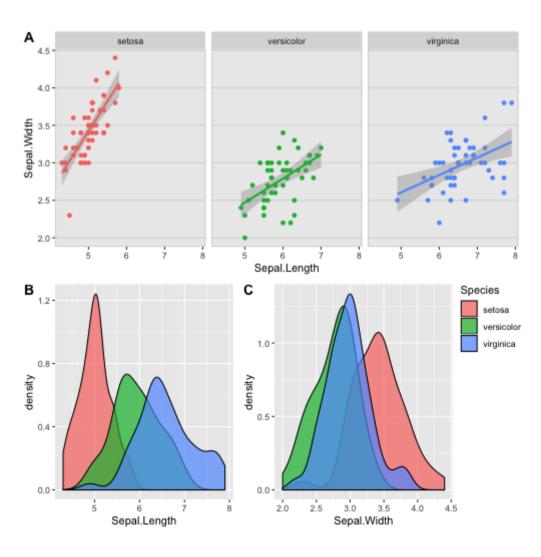
Matejka & Fitzmaurice, 2017

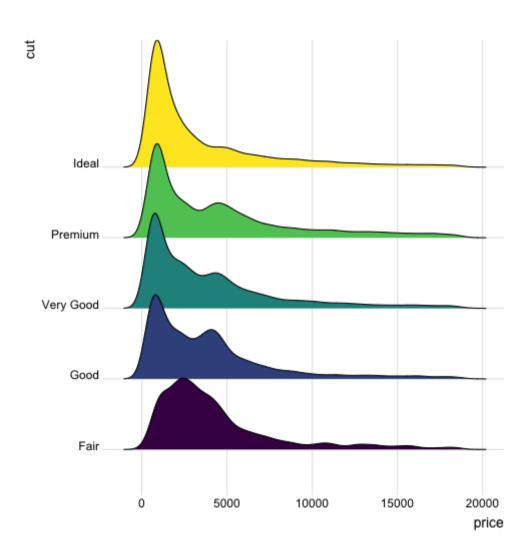
### **Bottom line**

- Summary statistics cannot always distinguish datasets
- Take advantage of humans' ability to visually recognize and remember patterns
- Find discrepancies in the data more easily

# Some examples







## **Manhattan plot**

**Circular Manhattan plot** 

## Maps

**Interactive graphs** 

# Data visualization with ggplot2

## What is ggplot2?

- A second (and final) iteration of the ggplot
- Implementation of Wilkerson's Grammar of Graphics in R
- Conceptually, a way to layer different elements onto a canvas to create a data visualization
- Started as Dr. Hadley Wickham's PhD thesis (with Dr. Dianne Cook)
- Won the John M. Chambers Statistical Software Award in 2006

- Mimicked in other software platforms
  - ggplot and seaborn in Python
  - Translated in plotly

## ggplot2 uses the grammar of graphics

#### A grammar ...

- compose and re-use small parts
- build complex structures from simpler units

#### of graphics ...

- Think of yourself as a painter
- Build a visualization using layers on a canvas
- Draw layers on top of each other

The ggplot2 package is a very flexible and (to me) intuitive way of visualizing data. It is based on the concept of layering elements on a canvas.

This idea of layering graphics on a canvas is, to me, a nice way of building graphs

#### You need:

- A data.frame object
- Aesthetic mappings (aes) to say what data is used for what purpose in the viz
  - x- and y-direction
  - shapes, colors, lines
- A geometry object (geom) to say what to draw
  - You can "layer" geoms on each other to build plots

ggplot used pipes before pipes were a thing.

However, it uses the + symbol for piping rather than the %>% operator, since it pre-dates the tidyverse

```
library(ggplot2)
ggplot(mtcars, aes(x = wt, y = mpg)) + geom_point()
```

- A data.frame object: mtcars
- Aesthetic mapping:
  - x-axis: wt
  - y-axis: mpg
- Geometry:
  - geom\_point: draw points

```
library(ggplot2)
ggplot(mtcars, aes(x = wt, y = mpg)) + geom_point()+ geom_smooth()
```

- A data. frame object: mtcars
- Aesthetic mapping:
  - x-axis: wt
  - y-axis: mpg
- Geometry:
  - geom\_point: draw points
  - geom\_smooth: Add a layer which draws a best-fitting line

#### A dataset

We will use the beaches dataset

```
library(tidyverse)
library(rio)
beaches <- import('data/sydneybeaches3.csv')</pre>
```

```
date year month day season rainfall temperature enterococci
    1 2013-01-02 2013
                                             0.0
                                                        23.4
                                                                      6.7
   2 2013-01-06 2013
                              6
                                             0.0
                                                        30.3
                                                                      2.0
   3 2013-01-12 2013
                                             0.0
                                                        31.4
                                                                     69.1
                             18
   4 2013-01-18 2013
                                             0.0
                                                        46.4
                                                                     9.0
   5 2013-01-24 2013
                             24
                                             0.0
                                                        27.5
                                                                     33.9
   6 2013-01-30 2013
                             30
                                             0.6
                                                        26.6
                                                                    26.5
      day_num month_num month_name season_name
#>
                           January
                                         Summer
            6
                           January
                                         Summer
                           January
                                         Summer
           18
                           January
                                         Summer
           24
                           January
                                         Summer
           30
                           January
                                         Summer
```

