dplyr, tidyr, and ggplot2
Intro to the *tidyverse*

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Part 2: ggplot2

Motivation

What is `ggplot2`?

- `ggplot` philosophy

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- `ggplot` philosophy
- Simple plots
- Some useful techniques
- More complicated plots

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- Data input centered around around `data.frames` or `tibbles`
- Data display centered around geoms (geometric objects)
- Columns from data frames are mapped into geoms using `aesthetics`
- geoms are displayed according to `themes`

Simple example - scatterplot

```
data(mtcars) # mtcars dataset (built into R)
```

```
##                    mpg cyl disp  hp drat    wt  qsec vs am gear carb
## Mazda RX4          21.0   6  160 110 3.90 2.620 16.46  0  1    4    4
## Mazda RX4 Wag      21.0   6  160 110 3.90 2.875 17.02  0  1    4    4
## Datsun 710         22.8   4  108  93 3.85 2.320 18.61  1  1    4    1
## Hornet 4 Drive     21.4   6  258 110 3.08 3.215 19.44  1  0    3    1
## Hornet Sportabout  18.7   8  360 175 3.15 3.440 17.02  0  0    3    2
```

Top line of code says: - data from mtcars
dataframe - aes = aesthetics from
dataframe - map disp to x-axis, mpg to
y-axis

```
ggplot(data = mtcars, aes(x = disp, y = mpg))+
  geom_point() # Display data using points
```
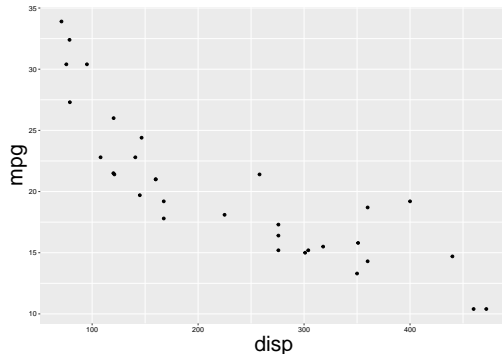
Simple example - bar plot

```
data(mtcars) # mtcars dataset (built into R)
```

```
##                    mpg cyl disp  hp drat    wt  qsec vs am gear carb
## Mazda RX4         21.0   6  160 110 3.90 2.620 16.46  0  1    4    4
## Mazda RX4 Wag     21.0   6  160 110 3.90 2.875 17.02  0  1    4    4
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## Hornet Sportabout 18.7   8  360 175 3.15 3.440 17.02  0  0    3    2
```

```
# Top line of code says:
# - map gear (number of gears) to x-axis
#   - first converted to a factor
ggplot(data = mtcars, aes(x = factor(gear)))+
  geom_bar()
# Display number of data points for each factor

# Automatically uses stat='count' to group
#   data according to factor
```
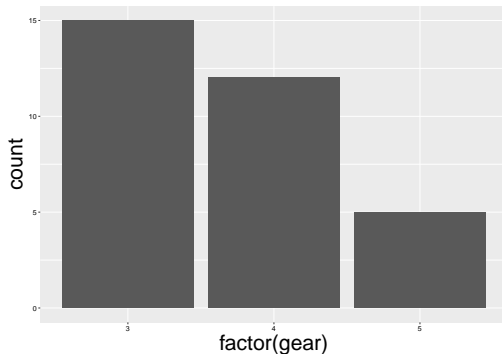
Simple example - histogram

```r
data(mtcars) # mtcars dataset (built into R)
head(mtcars,5) # Show first 5 rows
```

```
##                    mpg cyl disp  hp drat    wt  qsec vs am gear carb
## Mazda RX4         21.0   6  160 110 3.90 2.620 16.46  0  1    4    4
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## Hornet Sportabout 18.7   8  360 175 3.15 3.440 17.02  0  0    3    2
```

```r
# Top line of code says:
# - map disp (displacement) to x-axis
ggplot(data = mtcars, aes(x = disp))+
  # Group disp into bins, and display
  #   count in each bin
  geom_histogram()
```
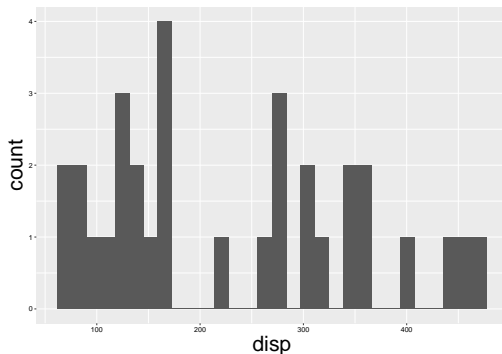
Simple example - histograms and density plots

```
# Histogram
ggplot(data=mtcars,aes(x=disp))+
  geom_histogram()
```
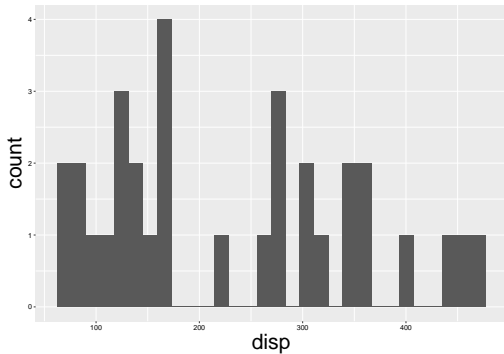

Histogram

```
# Density plot
ggplot(data=mtcars,aes(x=disp))+
  geom_density()
```
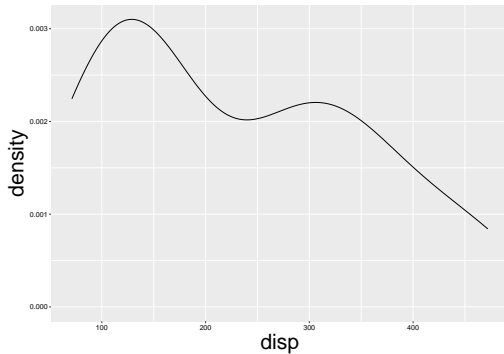

