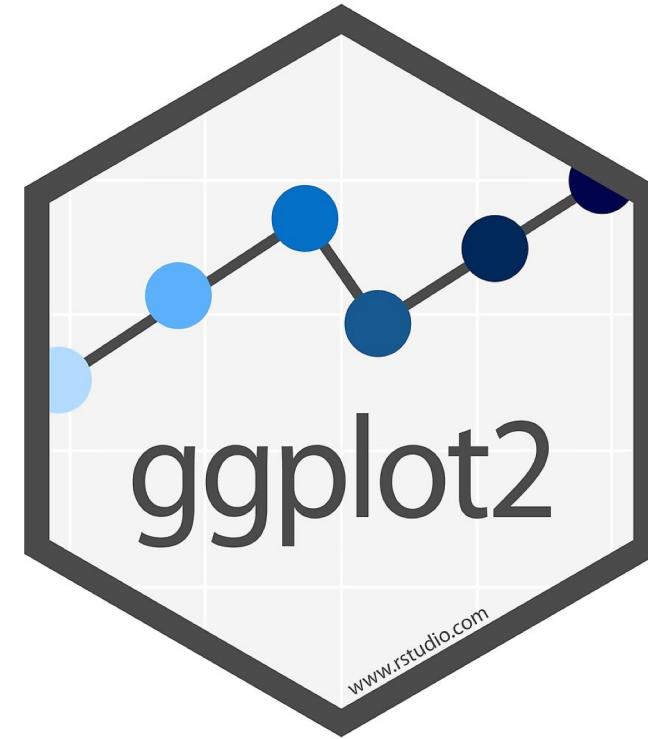




Environmental Computing

Better figures with **ggplot2**



Fonti Kar

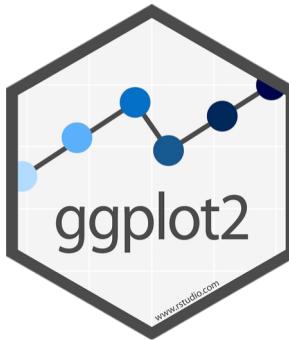


@fonti_kar



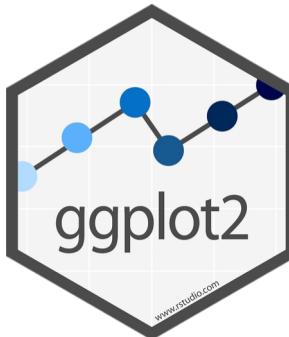


VS.





VS.



- `plot()` and `par()`
- Error bars are a nightmare
- Long chunky scripts
- Very specific functions/packages for different plots e.g. `lattice::`

- **Tidy, intuitive coding**
- **Many handy built-in functions**
- **Less time spent on grumpy coding**
- **The ‘profesh’ look**
- Complex customising (if you need it)

The recipe

```
ggplot(data, aes(x_var, y_var))  
+ geom_plot()  
+ theme_()
```

'aesthetics' – what you want to be depict in your plot

what type of plot do you want?

the 'look' of the plot and other nitty gritty things
(e.g. axes, font size, legend position)

An example

```
ggplot(iris, aes(x = Species, y = Petal.Length))
```

```
+ geom_boxplot()
```

```
+ theme_bw()
```

the 'look' of the plot and other nitty gritty things
(e.g. axes, font size, legend position)

'aesthetics' – what you want to be depict in your plot

what type of plot do you want?

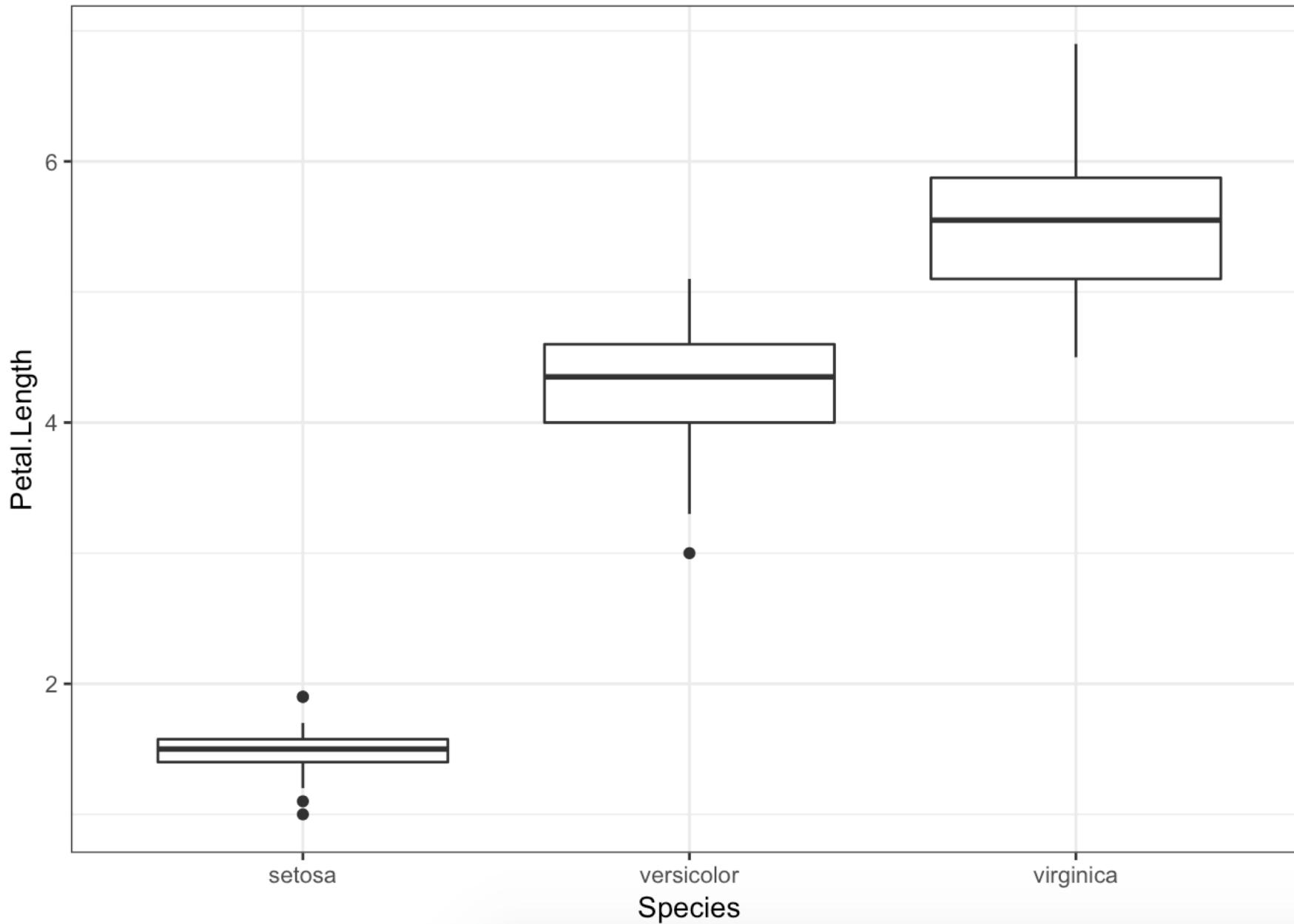


Iris Versicolor

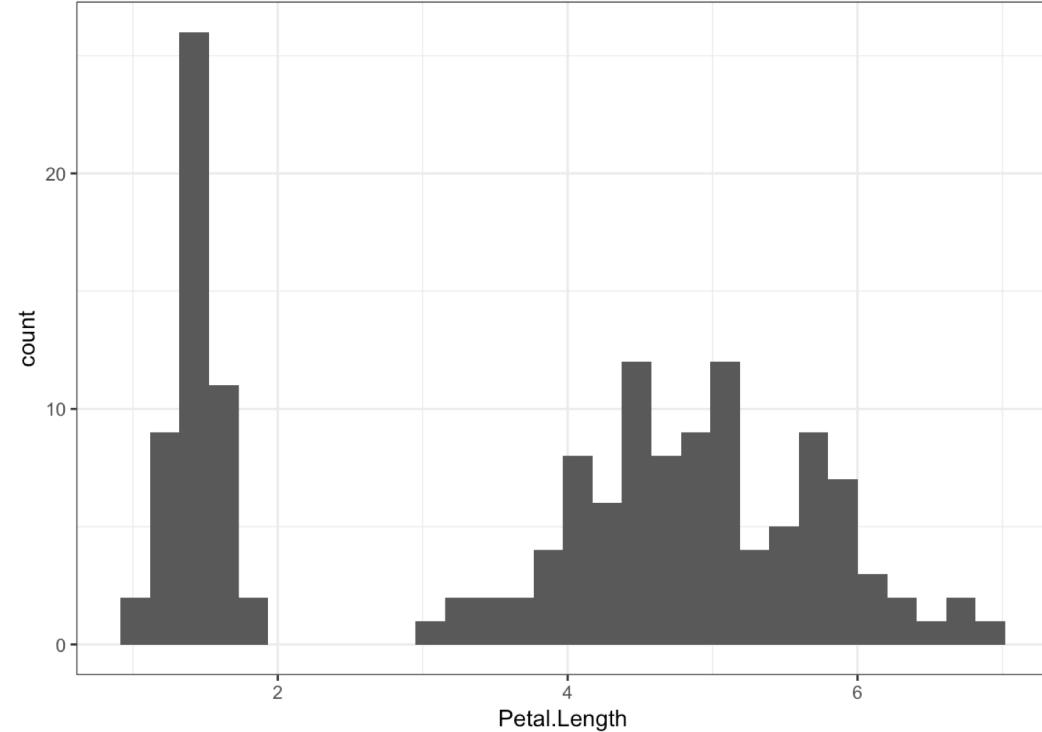
Iris Setosa

Iris Virginica

```
ggplot(iris, aes(x = Species, y = Petal.Length)) + geom_boxplot() + theme_bw()
```