Credit: D. J. Navarro

# Building a graph

### Start with a blank canvas

ggplot()

### Add a data set

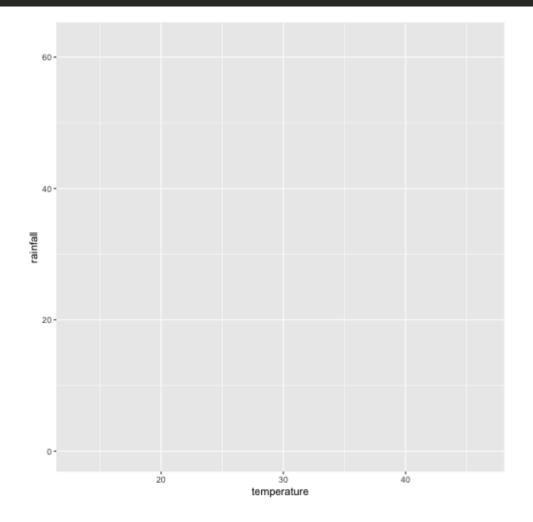
```
ggplot(
  data = beaches
)
```

# Add a mapping from data to elements

```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature,
    y = rainfall
  )
)
```

#### What goes in

- the x and y axes
- the color of markers
- the shape of markers

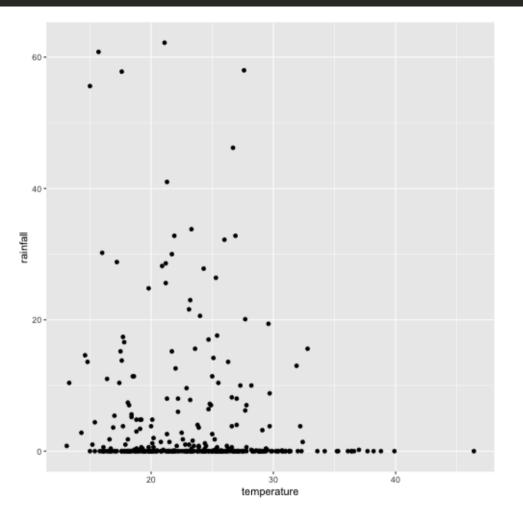


## Add a geometry to draw

```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature,
    y = rainfall
  )
) +
  geom_point()
```

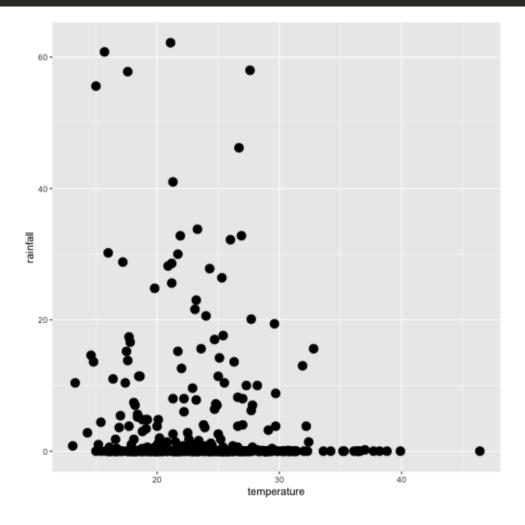
#### What to draw:

- Points, lines
- histogram, bars, pies



## Add options for the geom

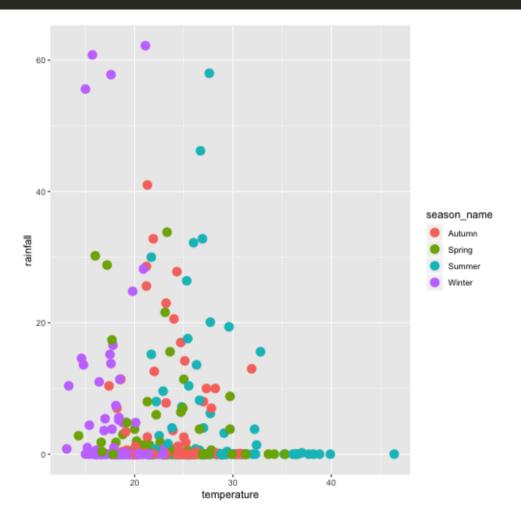
```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature,
    y = rainfall
  )
) +
  geom_point(size = 4)
```