Probability density plot
$a \int_{-\infty}^{\infty} f(x)\, dx = 1$

Colours in plots

- Colours can be *mapped* (via aes) or *set* (outside of aes)

```
ggplot(data=mtcars,aes(x=disp,y=mpg))+
  # Maps gear to colour
  geom_point(aes(col=factor(gear)))
```
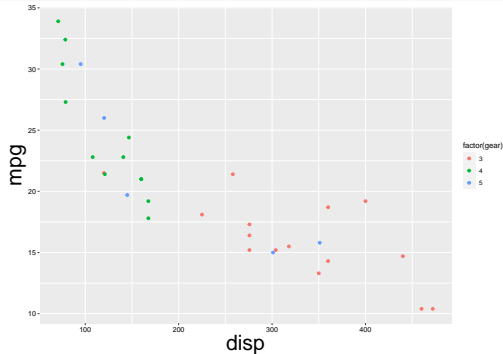


```
ggplot(data=mtcars,aes(x=disp,y=mpg))+
  geom_point(colour='red') #Sets colour
```

Colours in plots
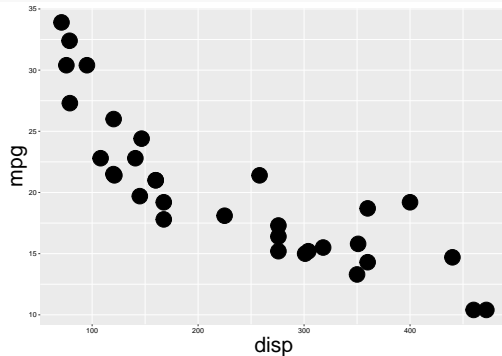
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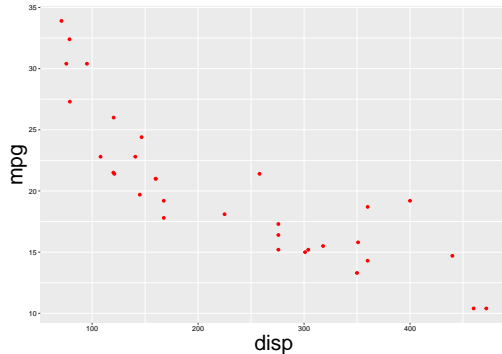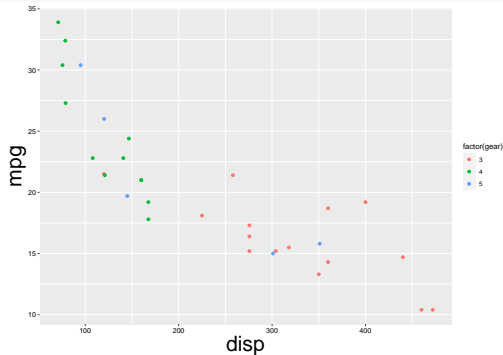
```
ggplot(data=mtcars,aes(x=disp,y=mpg))+
  geom_point(colour='red') #Sets colour
```
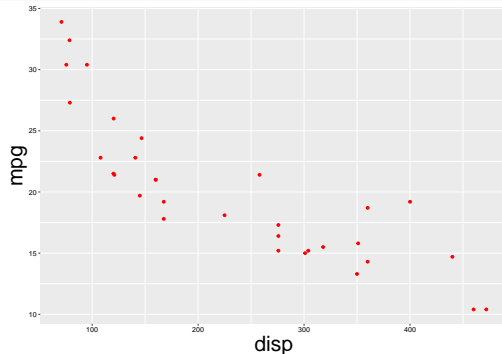

- Notice how aes was used twice in Figure 1? If used within the ggplot command, the rest of the geoms will remember it. Used within a geom, it will *update* the aesthetic

What if I want different colours?

- Default colour themes are pretty bad. Change them with `scale_colour_manual`


```
ggplot(data=mtcars,aes(x=disp,y=mpg))+
  geom_point(aes(col=factor(gear)))+
  scale_colour_manual(values=c('blue','purple','red'))
```
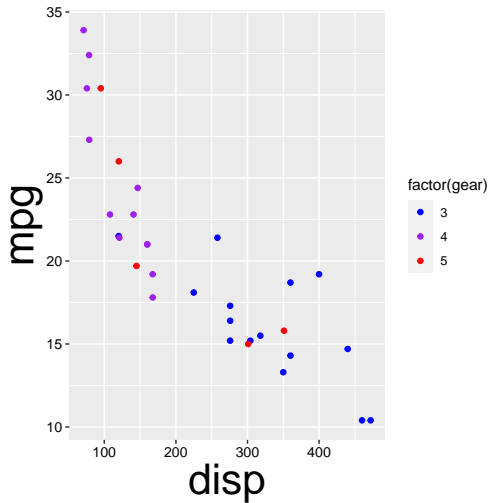
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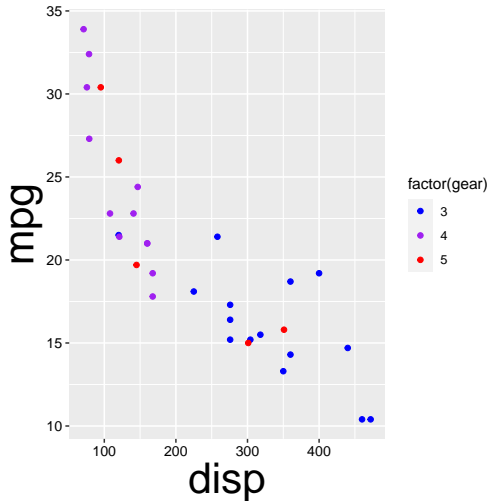
What if I want different colours?