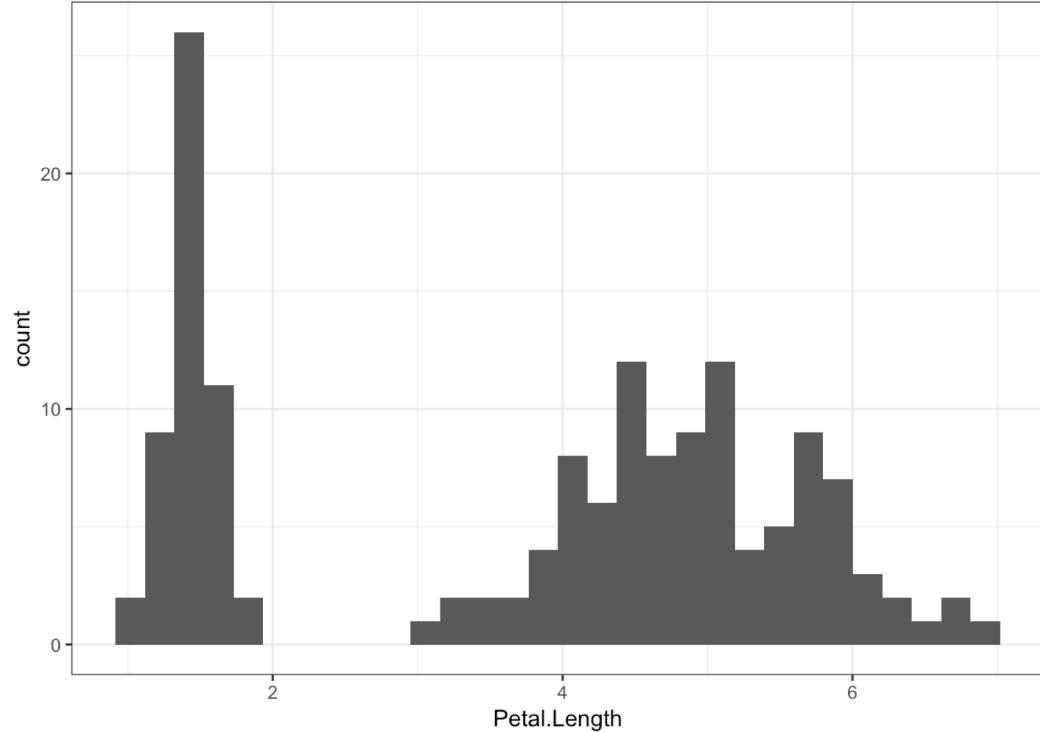


Some other key plots

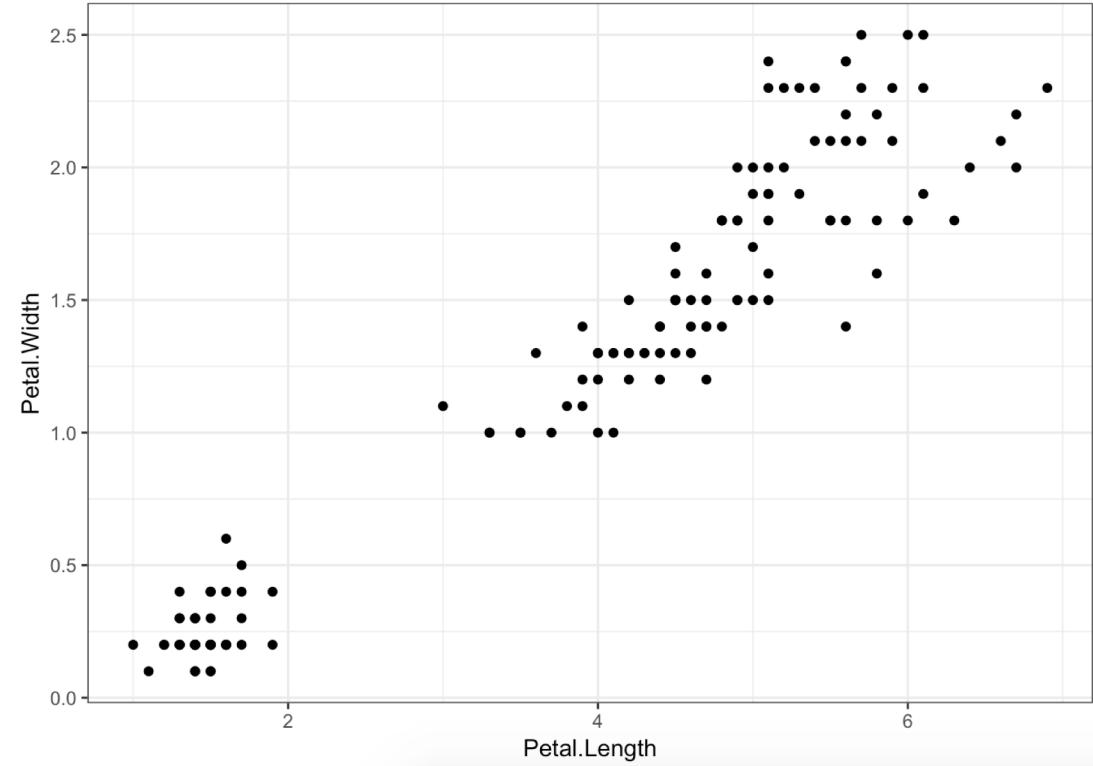


```
ggplot(iris, aes(x = Petal.Length))  
+ geom_histogram()  
+ theme_bw()
```

Some other key plots

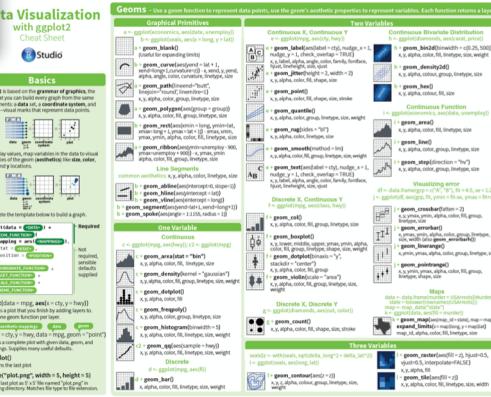


```
ggplot(iris, aes(x = Petal.Length))  
+ geom_histogram()  
+ theme_bw()
```



```
ggplot(iris, aes(x = Petal.Length,  
y = Petal.Width))  
+ geom_point()  
+ theme_bw()
```

Geom bonanza!



Get the cheat sheet!

Geoms - Use a geom function to represent data points, use the geom's aesthetic properties to represent variables. Each function returns a layer.

Graphical Primitives

- a <- ggplot(economics, aes(date, unemploy))
b <- ggplot(seals, aes(x = long, y = lat))
- a + **geom_blank()**
(Useful for expanding limits)
- b + **geom_curve(aes(yend = lat + 1, xend = long + 1, curvature = z))** - x, yend, y, yend, alpha, angle, color, curvature, linetype, size
- a + **geom_path(lineend = "butt", linejoin = "round", linemitre = 1)**
x, y, alpha, color, group, linetype, size
- a + **geom_polygon(aes(group = group))**
x, y, alpha, color, fill, group, linetype, size
- b + **geom_rect(aes(xmin = long, ymin = lat, xmax = long + 1, ymax = lat + 1))** - xmax, xmin, ymax, ymin, alpha, color, fill, linetype, size
- a + **geom_ribbon(aes(ymin = unemploy - 900, ymax = unemploy + 900))** - x, ymax, ymin, alpha, color, fill, group, linetype, size

Line Segments

- common aesthetics: x, y, alpha, color, linetype, size
- b + **geom_abline(aes(intercept = 0, slope = 1))**
 - b + **geom_hline(aes(yintercept = lat))**
 - b + **geom_vline(aes(xintercept = long))**
 - b + **geom_segment(aes(yend = lat + 1, xend = long + 1))**
 - b + **geom_spoke(aes(angle = 1:1155, radius = 1))**

One Variable

Continuous

- c <- ggplot(mpg, aes(hwy)); c2 <- ggplot(mpg)
- c + **geom_area(stat = "bin")**
x, y, alpha, color, fill, linetype, size
- c + **geom_density(kernel = "gaussian")**
x, y, alpha, color, fill, group, linetype, size, weight
- c + **geom_dotplot()**
x, y, alpha, color, fill
- c + **geom_freqpoly()**
x, y, alpha, color, group, linetype, size
- c + **geom_histogram(binwidth = 5)**
x, y, alpha, color, fill, linetype, size, weight
- c2 + **geom_qq(aes(sample = hwy))**
x, y, alpha, color, fill, linetype, size, weight

Discrete

- d <- ggplot(mpg, aes(f1))
- d + **geom_bar()**
x, alpha, color, fill, linetype, size, weight

Two Variables

Continuous X, Continuous Y

- e <- ggplot(mpg, aes(cty, hwy))
- e + **geom_label(aes(label = cty, nudge_x = 1, nudge_y = 1, check_overlap = TRUE))**
x, y, label, alpha, angle, color, family, fontface, hjust, lineheight, size, vjust
- e + **geom_jitter(height = 2, width = 2)**
x, y, alpha, color, fill, shape, size
- e + **geom_point()**
x, y, alpha, color, fill, shape, size, stroke
- e + **geom_quantile()**
x, y, alpha, color, group, linetype, size, weight
- e + **geom_rug(sides = "bl")**
x, y, alpha, color, linetype, size
- e + **geom_smooth(method = lm)**
x, y, alpha, color, fill, group, linetype, size, weight
- e + **geom_text(aes(label = cty, nudge_x = 1, nudge_y = 1, check_overlap = TRUE))**
x, y, label, alpha, angle, color, family, fontface, hjust, lineheight, size, vjust

Discrete X, Continuous Y

- f <- ggplot(mpg, aes(class, hwy))

- f + **geom_col()**
x, y, max, ymin, alpha, color, fill, group, linetype, size
- f + **geom_boxplot()**
x, y, lower, middle, upper, ymax, ymin, alpha, color, fill, group, linetype, shape, size, weight
- f + **geom_dotplot(binaxis = "y", stackdir = "center")**
x, y, alpha, color, fill, group
- f + **geom_violin(scale = "area")**
x, y, alpha, color, fill, group, linetype, size, weight

Discrete X, Discrete Y

- g <- ggplot(diamonds, aes(cut, color))

- g + **geom_count()**
x, y, alpha, color, fill, shape, size, stroke

Three Variables

- seals\$z <- with(seals, sqrt(delta_long^2 + delta_lat^2))
- l <- ggplot(seals, aes(long, lat))

- l + **geom_raster(aes(fill = z), hjust = 0.5, vjust = 0.5, interpolate = FALSE)**
x, y, alpha, fill
- l + **geom_contour(aes(z = z))**
x, y, z, alpha, colour, group, linetype, size, weight
- l + **geom_tile(aes(fill = z))**
x, y, alpha, color, fill, linetype, size, width

Continuous Bivariate Distribution

- h <- ggplot(diamonds, aes(carat, price))
- h + **geom_bin2d(binwidth = c(0.25, 500))**
x, y, alpha, color, fill, linetype, size, weight
- h + **geom_density2d()**
x, y, alpha, colour, group, linetype, size
- h + **geom_hex()**
x, y, alpha, colour, fill, size

Continuous Function

- i <- ggplot(economics, aes(date, unemploy))
- i + **geom_area()**
x, y, alpha, color, fill, linetype, size
- i + **geom_line()**
x, y, alpha, color, group, linetype, size
- i + **geom_step(direction = "hv")**
x, y, alpha, color, group, linetype, size

Visualizing error

- df <- data.frame(grp = c("A", "B"), fit = 4:5, se = 1:2)
- j <- ggplot(df, aes(grp, fit, ymin = fit - se, ymax = fit + se))

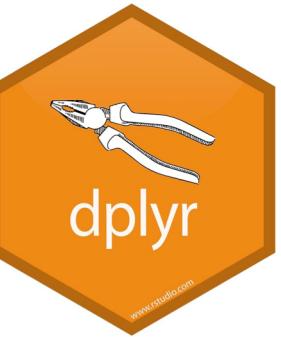
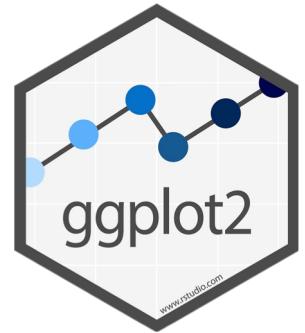
- j + **geom_crossbar(fatten = 2)**
x, y, max, ymin, alpha, color, fill, group, linetype, size
- j + **geom_errorbar()**
x, y, max, ymin, alpha, color, group, linetype, size, width (also **geom_errorbarh()**)
- j + **geom_linerange()**
x, ymin, ymax, alpha, color, group, linetype, size
- j + **geom_pointrange()**
x, y, ymin, ymax, alpha, color, fill, group, linetype, shape, size

Maps

- data <- data.frame(murder = USAArrests\$Murder, state = tolower(rrownames(USAArrests)))
- map <- map_data("state")
- k <- ggplot(data, aes(fill = murder))
- k + **geom_map(aes(map_id = state), map = map) + expand_limits(x = map\$long, y = map\$lat)**
map_id, alpha, color, fill, linetype, size

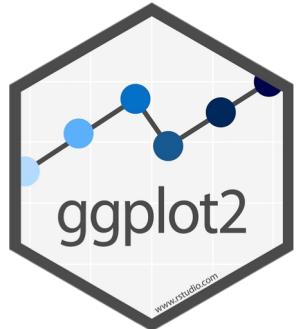
Three Variables

- l + **geom_raster(aes(fill = z), hjust = 0.5, vjust = 0.5, interpolate = FALSE)**
x, y, alpha, fill
- l + **geom_contour(aes(z = z))**
x, y, z, alpha, colour, group, linetype, size, weight
- l + **geom_tile(aes(fill = z))**
x, y, alpha, color, fill, linetype, size, width



Wrangle data and plot it!

I want plot of **mean** Petal.Length and
standard error bars for each species

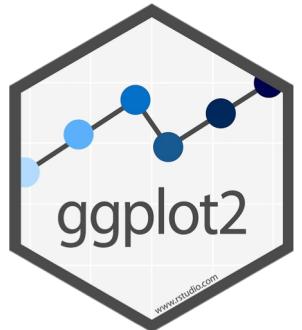


Wrangle data and plot it!

I want plot of **mean** Petal.Length and
standard error bars for each species

1) Compute what you want

```
iris.barplot <- iris %>% group_by(Species) %>%  
summarise(meanPetal.Length = mean(Petal.Length),  
         stderPetal.Length = sd(Petal.Length)/sqrt(length(Petal.Length)),  
         lowerErr = (meanPetal.Length - stderPetal.Length),  
         upperErr = (meanPetal.Length + stderPetal.Length))
```

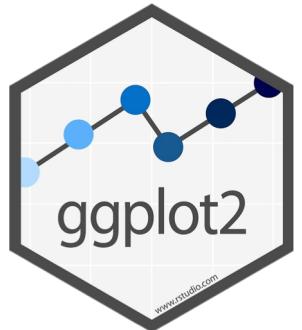


Wrangle data and plot it!