# Add a mapping to modify the geom

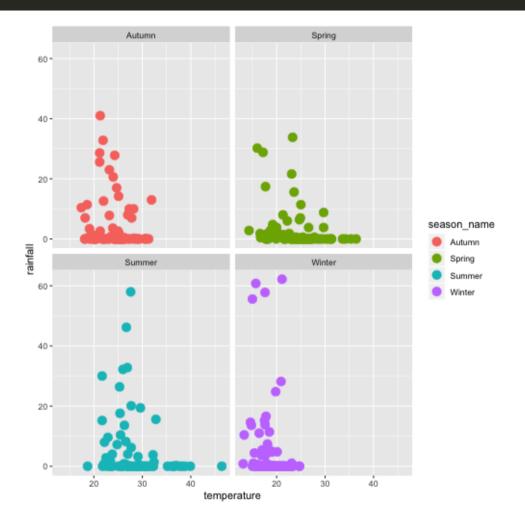
```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature,
    y = rainfall
  )
) +
  geom_point(
    mapping = aes(color = season_name),
    size = 4
  )
```

Anything data-driven has to be a mapping, driven by the aes function



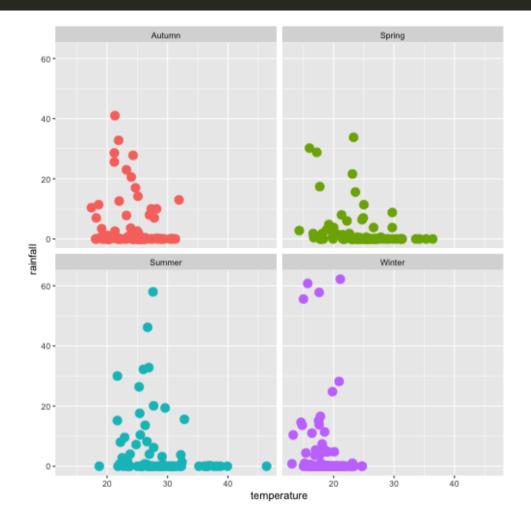
## **Split into facets**

```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature,
    y = rainfall
  )
) +
  geom_point(
    mapping = aes(color = season_name),
    size = 4
  ) +
  facet_wrap( ~ season_name)
```



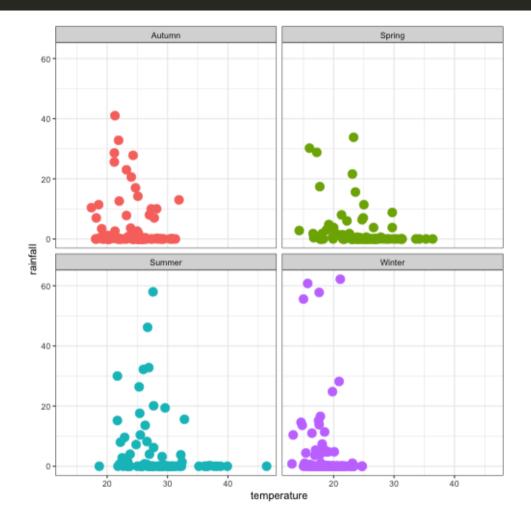
## Remove the legend

```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature,
    y = rainfall
  )
) +
  geom_point(
    mapping = aes(color = season_name),
    size = 4,
    show.legend = FALSE
  ) +
  facet_wrap( ~ season_name)
```



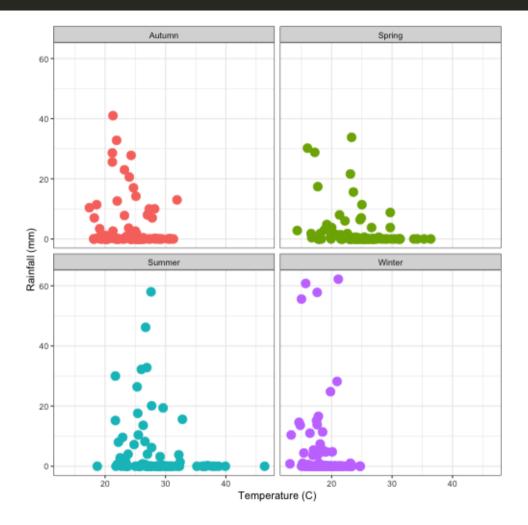
## **Change the background**

```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature,
    y = rainfall
  )
) +
  geom_point(
    mapping = aes(color = season_name),
    size = 4,
    show.legend = FALSE
  ) +
  facet_wrap( ~ season_name) +
  theme_bw()
```