- Default colour themes are pretty bad. Change them with `scale_colour_manual`
- Use `scale_fill_manual` for area-based colours (e.g. bar plots, polygons)
- Remember, 10% of males are red-green colourblind!
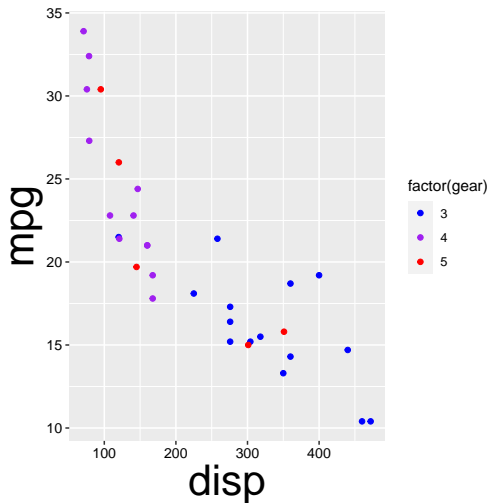

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```

Sizes in plots

- Sizes can also be *mapped* (via aes) or *set* (outside of aes)

```
ggplot(data=mtcars,aes(x=disp,y=mpg))+
  # Maps gear to size
  geom_point(aes(size=gear))
```
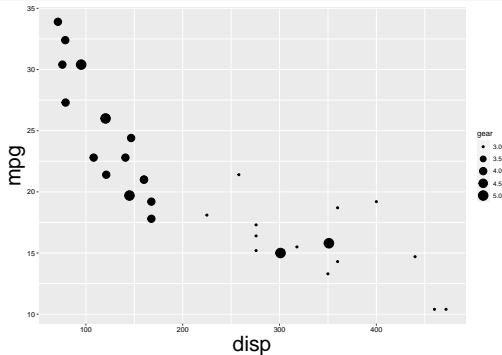


```
ggplot(data=mtcars,aes(x=disp,y=mpg))+
  geom_point(size=10) #Sets size
```

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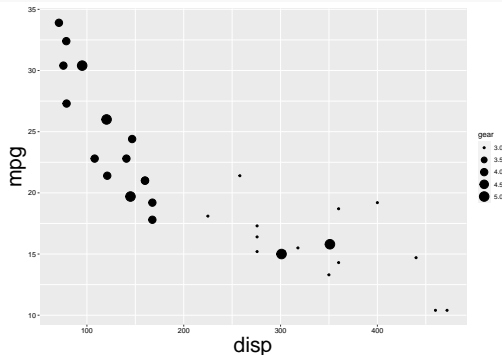


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ggplot(data=mtcars,aes(x=disp,y=mpg))+
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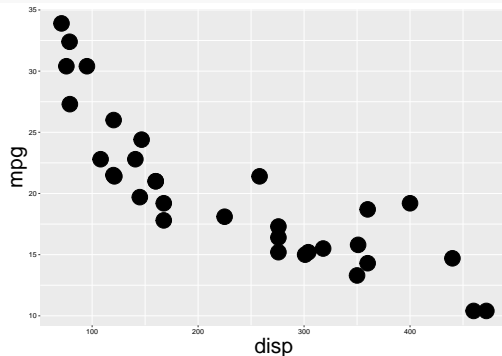

- Similar to colour choices, you can alter mapped sizes using scale_size

Change plot theme

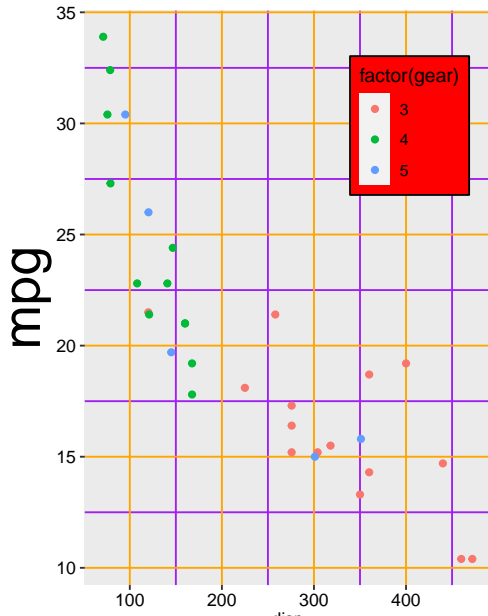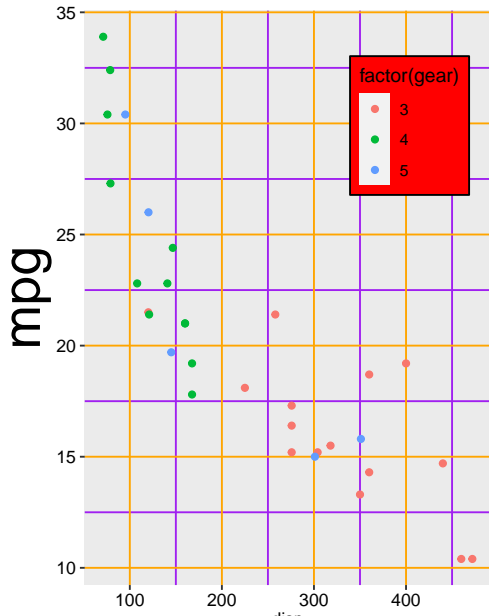- theme controls almost all non-data elements of plots

```r
ggplot(data=mtcars,aes(x=disp,y=mpg))+
  # Maps gear to colour
  geom_point(aes(col=factor(gear))) +
  #Changes plot theme
  theme(axis.title.x=element_text(size=10),
    legend.background=element_rect(fill='red'),
    legend.position=c(0.8,0.8),
    panel.grid.minor=element_line(colour='purple'),
    panel.grid.major=element_line(colour='orange'))
```
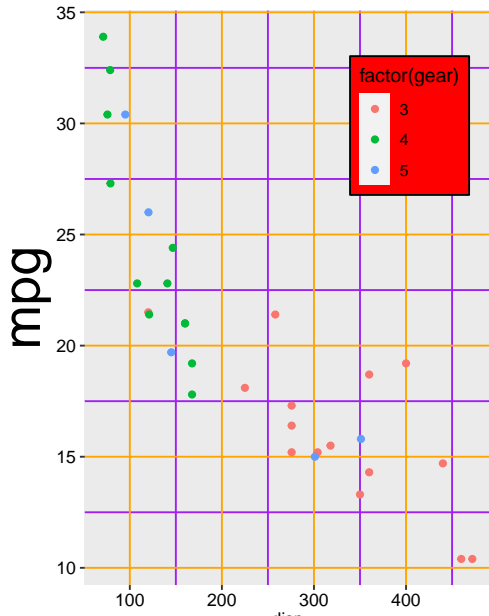
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- Made up of *elements*: element_line(), element_text(), element_rect()