I want plot of **mean** Petal.Length and
standard error bars for each species

2) Check your data

```
> iris.barplot
# A tibble: 3 × 5
  Species    meanPetal.Length stdErrorPetal.Length lowerErr upperErr
  <fct>            <dbl>             <dbl>           <dbl>      <dbl>
1 setosa            1.46             0.0246          1.44       1.49
2 versicolor        4.26             0.0665          4.19       4.33
3 virginica         5.55             0.0780          5.47       5.63
```

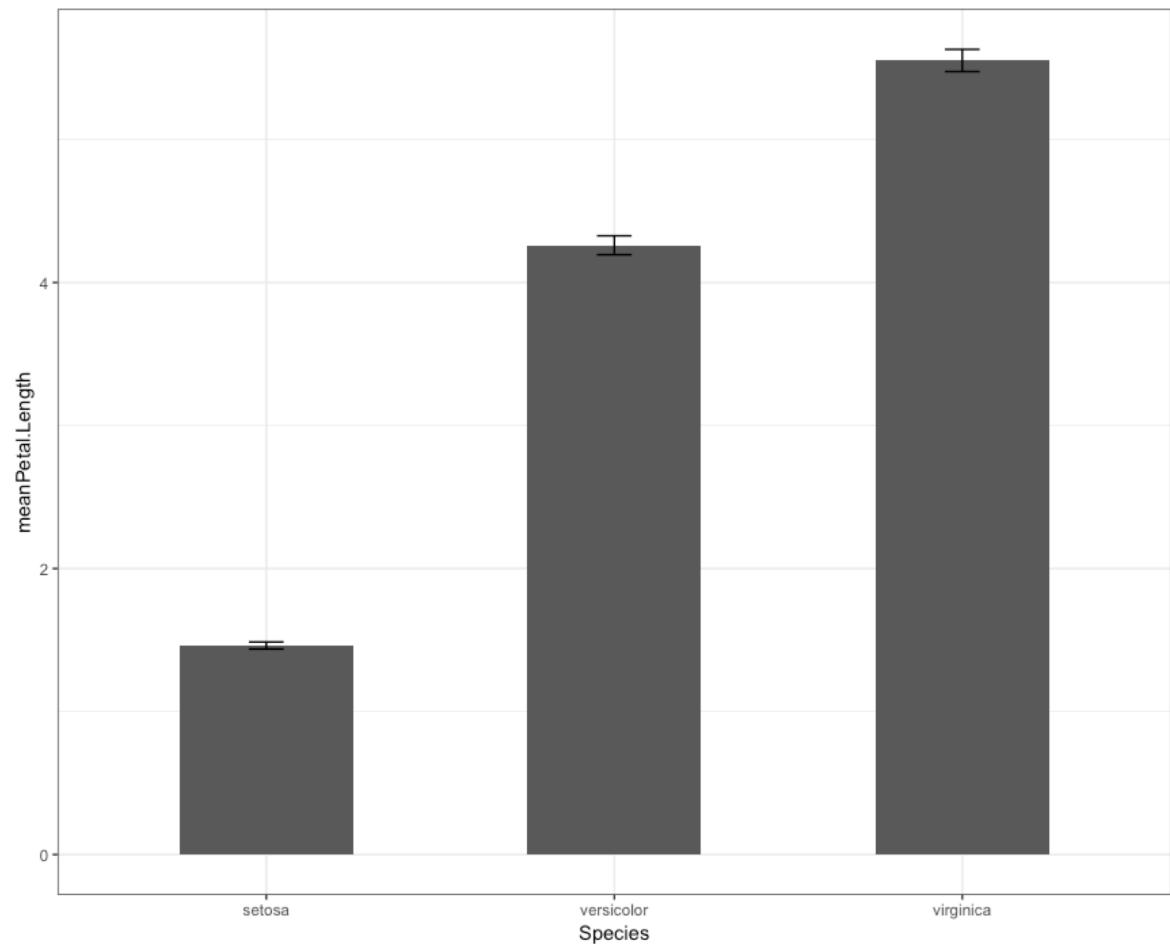


Wrangle data and plot it!

I want plot of **mean** Petal.Length and
standard error bars for each species

3) Plot your data

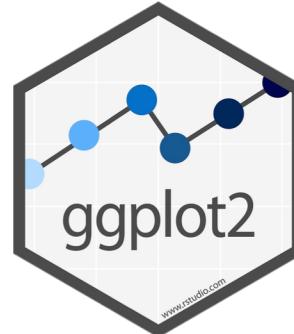
```
ggplot(iris.barplot,  
aes(x = Species,  
    y = meanPetal.Length))  
+ geom_col()  
+ geom_errorbar(  
aes(ymin = lowerErr,  
    ymax = upperErr),  
width = 0.2)  
+ theme_bw()
```



A quick comparison between



VS.



```
meanPetal.length <- tapply(iris$Petal.Length, iris$Species,  
mean)  
nPetal.length <- tapply(iris$Petal.Length, iris$Species,  
length)  
sePetal.length <- tapply(iris$Petal.Length,  
iris$Species, sd)/sqrt(nPetal.length)
```

WARNING: GRUMPY CODING AHEAD

```
xx <- barplot(meanPetal.length,  
ylim=c(0,max(meanPetal.length+sePetal.length)*1.1))  
  
arrows(xx,meanPetal.length-sePetal.length,  
xx, meanPetal.length+sePetal.length,  
angle=90,code=3,length = 0)
```

```
box()
```

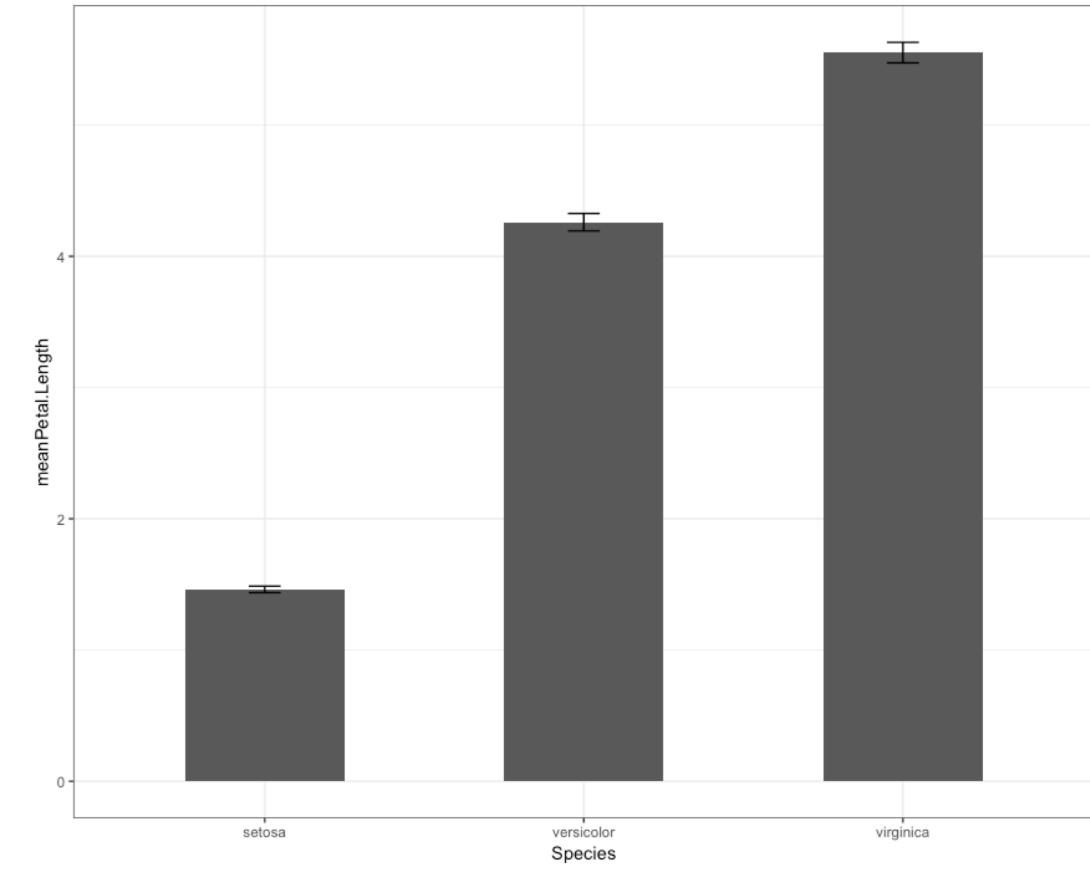
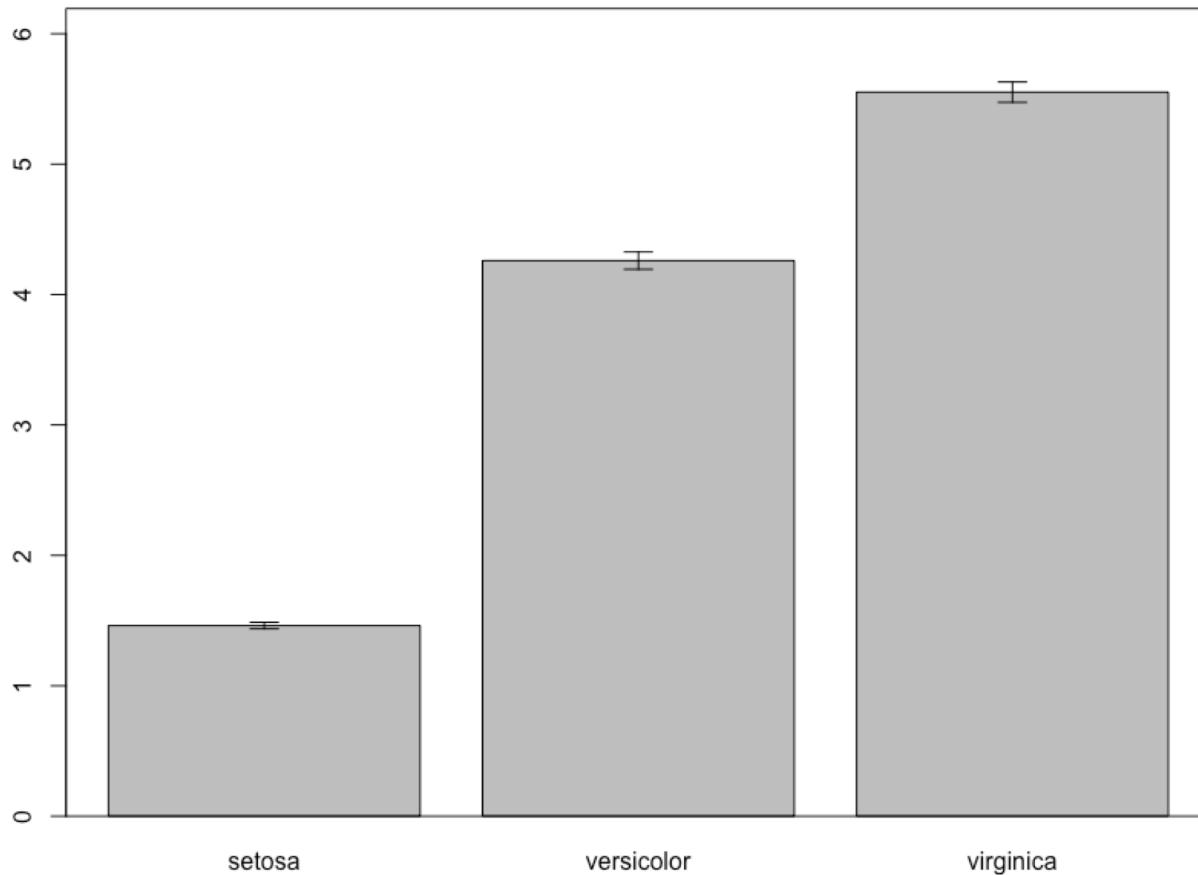
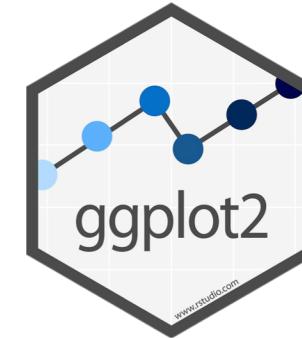
```
iris.barplot <- iris %>% group_by(Species) %>%  
summarise(meanPetal.Length = mean(Petal.Length),  
stderPetal.Length =  
sd(Petal.Length)/sqrt(length(Petal.Length)),  
lowerErr = (meanPetal.Length - stderPetal.Length),  
upperErr = (meanPetal.Length + stderPetal.Length))
```

```
ggplot(iris.barplot,  
aes(x = Species,  
y = meanPetal.Length)) + geom_col()  
+ geom_errorbar(  
aes(ymin = lowerErr,  
ymax = upperErr),  
width = 0)  
+ theme_bw()
```