### **Update the labels**

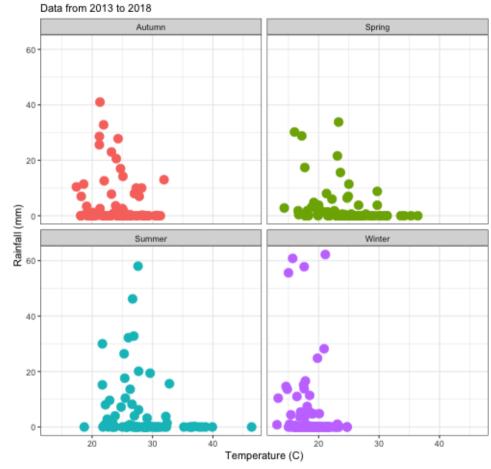
```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature,
    y = rainfall
  )
) +
  geom_point(
    mapping = aes(color = season_name),
    size = 4,
    show.legend = FALSE
) +
  facet_wrap( ~ season_name) +
  theme_bw() +
  labs(x = 'Temperature (C)', y = 'Rainfall (mm)')
```



#### **Add titles**

```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature,
    y = rainfall
  geom_point(
    mapping = aes(color = season_name),
    size = 4,
    show.legend = FALSE
  facet_wrap( ~ season_name) +
  theme_bw() +
  labs(x = 'Temperature (C)',
    y = 'Rainfall (mm)',
       title = 'Sydney weather by season',
       subtitle = "Data from 2013 to 2018")
```

#### Sydney weather by season



## The grammar

- Data
- Aesthetics (or aesthetic mappings)
- Geometries (as layers) or Statistics (as computed layers)
- Facets
- Themes
- (Coordinates)
- (Scales)

# Peeking under the hood

#### If I write...

```
ggplot(
  data = beaches,
  aes(x = temperature,
            y = rainfall)
) +
  geom_point()
```

#### what's really run is ...

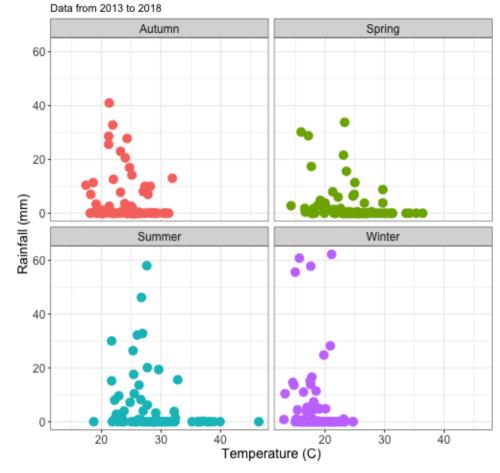
```
ggplot(
  data = beaches,
  mapping = aes(
    x = temperature, y = rainfall)) +
layer(
  geom = "point",
  stat = "identity",
  position = "identity") +
facet_null() +
theme_grey() +
coord_cartesian() +
scale_x_continuous() +
scale_y_continuous()
```

Each element can be adapted and tweaked to create graphs

#### **Customize**

```
ggplot(
  data = beaches,
 mapping = aes(
   x = temperature,
   v = rainfall
  geom_point(
   mapping = aes(color = season_name),
   size = 4,
   show.legend = FALSE
  facet_wrap( ~ season_name) +
  theme_bw() +
  labs(x = 'Temperature (C)',
      y = 'Rainfall (mm)',
      title = 'Sydney weather by season',
      subtitle = "Data from 2013 to 2018") +
  theme(axis.title = element_text(size = 14),
       axis.text = element_text(size = 12),
        strip.text = element_text(size = 12))
```

#### Sydney weather by season

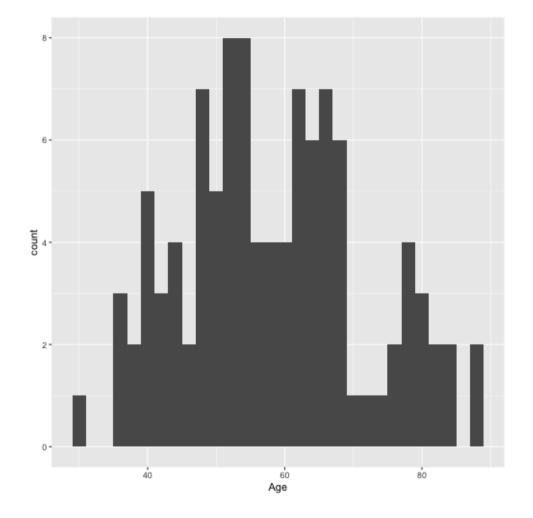


#### **Using the BRCA data**

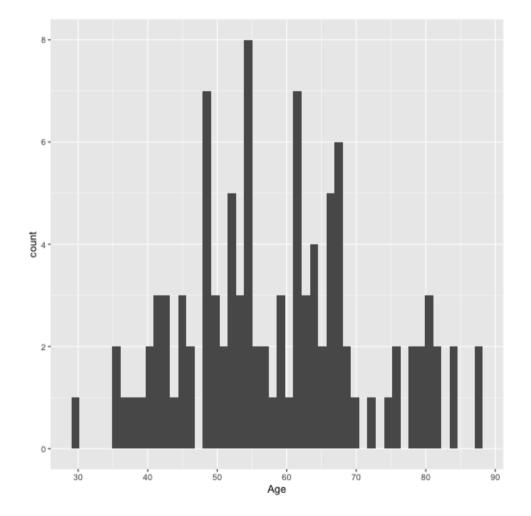
We'll use the brca data developed during the homework. The RDS file is available here.

```
brca_clean <- readRDS('data/brca.rds')
brca_clean <- brca_clean %>%
  rename('Age' = 'Age.at.Initial.Pathologic.Diagnosis')
```