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- Let's make some changes:

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Preset themes
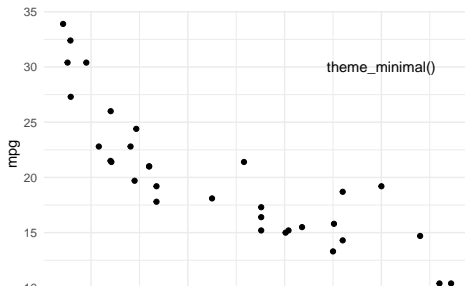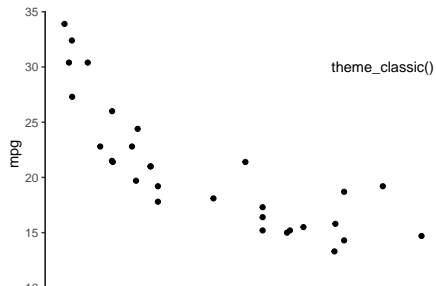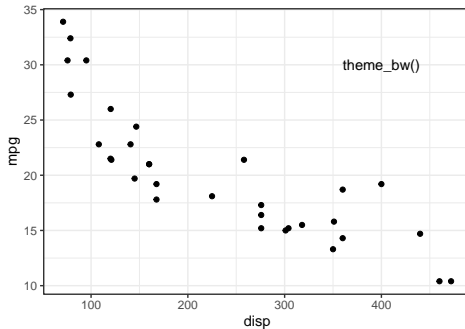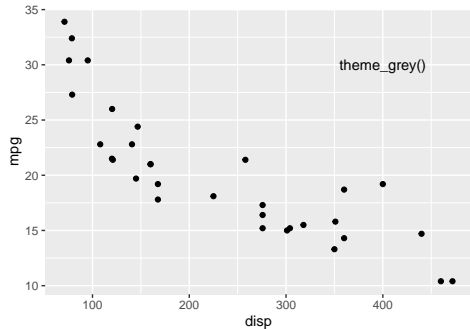
Make your own themes!

- You can modify existing themes in order to create your own

```
myTheme <- theme_classic()+ #Existing theme
  #Makes axis text bigger
  theme(axis.title=element_text(size=30),
        axis.text=element_text(size=10),
        legend.position='bottom')
#Sets up this theme as "default"
theme_set(myTheme)
```
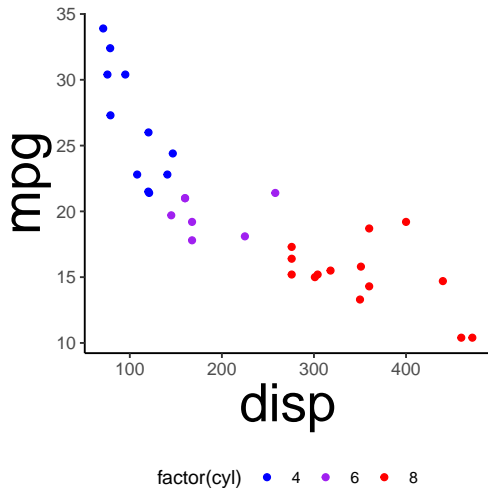
Make your own themes!

- You can modify existing themes in order to create your own
- Try using `theme_set()` at the start of your script to pre-set the theme for the rest of the script

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        axis.text=element_text(size=10),
        legend.position='bottom')
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```
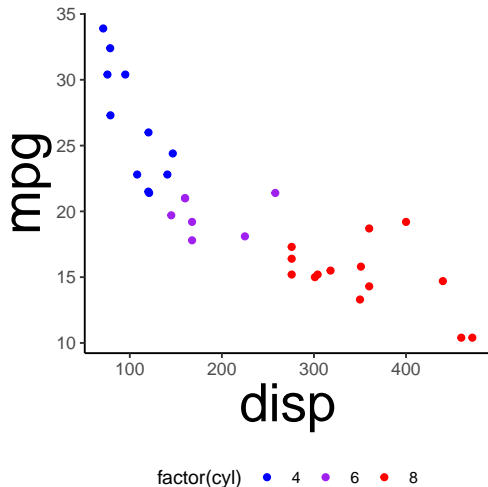
Complex plots - facets

