

# Enter the Tidyverse

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BIO5312 FALL2017

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# What is the “tidyverse”?

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A collection of R packages largely developed by Hadley Wickham and others at Rstudio

Have emerged as staples of modern-day data science in the past 5—10 years

We will focus on:

- Visualization/plotting with `ggplot2`
- Data management and “wrangling” with `dplyr` and `tidyr`
- Document presentation with `RMarkdown`



# Focus is on tidy dataframes

Each variable forms a column.

Each observation forms a row.

Each type of observational unit forms a table.

Tidy data provides a consistent approach to data management that greatly facilitates downstream analysis and viz

country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	2666	20595360
Brazil	1999	31737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272915272
China	2000	213766	128042583

variables

country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	2666	20595360
Brazil	1999	31737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272915272
China	2000	213766	128042583

observations

country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	2666	20595360
Brazil	1999	31737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272915272
China	2000	213766	128042583

values

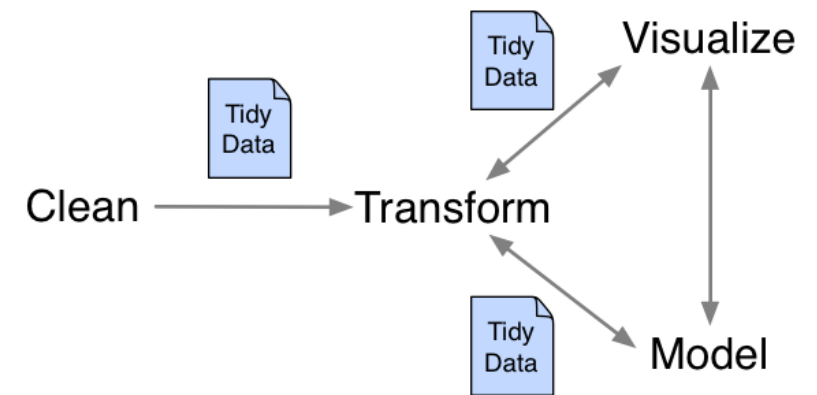
# Working with tidy data

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The package `dplyr` can manipulate and manage tidy data

The package `tidyr` can rearrange data to convert to/from tidy data

The package `ggplot2` is used for visualization/plotting



# The fundamental verbs of dplyr

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<code>filter()</code>	select rows
<code>select()</code>	select columns
<code>mutate()</code>	create new columns
<code>group_by()</code>	establish a data grouping
<code>tally()</code>	count observations in a grouping
<code>summarize()</code>	calculate summary statistic
<code>arrange()</code>	arrange rows

There are more functions but these ones are key!

# The pipe operator %>%

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“Pipes” output from one function/operation as input to the next

```
## Start simple: display data  
head(iris)
```

```
## Using %>%  
iris %>% head()
```

```
## Find the mean of iris sepal lengths  
mean.sepal <- mean(iris$Sepal.Length)
```

```
## Using %>%  
mean.sepal <- iris$Sepal.Length %>% mean()
```

```
iris$Sepal.Length %>% mean() -> mean.sepal
```

```
iris %>% mean(Sepal.Length) -> mean.sepal
```

“forward assignment” operator follows the logical flow of piping

# dp1yr demo

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Commands in demo are on  
[sjspielman.org/bio5312\\_fall2017/day2\\_tidyverse1](https://sjspielman.org/bio5312_fall2017/day2_tidyverse1)

# Visualizing with ggplot2

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The package `ggplot2` is a graphics package that implements a **grammar of graphics**

- Operates on *data frames*, not vectors like Base R
- Explicitly differentiates between the data and the representation of the data