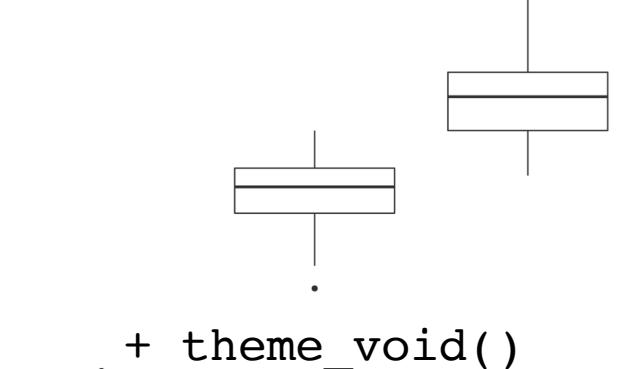
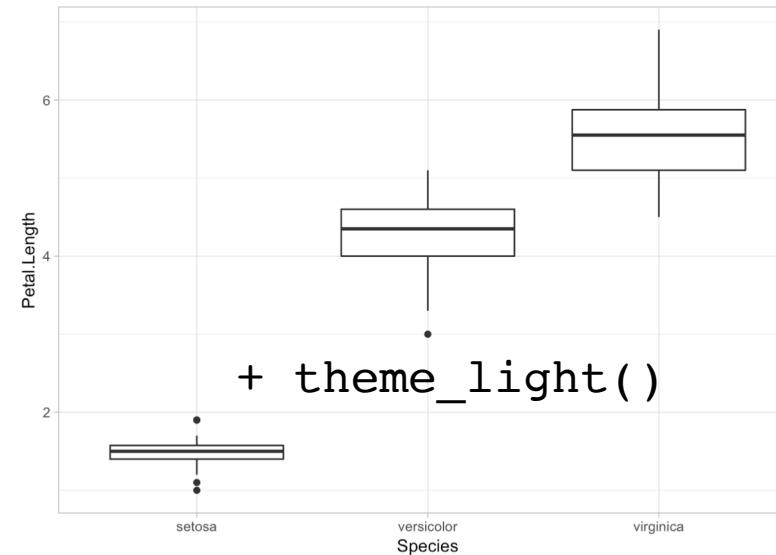
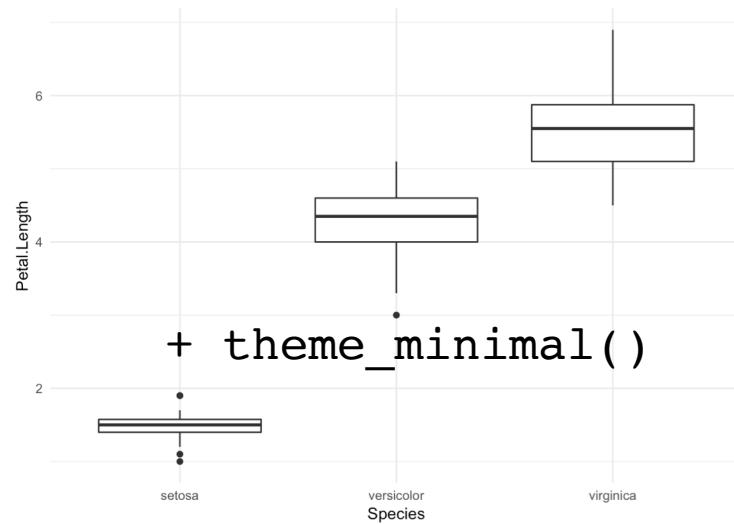
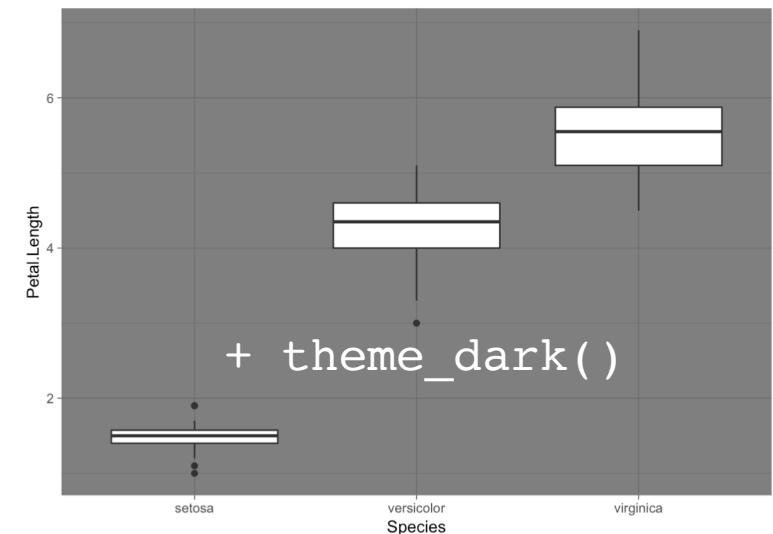
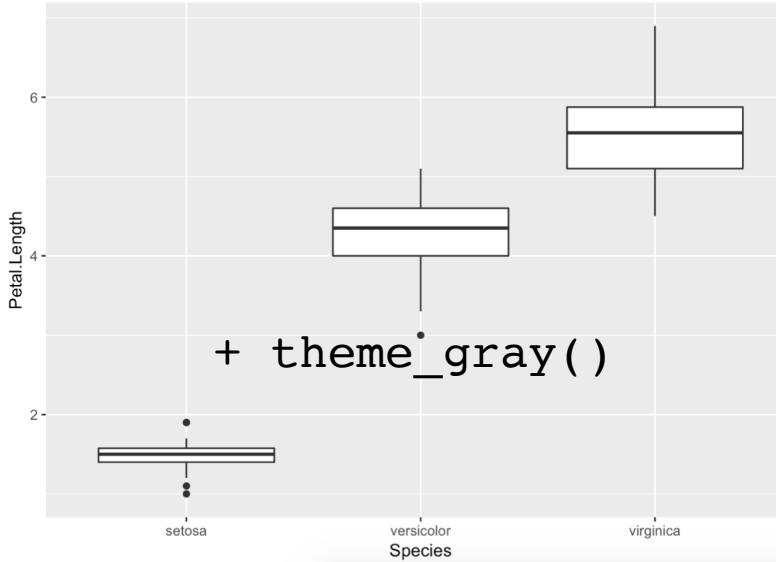
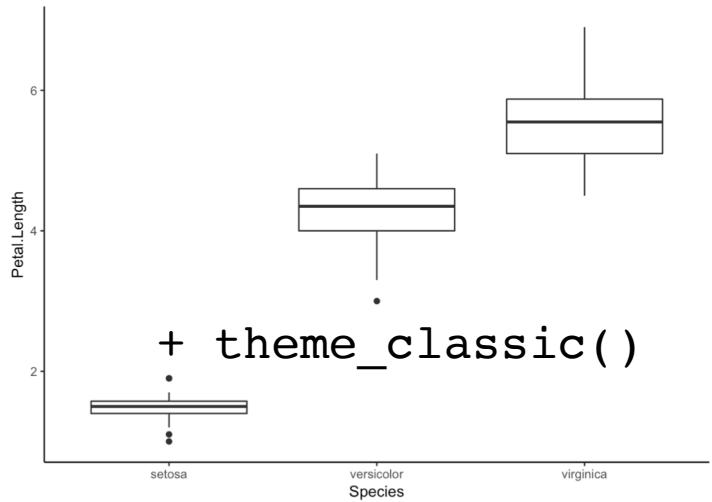
A quick comparison between



vs.

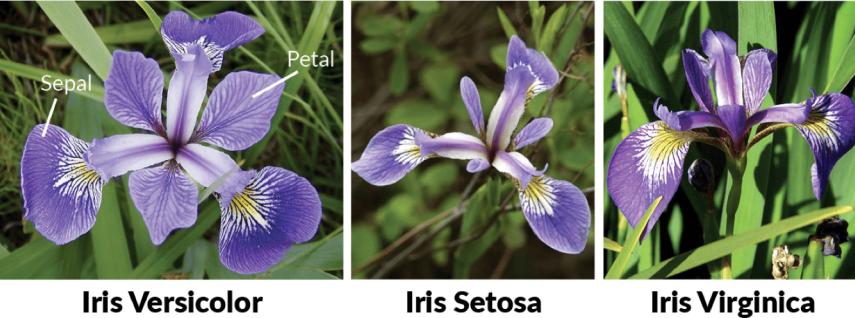


A quick look at themes





Customising your plot a bit more



Colour/shapes?

```
ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length, color = Species)) +  
  geom_point(size = 2.5, alpha = 0.6) +  
  theme_bw() +  
  theme(legend.position = "bottom",  
        legend.text = element_text(size = 12, family="Comic Sans MS"),  
        legend.title = element_text(size = 12, face = "bold", family="Comic Sans MS"),  
        panel.grid.major = element_blank(),  
        panel.grid.minor = element_blank(),  
        axis.text= element_text(size = 18),  
        axis.title= element_text(size = 20,face = "bold", family="Comic Sans MS"))
```

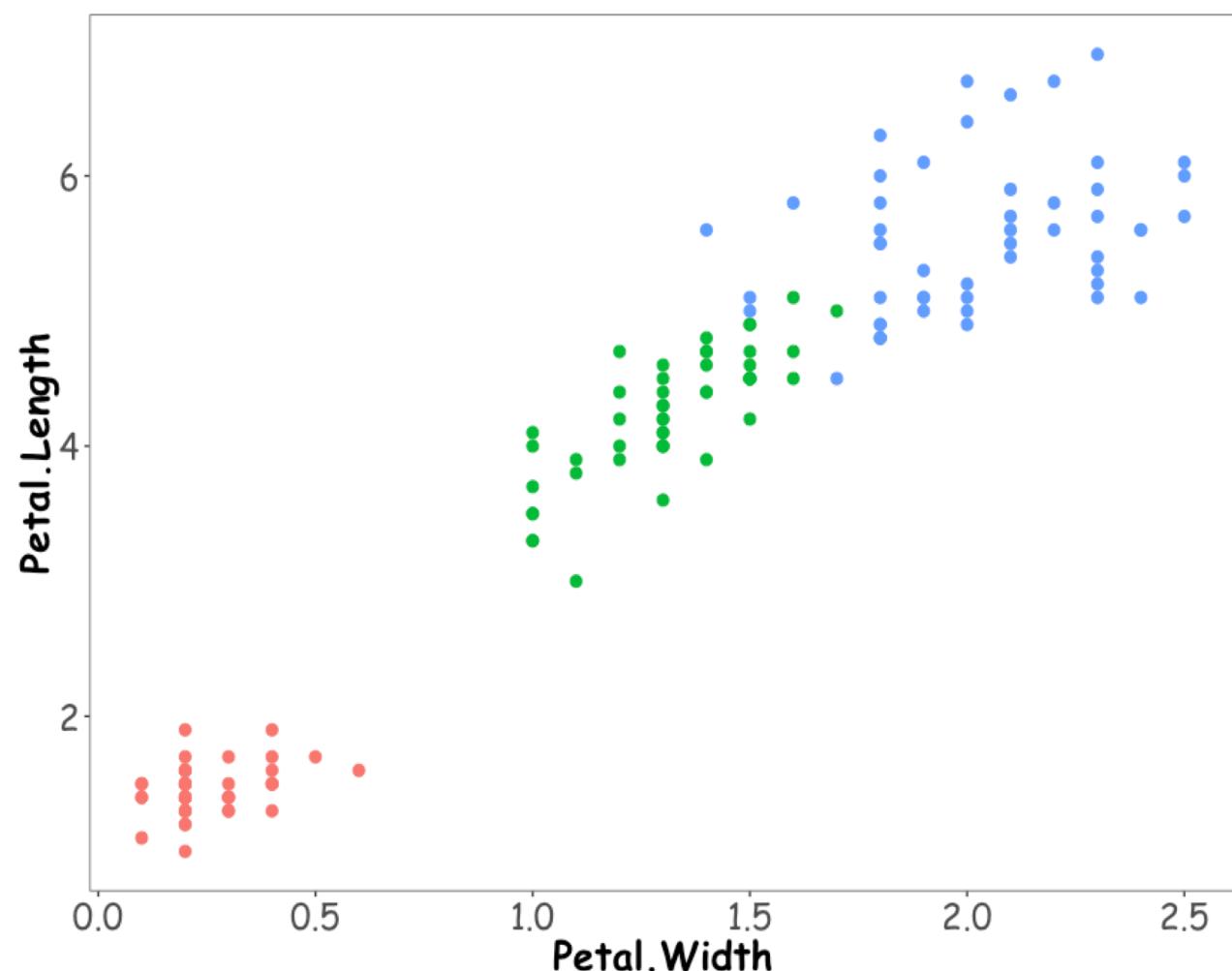
Fine-tuning your geom_

Fine-tuning your plot:
Legend?
Font type, size & style

Just because you can,
doesn't mean you should...