- It is possible to break up the plot into smaller `facets` that are mapped to a given variable

```
ggplot(mtcars,aes(x=disp,y=mpg))+ geom_point(aes(col=wt))+
  facet_grid(factor(am) ~ factor(gear))
```
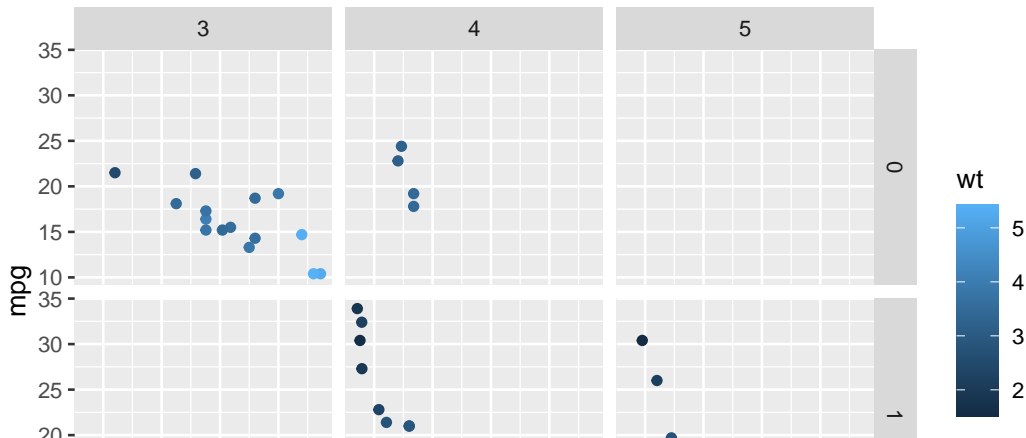
Complex plots - facets

- It is possible to break up the plot into smaller `facets` that are mapped to a given variable
- This can be combined with colour/size mappings

```
ggplot(mtcars,aes(x=disp,y=mpg))+ geom_point(aes(col=wt))+
  facet_grid(factor(am) ~ factor(gear))
```
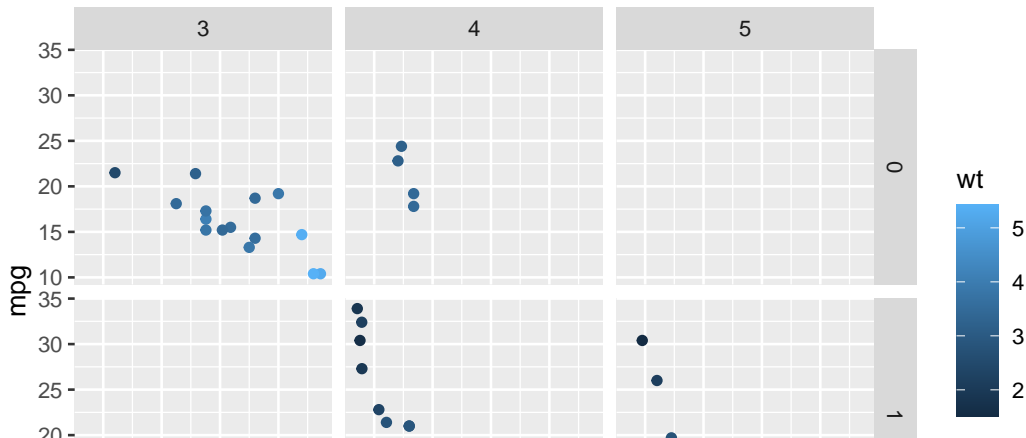
Complex plots - summary statistics (mean)

```
ggplot(mtcars,aes(x=factor(am),y=mpg))+
  geom_point(position=position_jitter(width=0.05))+ #Adds noise to data in x-dimension
  geom_point(stat='summary',fun=mean,col='red',size=3) #Mean only
```
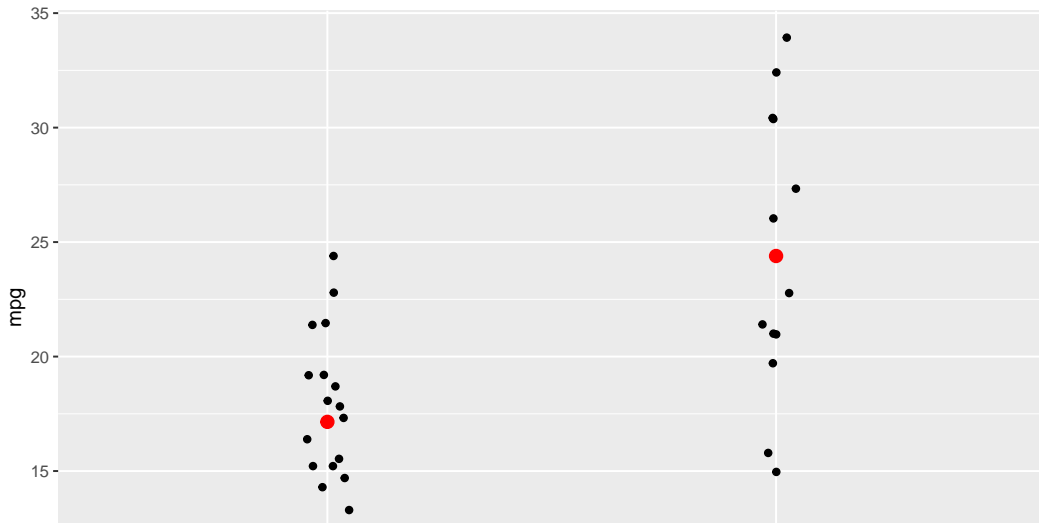
Complex plots - summary statistics (mean + SD)

```
ggplot(arrange(mtcars,am,disp),aes(x=factor(am),y=mpg))+
  geom_point(position=position_jitter(width=0.05))+
  geom_pointrange(stat='summary',fun.data=mean_se,
                  fun.args = list(mult = 2),col='red') #Mean + 2 SE
```
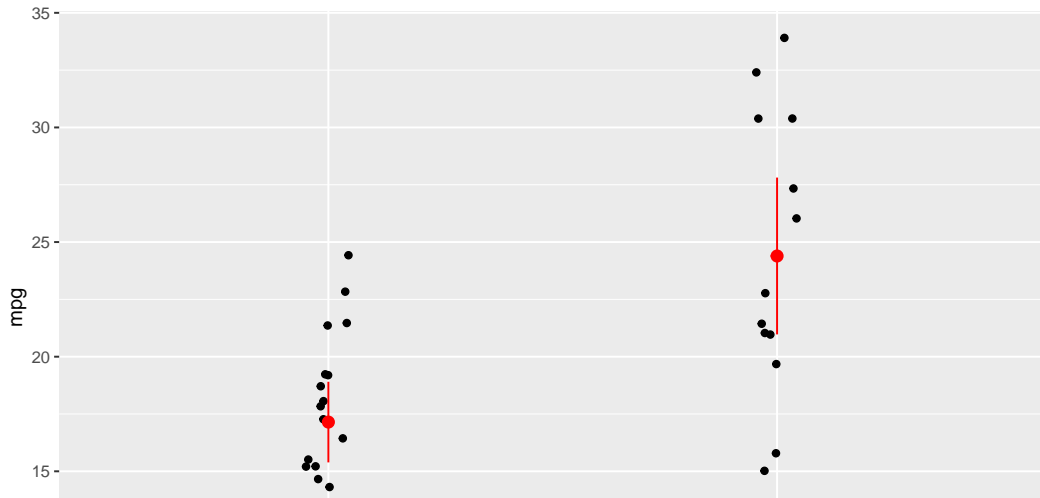
Complex plots - smoothers