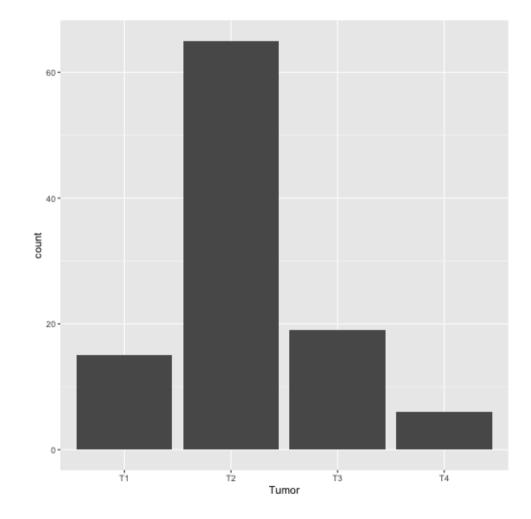
## **Univariate plots**



## **Univariate plots**

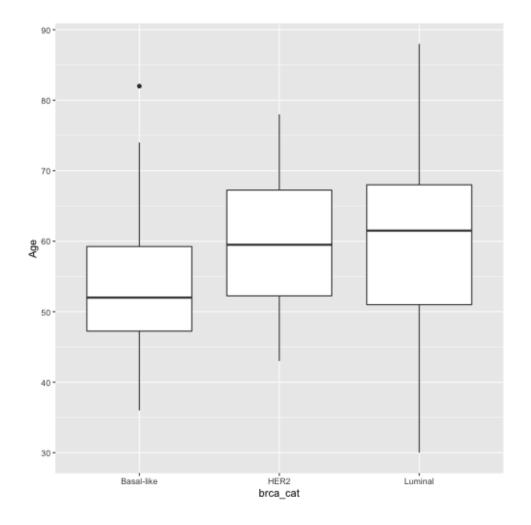


## **Univariate plots**

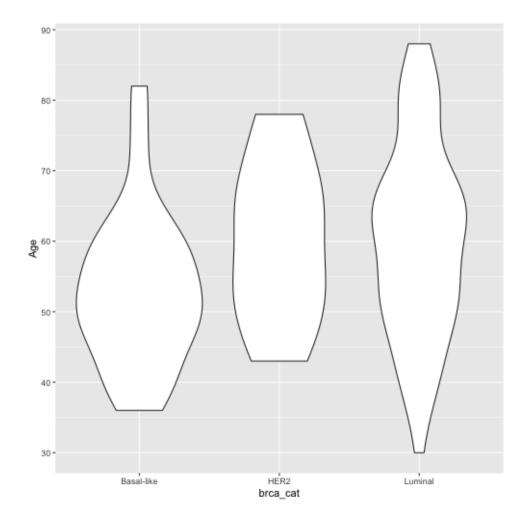


# **Bivariate plots**

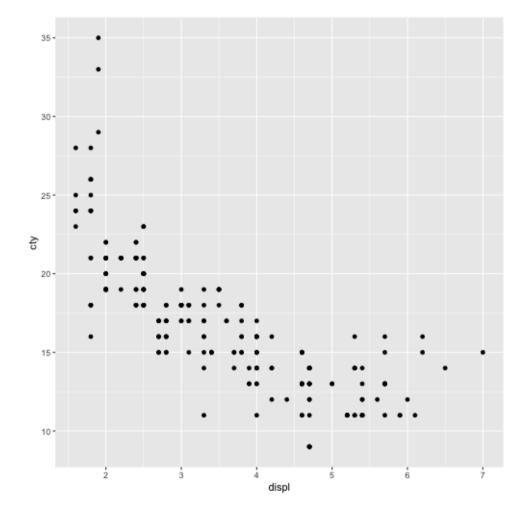
#### **Continuous with discrete**



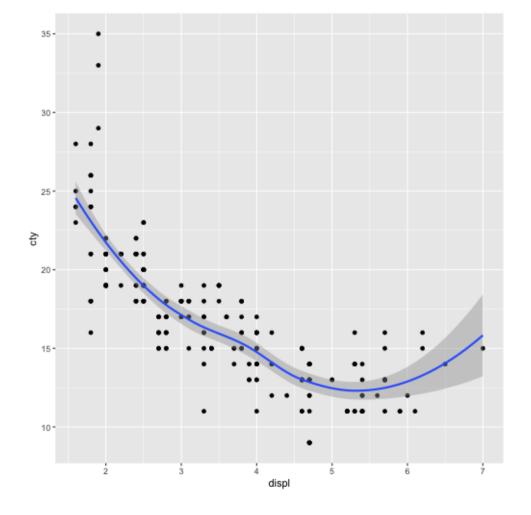
#### **Continuous with discrete**



#### Two continuous variables



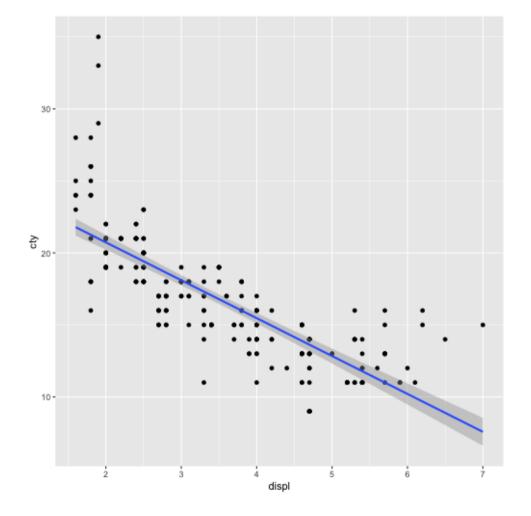