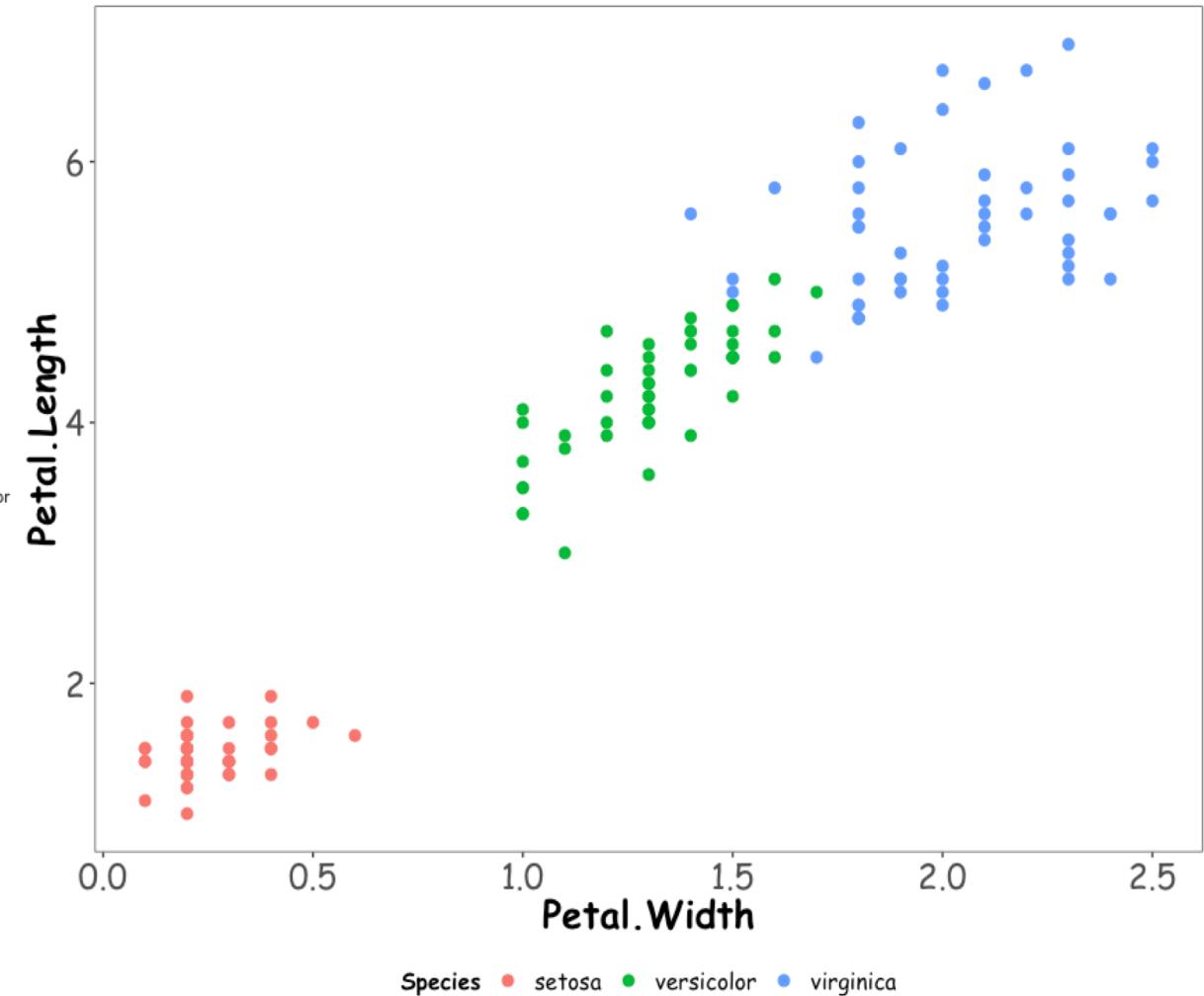
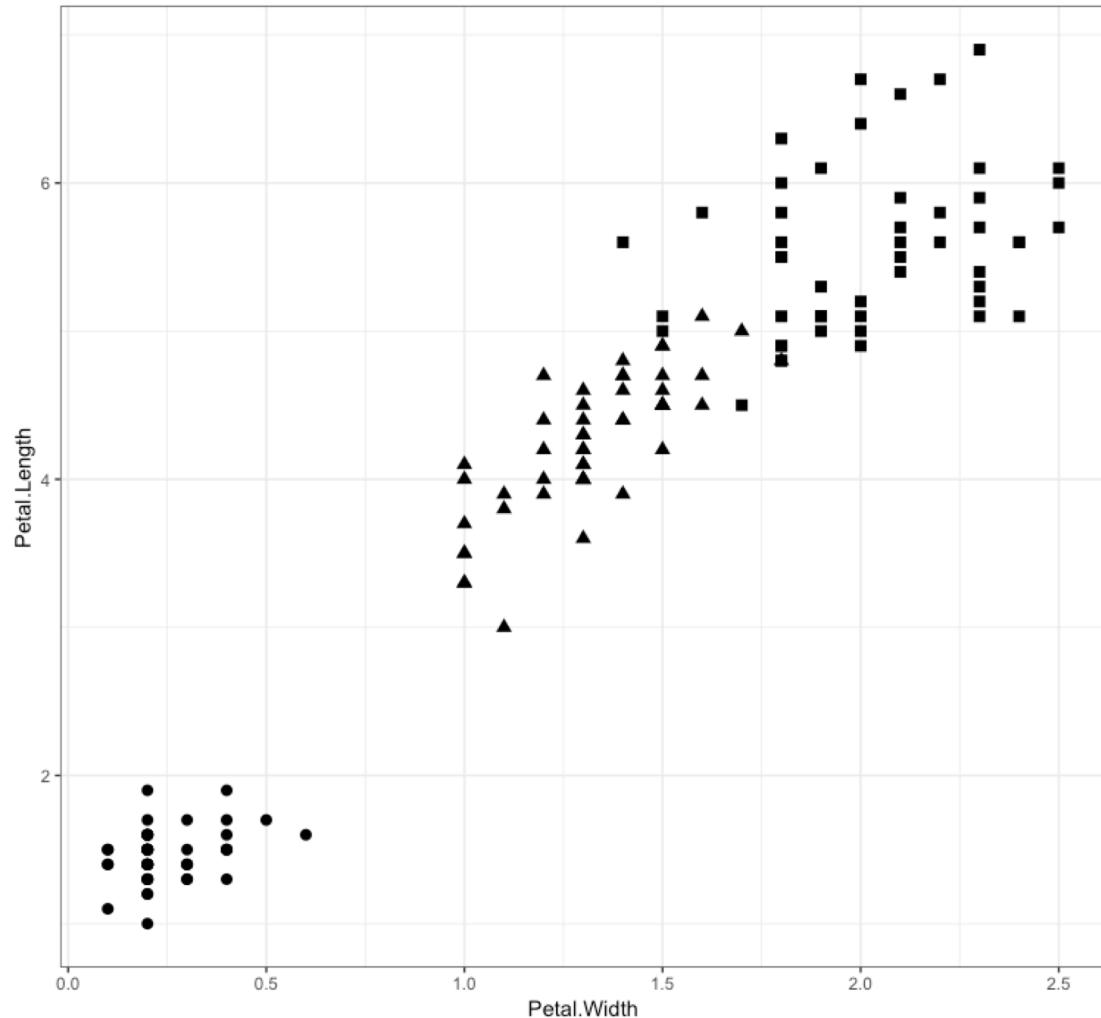
Consider:

- Target audience
- What is the best way to depict what I want?
- Could I simplify it further?
- How would I describe this in the figure legend in my paper?



```
ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length, color = Species)) +  
  geom_point(size = 2.5, alpha = 0.6) +  
  theme_bw() +  
  theme(legend.position = "bottom",  
        legend.text = element_text(size = 12, family="Comic Sans MS"),  
        legend.title = element_text(size = 12, face = "bold", family="Comic Sans MS"),  
        panel.grid.major = element_blank(),  
        panel.grid.minor = element_blank(),  
        axis.text= element_text(size = 18),  
        axis.title= element_text(size = 20,face = "bold", family="Comic Sans MS"))
```

Less is *more*



Creating your theme for consistency

This

```
my_thesis_theme <- theme(legend.position = "bottom",
                           legend.text = element_text(size = 12, family="Comic Sans MS"),
                           legend.title = element_text(size = 12, face = "bold", family="Comic Sans MS"),
                           panel.grid.major = element_blank(),
                           panel.grid.minor = element_blank(),
                           axis.text= element_text(size = 18),
                           axis.title= element_text(size = 20,face = "bold", family="Comic Sans MS"))

ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length, color = Species)) +
  geom_point(size = 2.5, alpha = 0.6) +
  theme_bw() +
  my_thesis_theme
```

Add this for all your plots from now on!

Instead of this

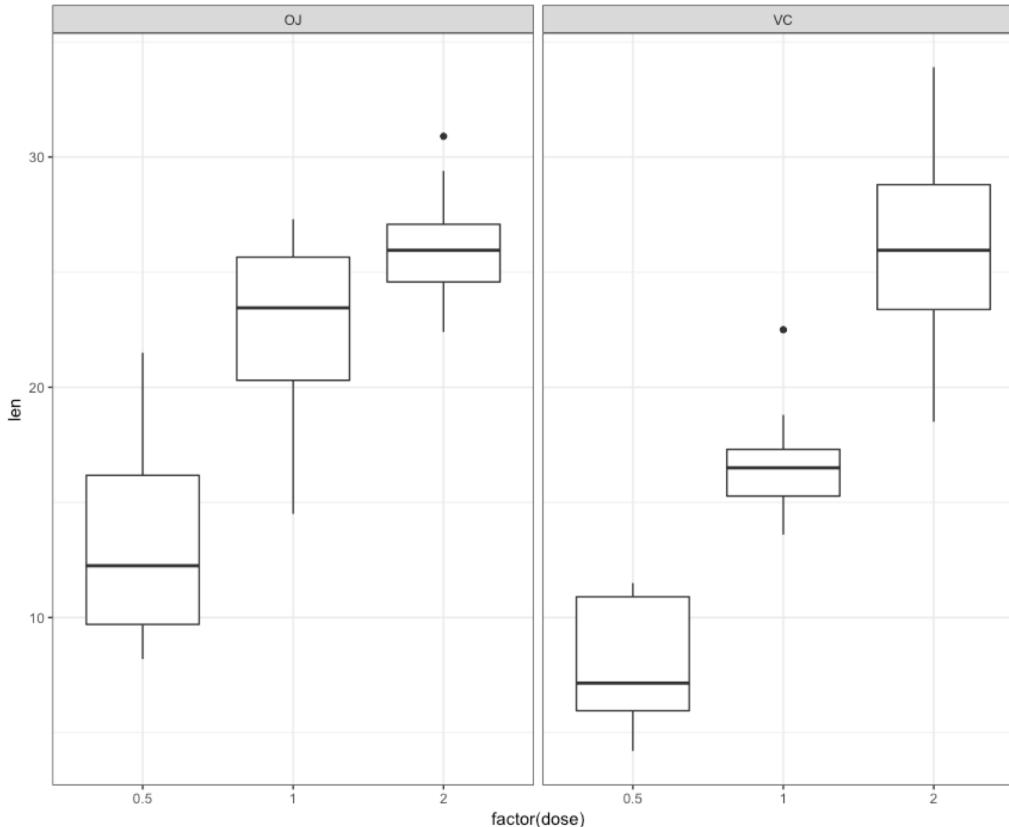
```
ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length, color = Species)) +
  geom_point(size = 2.5, alpha = 0.6) +
  theme_bw() +
  theme(legend.position = "bottom",
        legend.text = element_text(size = 12, family="Comic Sans MS"),
        legend.title = element_text(size = 12, face = "bold", family="Comic Sans MS"),
        panel.grid.major = element_blank(),
        panel.grid.minor = element_blank(),
        axis.text= element_text(size = 18),
        axis.title= element_text(size = 20,face = "bold", family="Comic Sans MS"))
```