- You can add `lm` (or other model) predictions to your plots:

```
ggplot(mtcars,aes(x=disp,y=mpg))+
  geom_point() +
  geom_smooth(method='lm',formula=y~x)
```
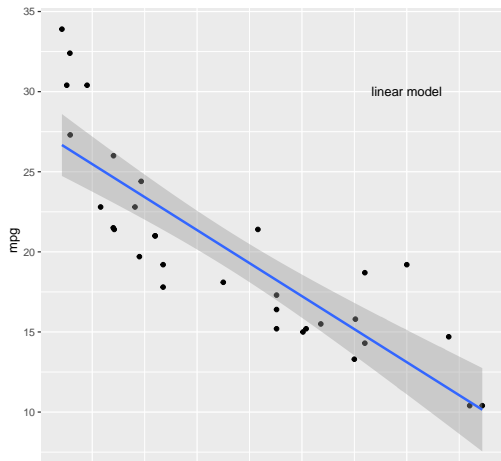
```
ggplot(mtcars,aes(x=disp,y=mpg))+
  geom_point() +
  geom_smooth(method='gam',formula=y~s(x))
```
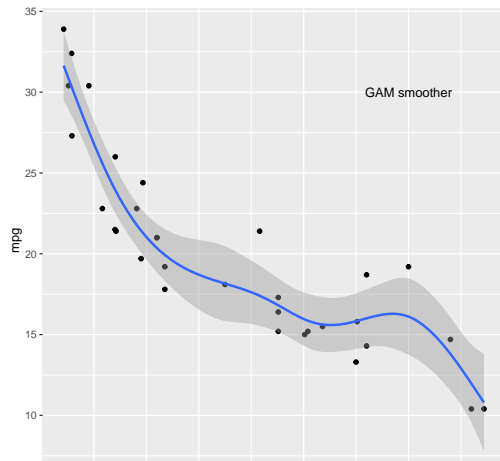
Complex plots - transformations

- You can show transformed data OR you can transform the axes themselves using `scale_*_log10` (x or y axis)

```r
ggplot(mtcars,aes(x=log(disp),y=log(mpg)))+
  geom_point() +
  geom_smooth(method='lm',formula=y~x)
# Harder to interpret, because people can't
#    usually do log(x) in their head
```
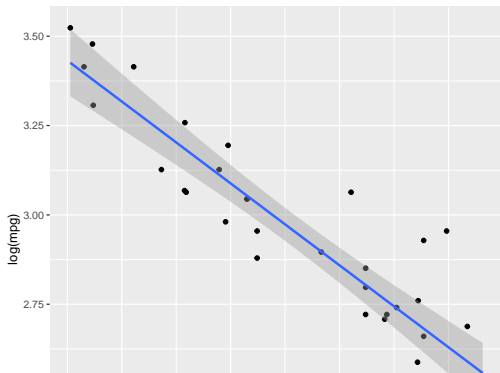


```r
ggplot(mtcars,aes(x=disp,y=mpg))+
  geom_point() +
  geom_smooth(method='lm',formula=y~x)+
  scale_x_log10() + scale_y_log10()
# sqrt is also popular
```
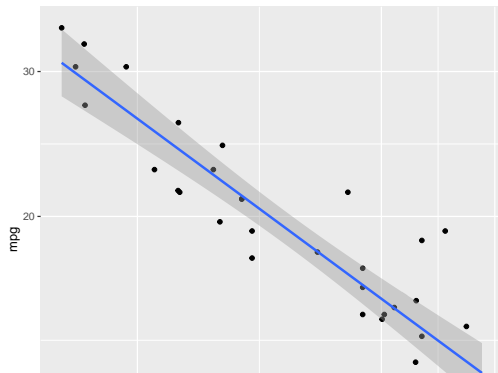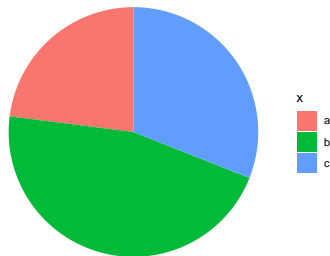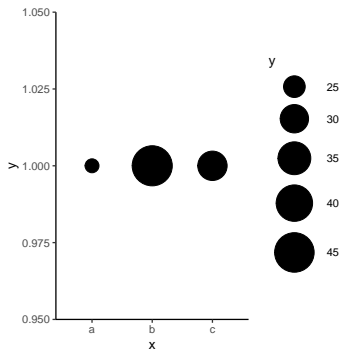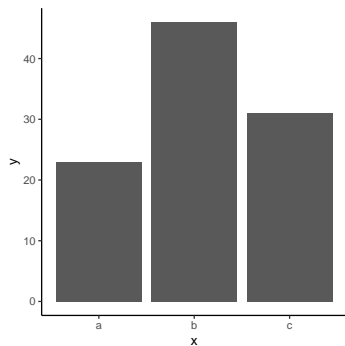
Things to remember:

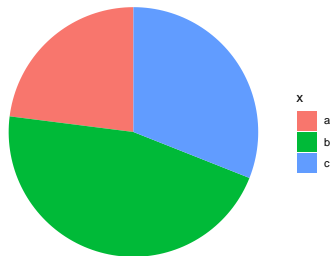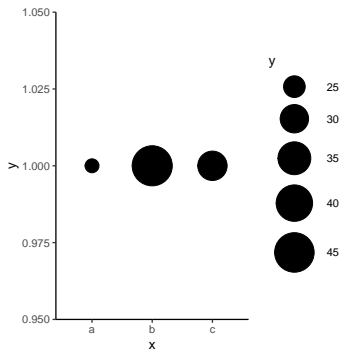- Simpler plots are often better. Try to keep it to 3 aesthetics per panel. Avoid 3D plots.
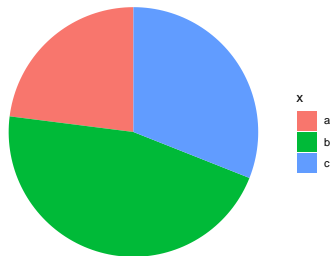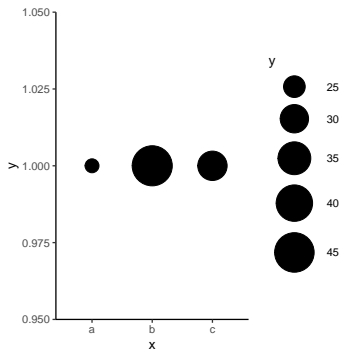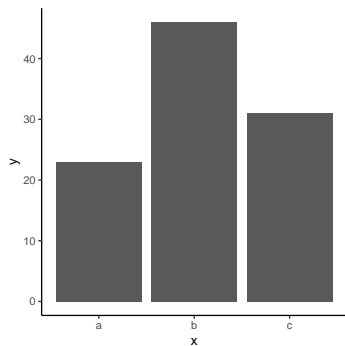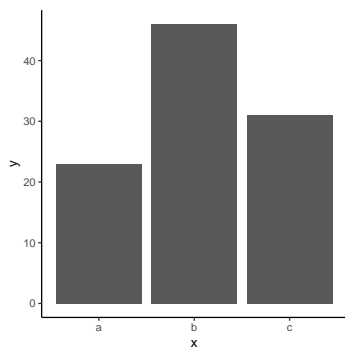
Things to remember:

- Simpler plots are often better. Try to keep it to 3 aesthetics per panel. Avoid 3D plots.
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Things to remember:

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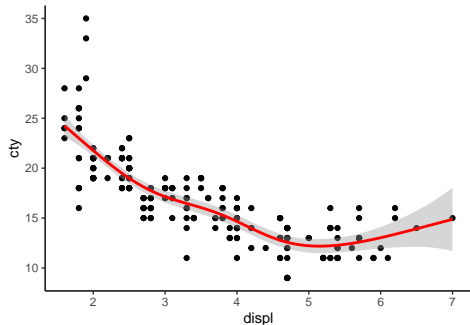- Avoid "non-data ink" (see Edward Tufte's work)
- Our eyes are good at estimating linear positions, but bad at estimating area, volume, colour shading, and angles:
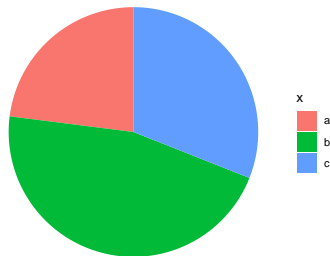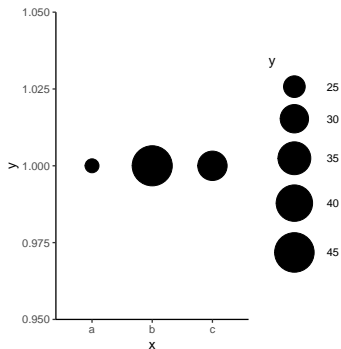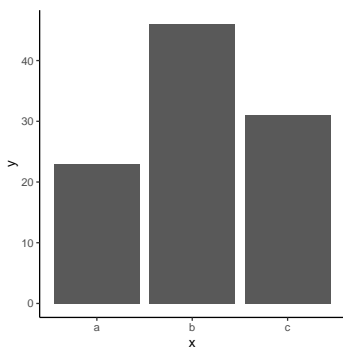
A challenger approaches:

Make these figures! Datasets are found in `mpg`, `msleep`, `trees`, and `starwars` (built into the `ggplot2` and `dplyr` packages)
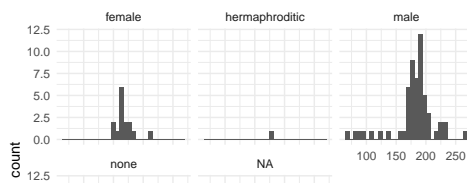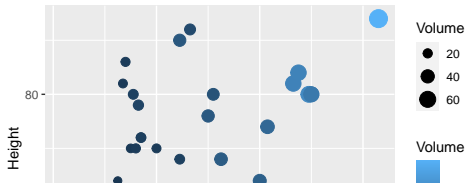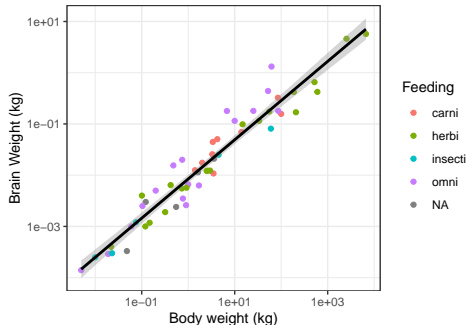
Final remarks

- dplyr & tidyr work with other parts of the tidyverse, such as ggplot2

```r
library(ggplot2)
#Code for dplyr begins here
CO2 %>% filter(conc==1000) %>%
  group_by(Type,Treatment) %>%
  summarize(meanUp=mean(uptake),
            maxUp=max(uptake),
            minUp=min(uptake)) %>%
  #Code for ggplot begins here
  ggplot(aes(x=Type,col=Treatment))+
  geom_pointrange(aes(y=meanUp,
                      ymax=maxUp,
                      ymin=minUp))+
  labs(x='Area',y='Uptake at 1000ppm')+
  scale_colour_manual(values=c('red','blue'))
```
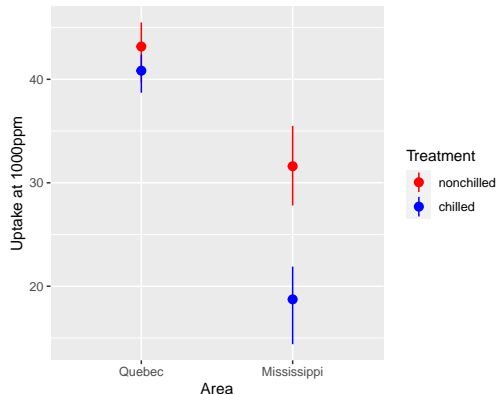
Final remarks

- dplyr & `tidyr` work with other parts of the tidyverse, such as `ggplot2`
- Example: filtered summary plot
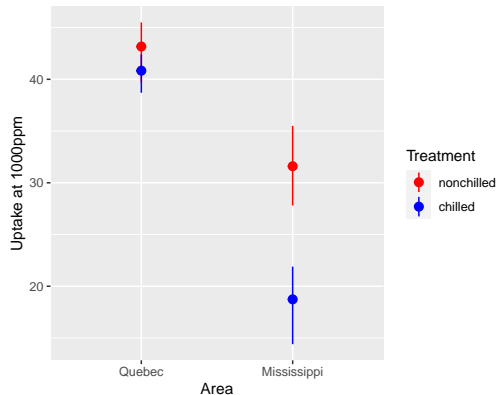
```
library(ggplot2)
#Code for dplyr begins here
CO2 %>% filter(conc==1000) %>%
  group_by(Type,Treatment) %>%
  summarize(meanUp=mean(uptake),
            maxUp=max(uptake),
            minUp=min(uptake)) %>%
  #Code for ggplot begins here
  ggplot(aes(x=Type,col=Treatment))+
  geom_pointrange(aes(y=meanUp,
                      ymax=maxUp,
                      ymin=minUp))+
  labs(x='Area',y='Uptake at 1000ppm')+
  scale_colour_manual(values=c('red','blue'))
```

Final remarks

- dplyr & tidyr can pass data frames to and from non-`tidyverse` functions: use '.' operator

```
co2mod <- CO2 %>%
  filter(Type=='Quebec') %>%
  #Code for nls begins here
  nls(uptake~SSasymp(conc,A,B,C),
      start=list(A=30,B=-15,C=-5),data=.)

data.frame(conc=seq(50,1000,20)) %>%
  predict(co2mod,newdata=.) %>%
  data.frame(conc=seq(50,1000,20),predUp=.) %>%
  #Code for ggplot begins here
  ggplot(aes(conc,predUp))+
  geom_line()+
  geom_point(data=filter(CO2,Type=='Quebec'),
             aes(conc,uptake))+
  labs(x='CO2 Concentration',y='Uptake')
```
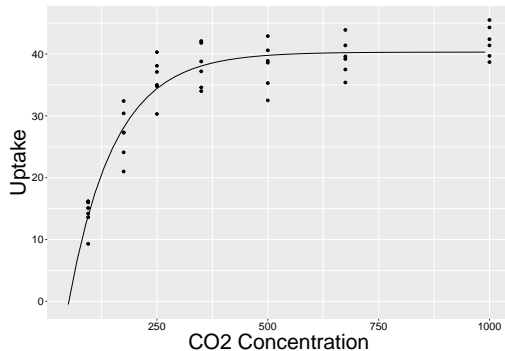
Final remarks

- dplyr & tidyr can pass data frames to and from non-`tidyverse` functions: use `.` operator
- Example: nonlinear growth model

```r
co2mod <- CO2 %>%
  filter(Type=='Quebec') %>%
  #Code for nls begins here
  nls(uptake~SSasymp(conc,A,B,C),
      start=list(A=30,B=-15,C=-5),data=.)

data.frame(conc=seq(50,1000,20)) %>%
  predict(co2mod,newdata=.) %>%
  data.frame(conc=seq(50,1000,20),predUp=.) %>%
  #Code for ggplot begins here
  ggplot(aes(conc,predUp))+
  geom_line()+
  geom_point(data=filter(CO2,Type=='Quebec'),
             aes(conc,uptake))+
  labs(x='CO2 Concentration',y='Uptake')
```