#### Two continuous variables



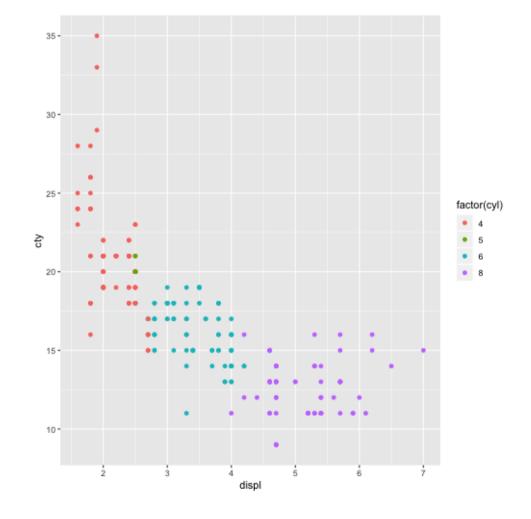
#### Two continuous variables

This forces a straight line.

1m stands for linear model



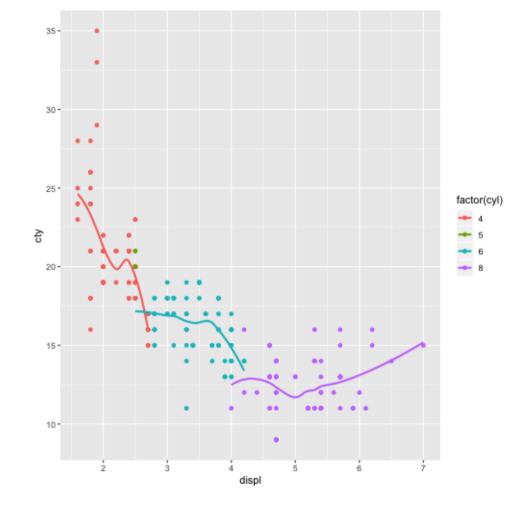
## **Adding layers**



### **Adding layers**

Separate lines for separate groups

se=F suppresses the confidence bands

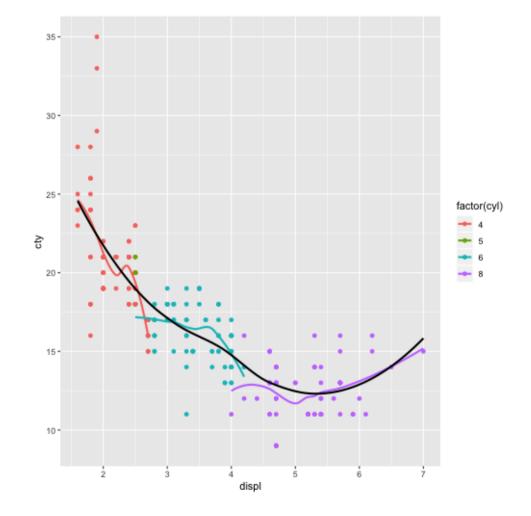


#### **Classwork checkin**

What would happen if I tried to do the previous graph without transforming cyl to factor?