Facets! New favourite thing

Great for multivariate data (i.e. a factor and a continuous variable)

```
Toothplot <- ggplot(data = ToothGrowth, aes(x = factor(dose), y = len)) +  
  geom_boxplot() +  
  theme_bw()
```



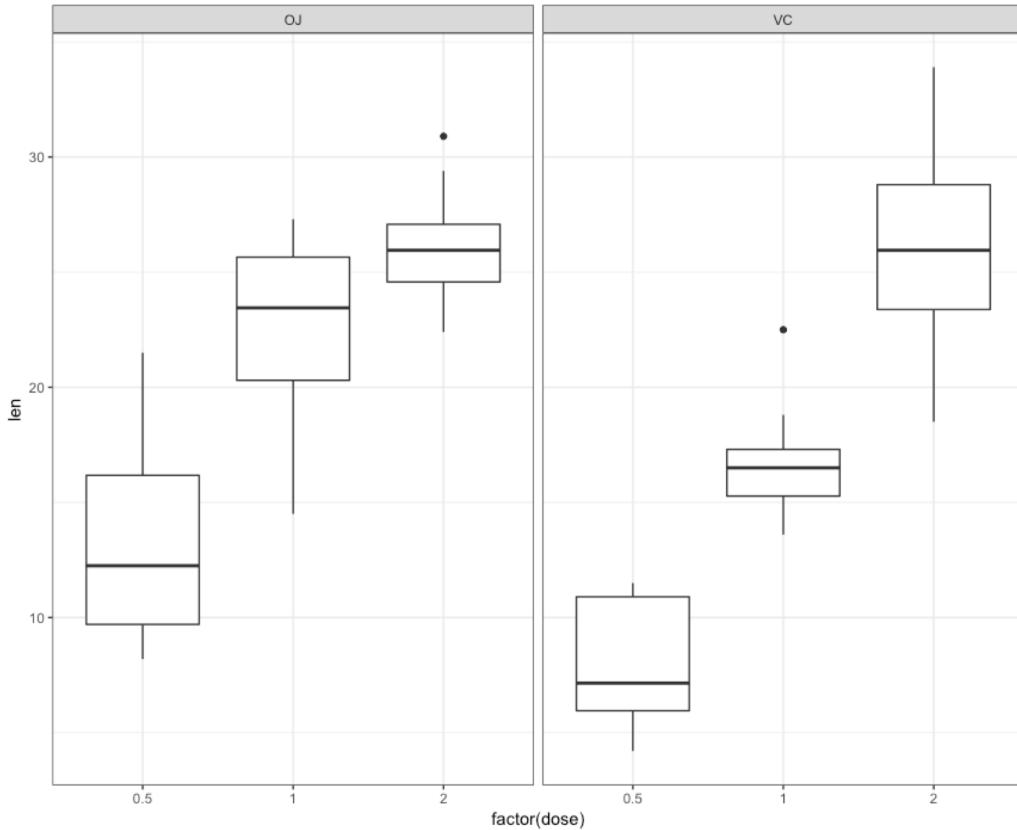
```
Toothplot + facet_wrap(~ supp)
```



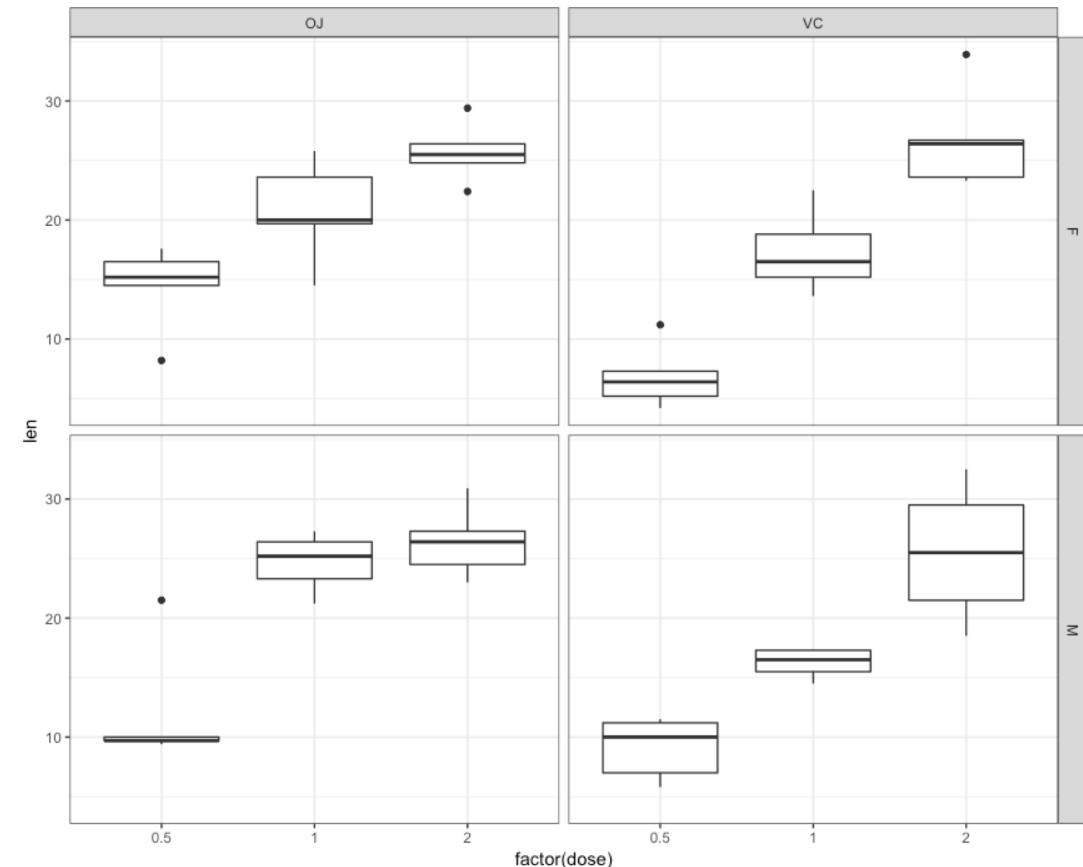
Facets! New favourite thing

Great for multivariate data (i.e. a factor and a continuous variable)

```
Toothplot <- ggplot(data = ToothGrowth, aes(x = factor(dose), y = len)) +  
  geom_boxplot() +  
  theme_bw()
```



```
Toothplot + facet_wrap(~ supp)
```

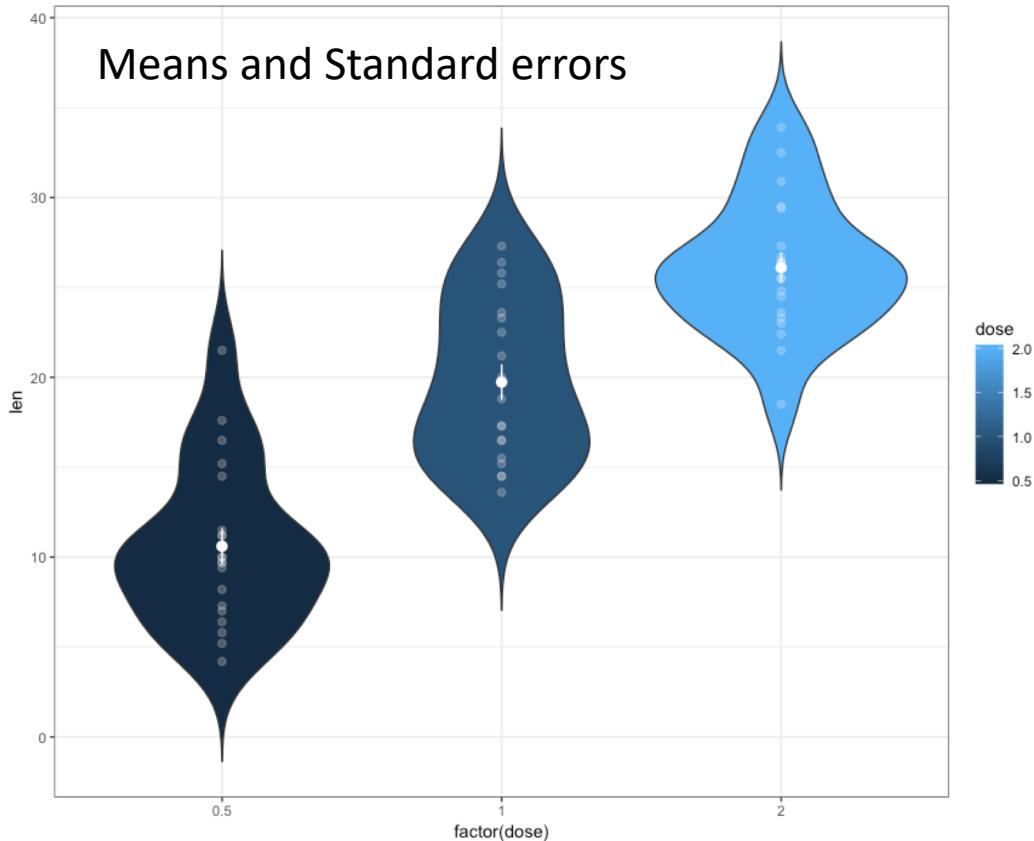


```
*Toothplot + facet_grid(Sex ~ supp)
```