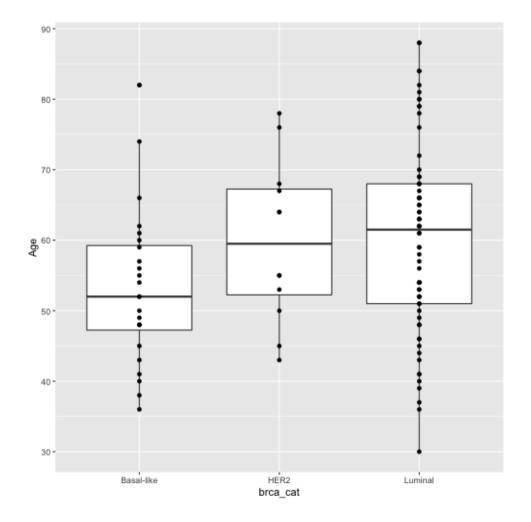
### **Adding layers**

- You can limit mappings to particular geometries
- Anything mapped from the original dataset has to be in aes()
- Anything that doesn't come from the data can be on its own



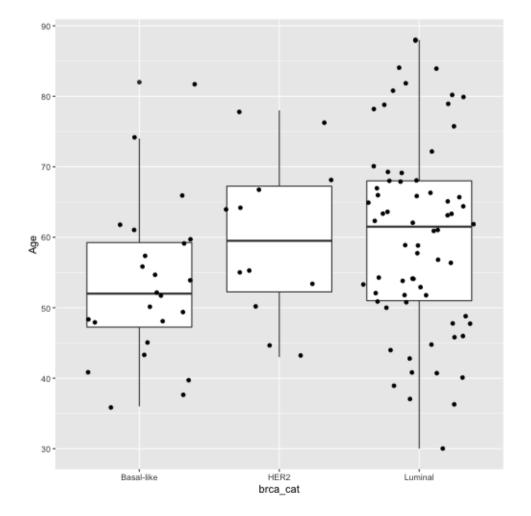
### Going back to the boxplots

Can't see the points since they are overlayed

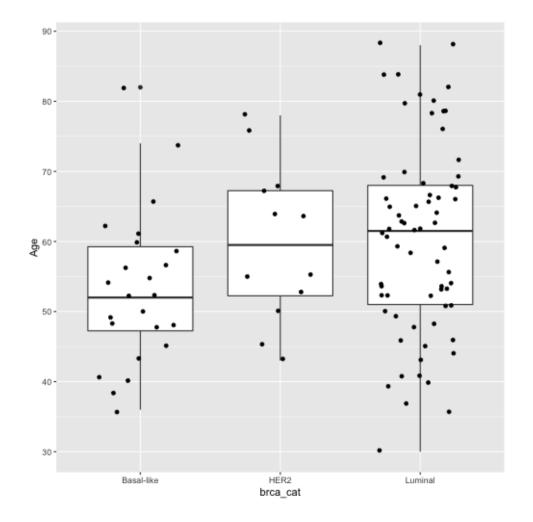


### Going back to the boxplots

• Maybe too wide?



### Going back to the boxplots



```
library(qqman)
data(gwasResults)
head(gwasResults)
```

```
#> SNP CHR BP P

#> 1 rs1 1 1 0.9148060

#> 2 rs2 1 2 0.9370754

#> 3 rs3 1 3 0.2861395

#> 4 rs4 1 4 0.8304476

#> 5 rs5 1 5 0.6417455

#> 6 rs6 1 6 0.5190959
```

```
gwasResults <- gwasResults %>%
  mutate(x_position = 1:n())
head(gwasResults)
```

```
#> SNP CHR BP P x_position

#> 1 rs1 1 1 0.9148060 1

#> 2 rs2 1 2 0.9370754 2

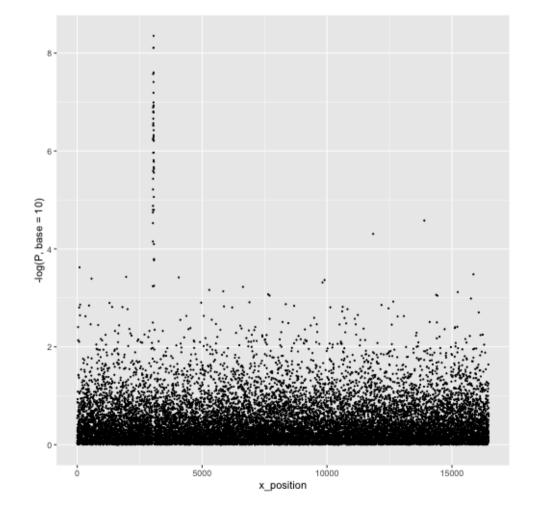
#> 3 rs3 1 3 0.2861395 3

#> 4 rs4 1 4 0.8304476 4

#> 5 rs5 1 5 0.6417455 5

#> 6 rs6 1 6 0.5190959 6
```

```
ggplot(gwasResults,
    aes(x = x_position,
        y = -log(P, base=10))
    )+
    geom_point(size = 0.2)
```