* Run this first `ToothGrowth$Sex <- rep(c("F", "M"), 2, 30)`

Built in functions

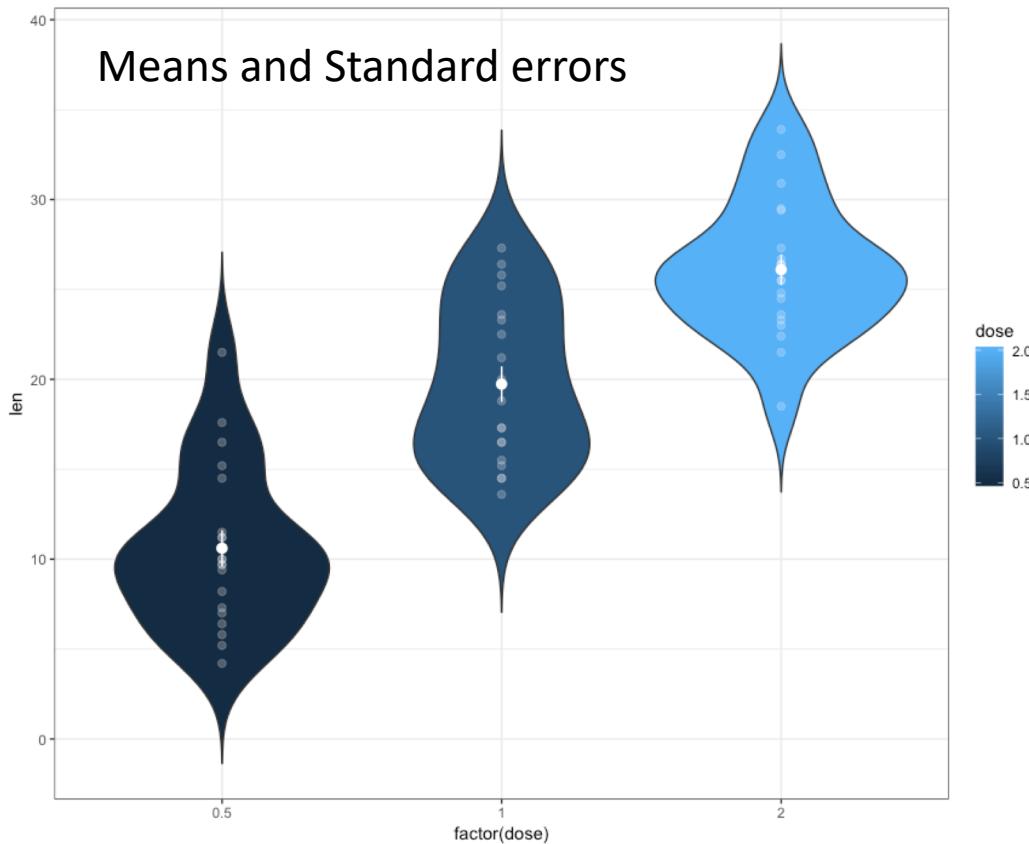
Some quick ways to explore data



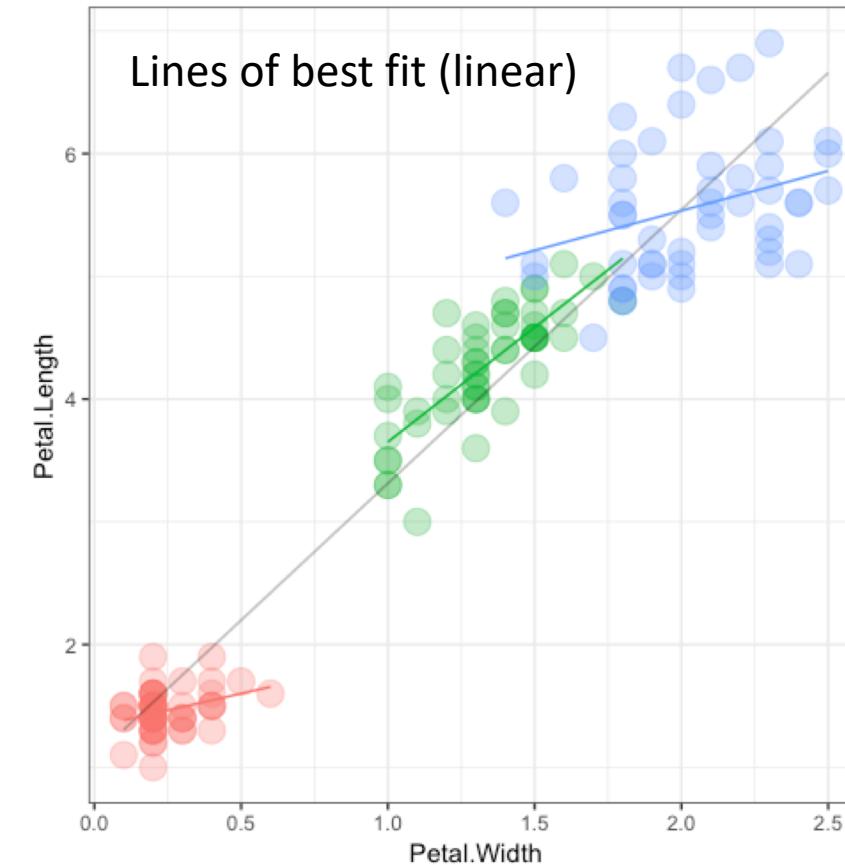
```
ggplot(data = ToothGrowth,  
aes(x = factor(dose), y = len, fill = dose)) +  
  geom_violin(trim = F) +  
  geom_point(col = "white", size = 2, alpha = 0.2) +  
  stat_summary(fun.data = "mean_se", col = "white") +  
  theme_bw()
```

Built in functions

Some quick ways to explore data



```
ggplot(data = ToothGrowth,  
aes(x = factor(dose), y = len, fill = dose)) +  
  geom_violin(trim = F) +  
  geom_point(col = "white", size = 2, alpha = 0.2) +  
  stat_summary(fun.data = "mean_se", col = "white") +  
  theme_bw()
```

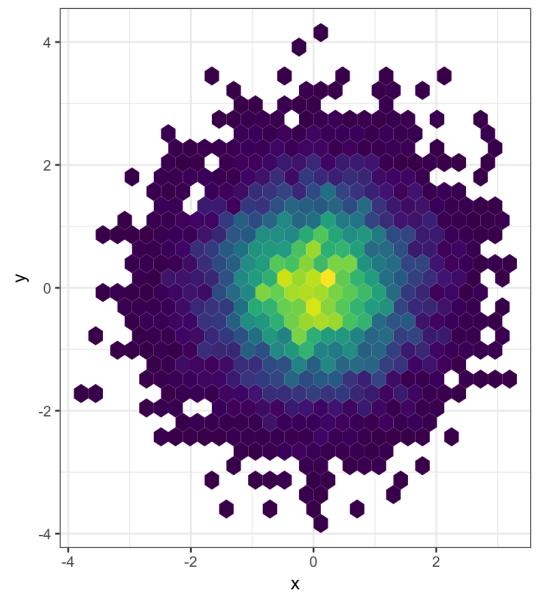


```
ggplot(data = iris, aes(x = Petal.Width, y = Petal.Length)) +  
  geom_point(aes(color = Species), size = 5, alpha = 0.3) +  
  geom_line(stat = "smooth", method = "lm", alpha = 0.3) +  
  geom_line(aes(group = Species, colour = Species), stat = "smooth",  
            method = "lm", lwd = 0.5) +  
  theme_bw()
```

Working with palettes

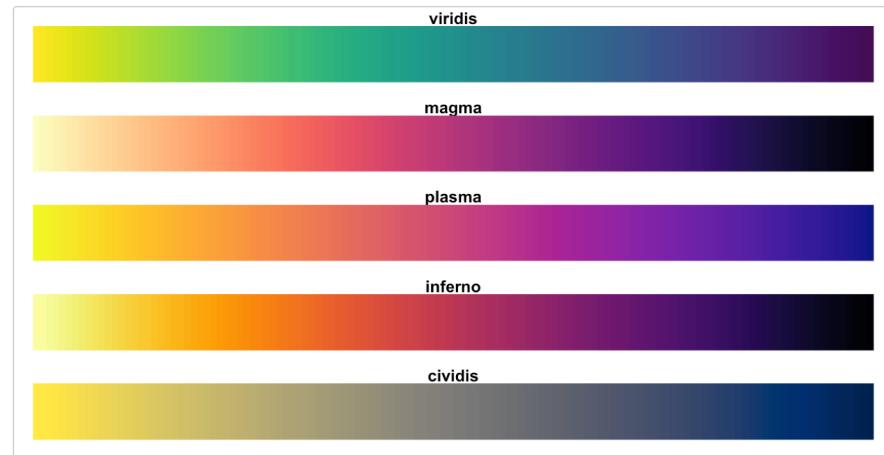
Some quick ways to make your colours look cohesive

```
install.packages("viridis")  
library(viridis)
```

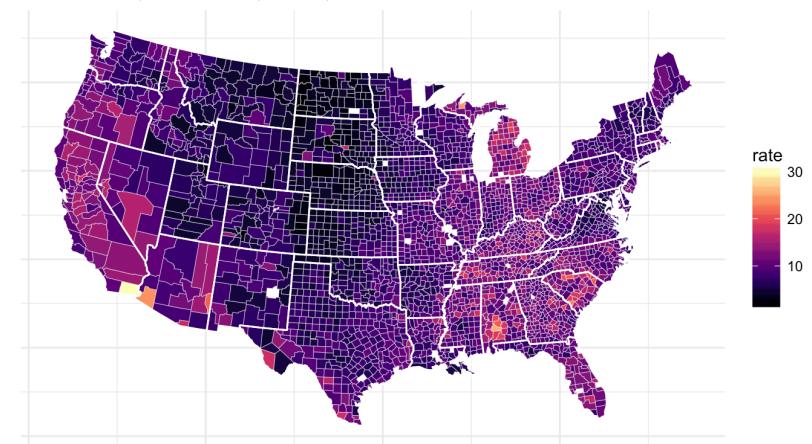


The Color Scales

The package contains four color scales: "Viridis", the primary choice, and three alternatives with similar properties, "magma", "plasma", and "inferno."

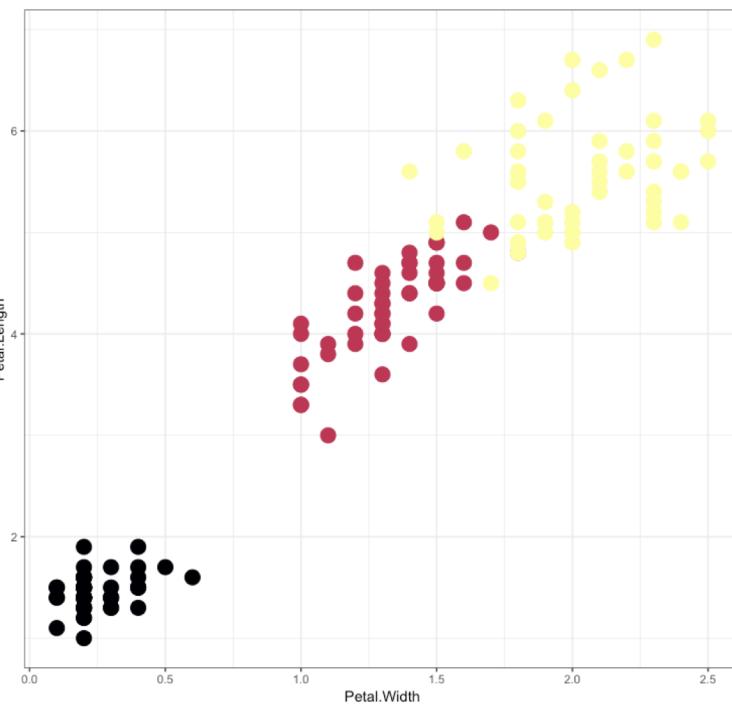
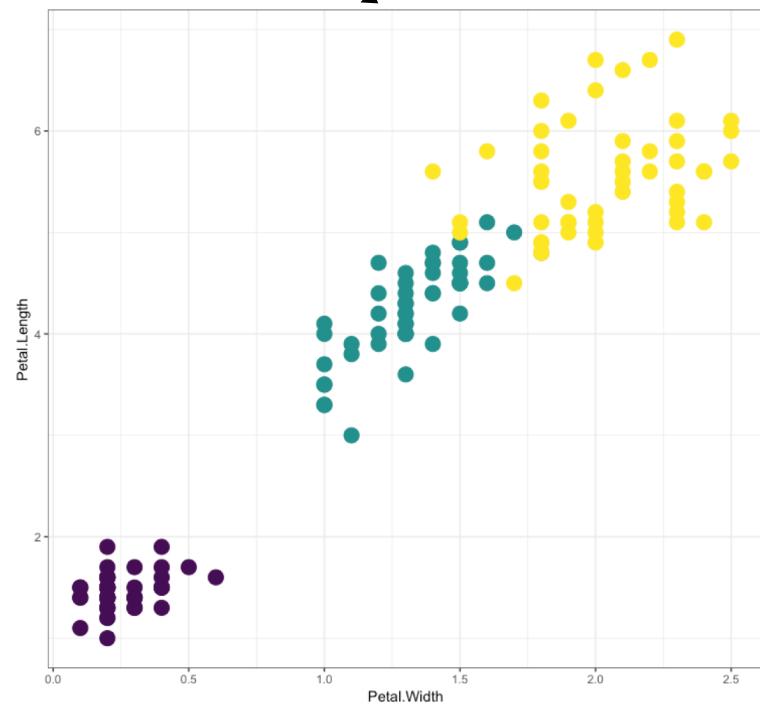


US unemployment rate by county

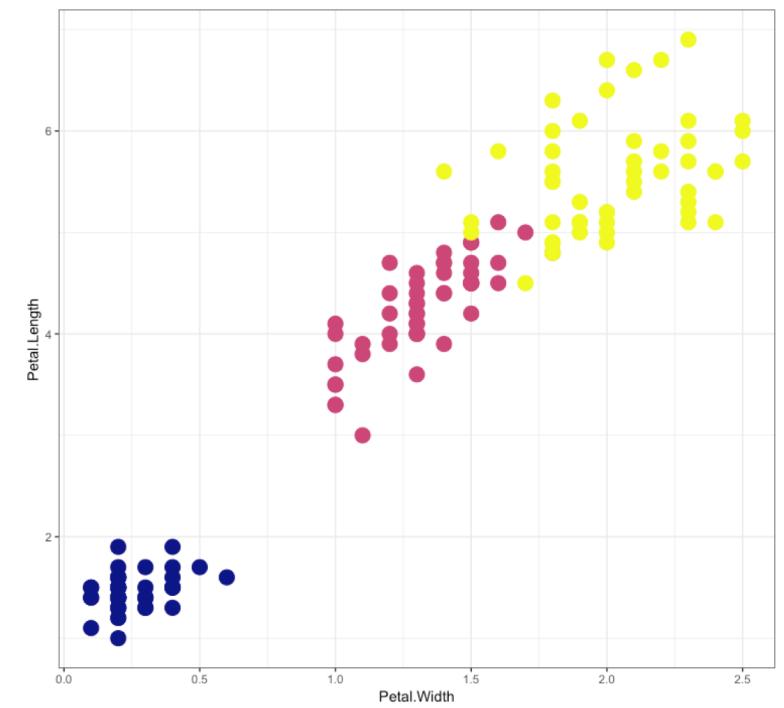


```
irisplot <- ggplot(data = iris,  
                    aes(x = Petal.Width, y = Petal.Length, color = Species)) +  
  geom_point(size = 5) +  
  theme_bw()
```

```
irisplot + scale_color_viridis(discrete = T)
```



```
scale_color_viridis(discrete = T,  
                     option = "B")
```