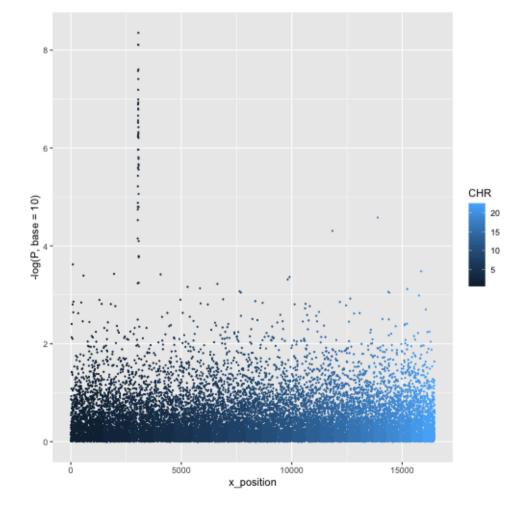


```
ggplot(gwasResults,
    aes(x = x_position,
        y = -log(P, base=10),
        group=CHR,
        color=CHR))+
    geom_point(size=0.2)
```

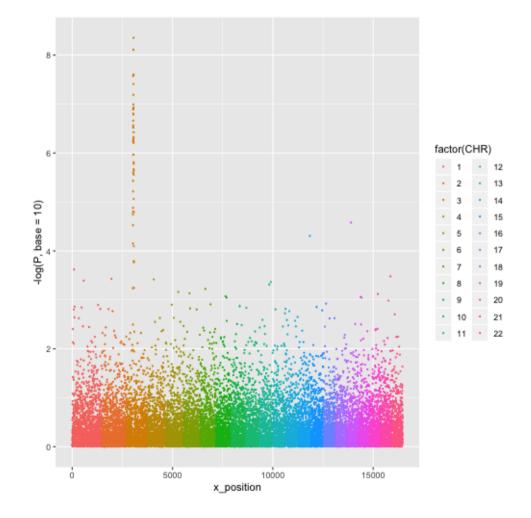
Oops!! We wanted points colored by chromosome.

But that didn't happen because we put CHR in as numeric.

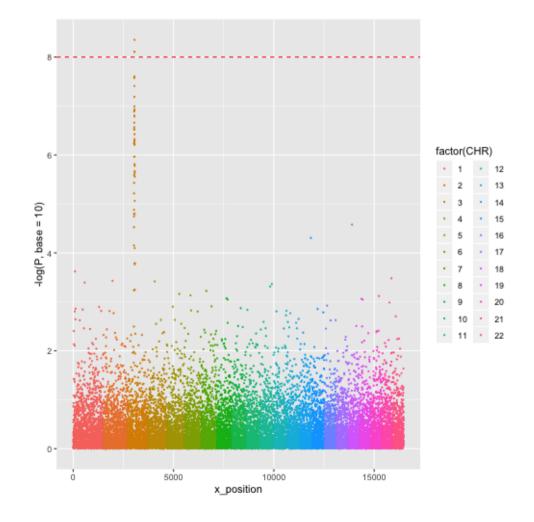
Need to convert to factor, i.e., discrete



```
ggplot(gwasResults,
    aes(x = x_position,
        y = -log(P, base=10),
        group=factor(CHR),
        color=factor(CHR)))+
    geom_point(size=0.2)
```

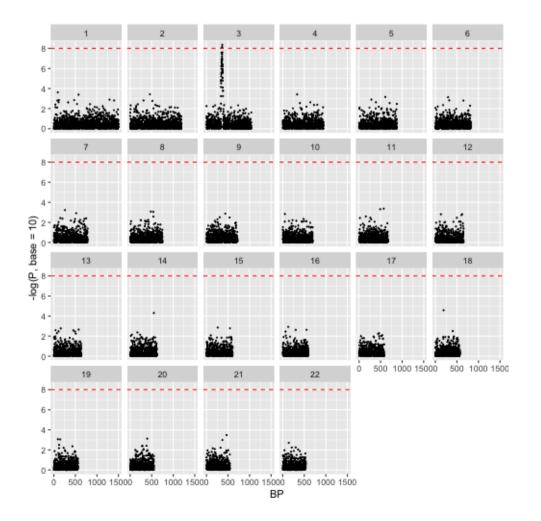


```
ggplot(gwasResults,
    aes(x = x_position,
        y = -log(P, base=10),
        group=factor(CHR), color=factor(CHR)))+
    geom_point(size=0.2)+
    geom_hline(yintercept = 8, color='red', linetype=2)
```

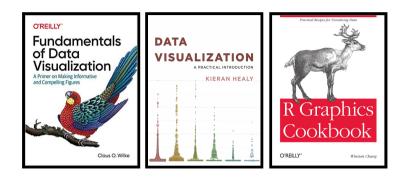


### Manhattan plot, exploded

- No more grouping variable
- A new function facet\_wrap



#### Resources



Data visualization cheatsheet (RStudio)

Chapter 3 of R4DS