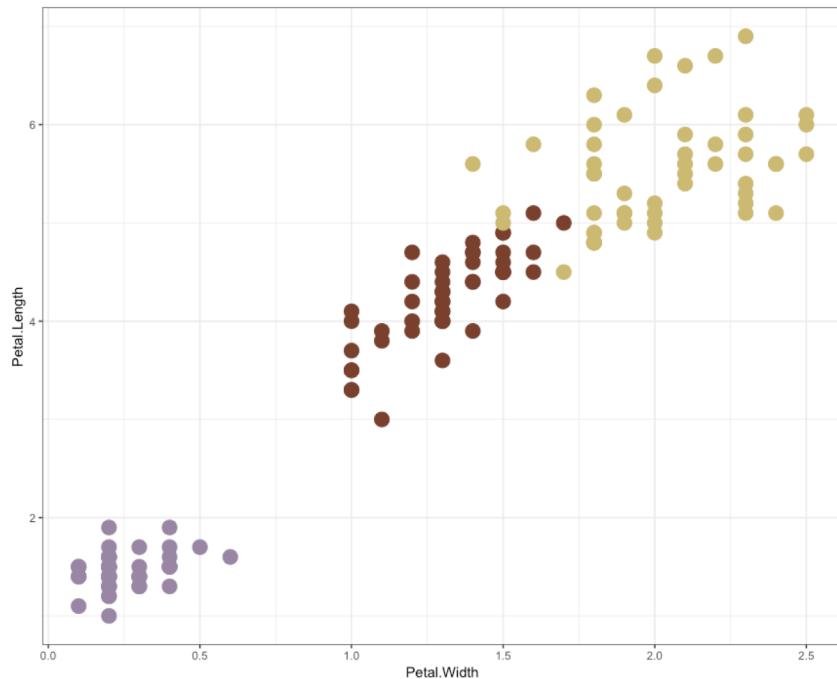
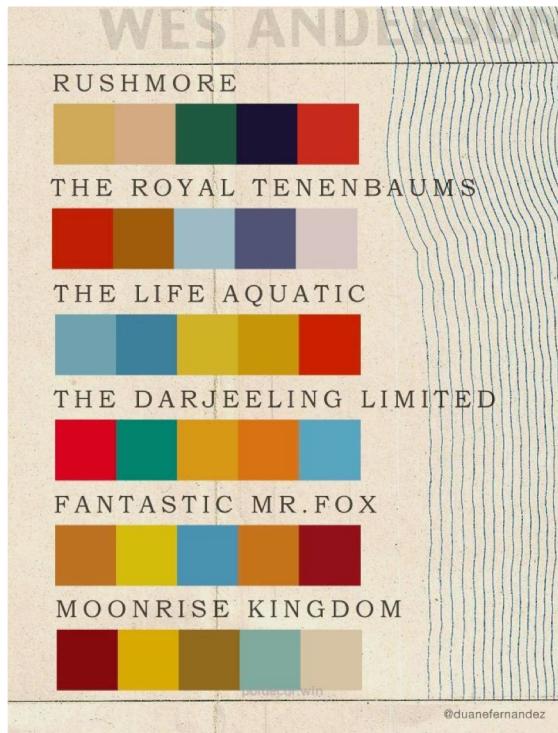
```
scale_color_viridis(discrete = T,  
                     option = "C")
```

Working with palettes

Some quick ways to make your colours look cohesive

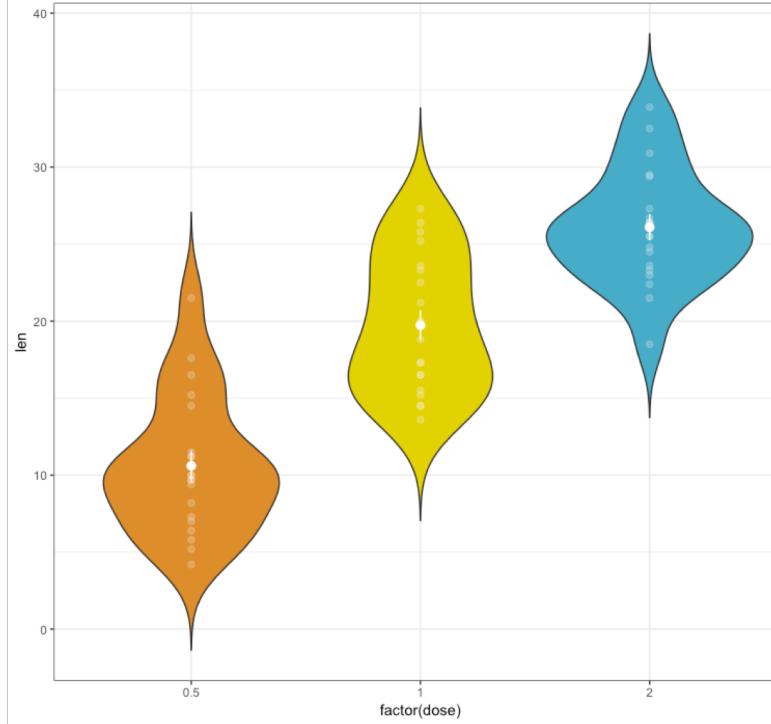
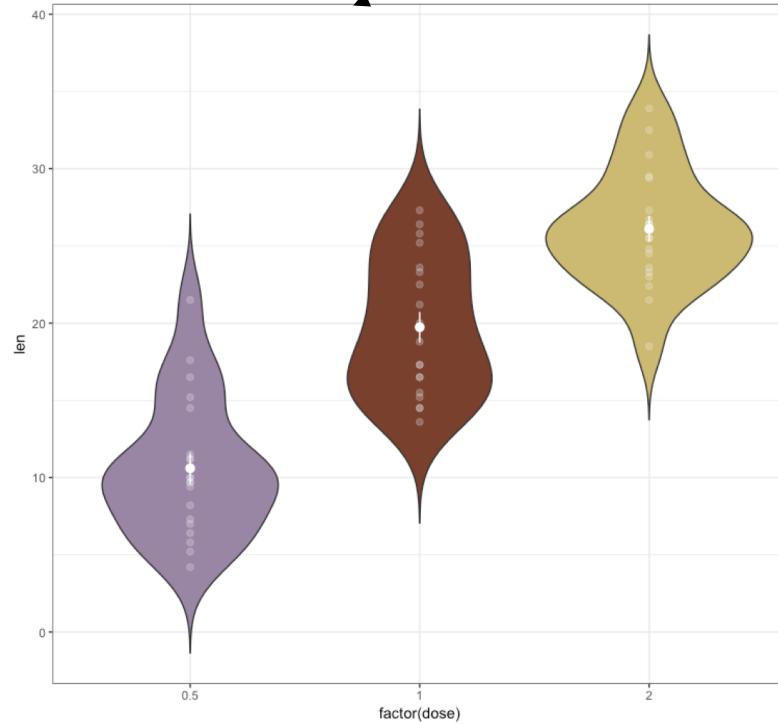


```
devtools::install_github("karthik/wesanderson")  
library(wesanderson)
```

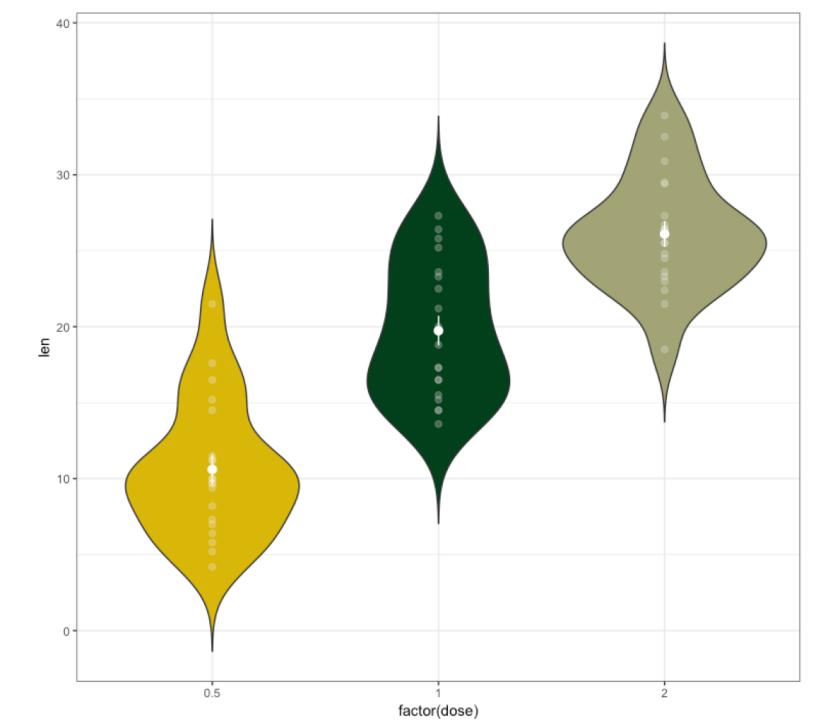


```
toothplot <- ggplot(data = ToothGrowth, aes(x = factor(dose), y = len, fill = factor(dose))) +  
  geom_violin(trim = F) +  
  geom_point(col = "white", size = 2, alpha = 0.2) +  
  stat_summary(fun.data = "mean_se", col = "white") +  
  theme_bw()
```

```
toothplot +  
scale_fill_manual(values = wes_palette("IsleofDogs1"))
```



wes_palette("FantasticFox1")



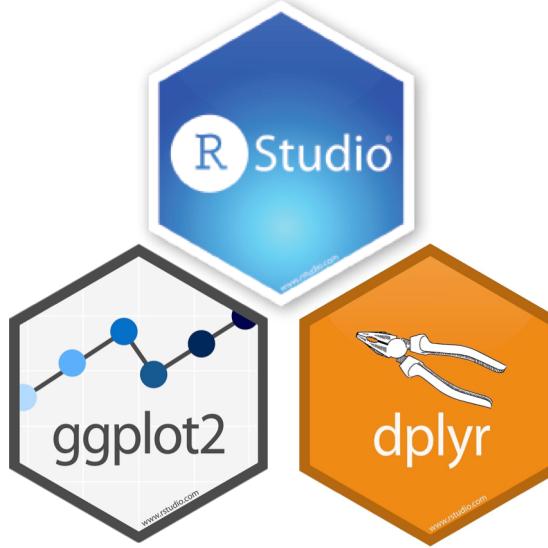
wes_palette("Cavalcanti1")

Final words...

- This is only the beginning
- LOTS of google-able help!
- Always think about what the **key message** you are trying to graphically represent
- Less is **more**



Environmental Computing



The screenshot displays the R Studio Data Visualization with ggplot2 Cheat Sheet. It is a comprehensive guide to the ggplot2 package, organized into several sections:

- Geoms**: Shows how to use a geom function to represent data points, using the geom's aesthetic properties to represent variables. Each function returns a layer.
- Graphical Primitives**: Lists basic ggplot2 components: aes, aes_string, aes_(), and geom_().
- Basics**: Provides examples for creating plots with a single data source and mapping variables to data.
- Continuous X, Continuous Y**: Examples for creating plots where both axes are continuous.
- Continuous Function**: Examples for creating plots where the x-axis is a continuous function.
- Two Variables**: Examples for creating plots with two variables.
- Visualizing error**: Examples for handling errors in data frames.
- Continuous X, Continuous Y**: Examples for creating plots with a single data source and mapping variables to data.
- Continuous Function**: Examples for creating plots where the x-axis is a continuous function.
- Two Variables**: Examples for creating plots with two variables.
- Geoms**: A detailed section showing how to use a geom function to represent data points, using the geom's aesthetic properties to represent variables. Each function returns a layer.
- Continuous X, Continuous Y**: Examples for creating plots with a single data source and mapping variables to data.
- Continuous Function**: Examples for creating plots where the x-axis is a continuous function.
- Two Variables**: Examples for creating plots with two variables.



<http://environmentalcomputing.net/plotting-with-ggplot/>