# The ggplot2 grammar

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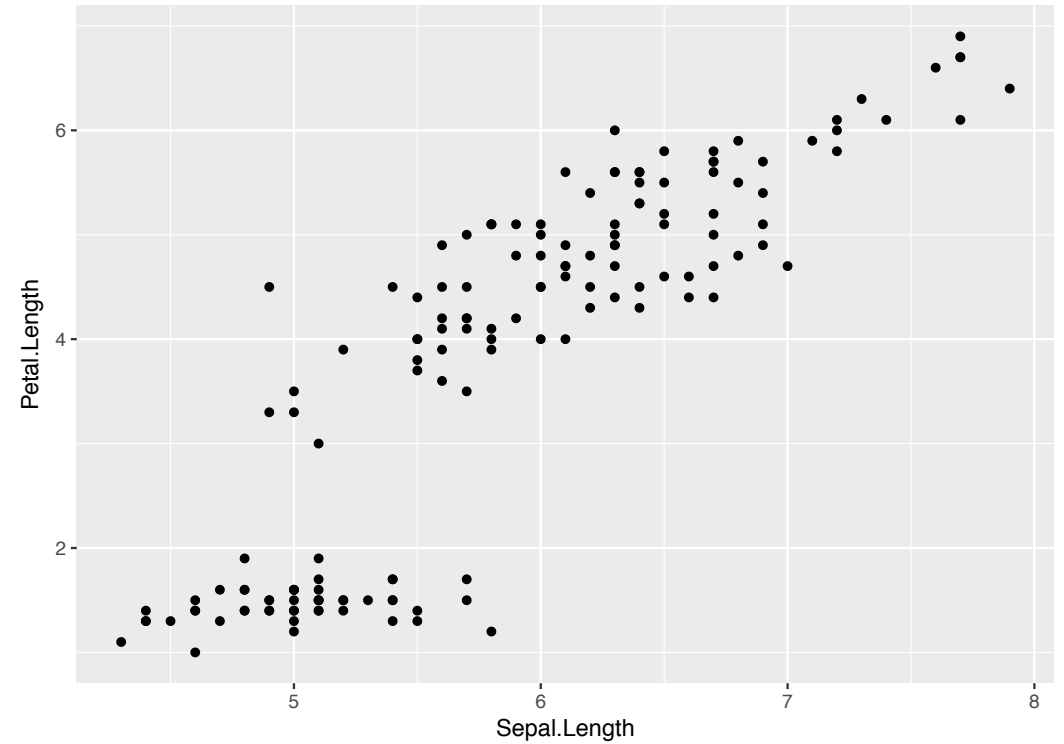
Grammar element*	What is it
<b>Data</b>	The data frame being plotted
<b>Geometrics</b>	The geometric shape that will represent the data <ul style="list-style-type: none"><li>• Point, boxplot, histogram, violin, bar, etc.</li></ul>
<b>Aesthetics</b>	The aesthetics of the geometric object <ul style="list-style-type: none"><li>• Color, size, shape, etc.</li></ul>

\*Table is tiny subset of what ggplot2 has to offer

# Example: scatterplot

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```
> ggplot(iris, aes(x = Sepal.Length, y = Petal.Length)) + geom_point()
```

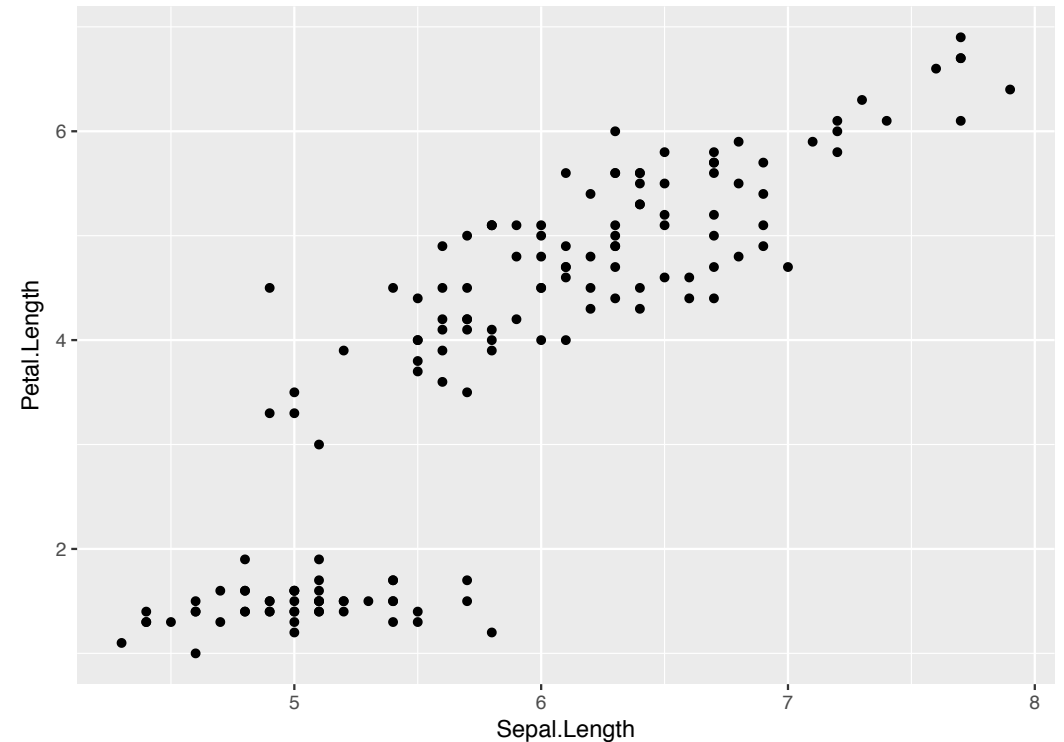


# Example: scatterplot

---

```
> ggplot( iris, aes(x = Sepal.Length, y = Petal.Length) ) + geom_point()
```

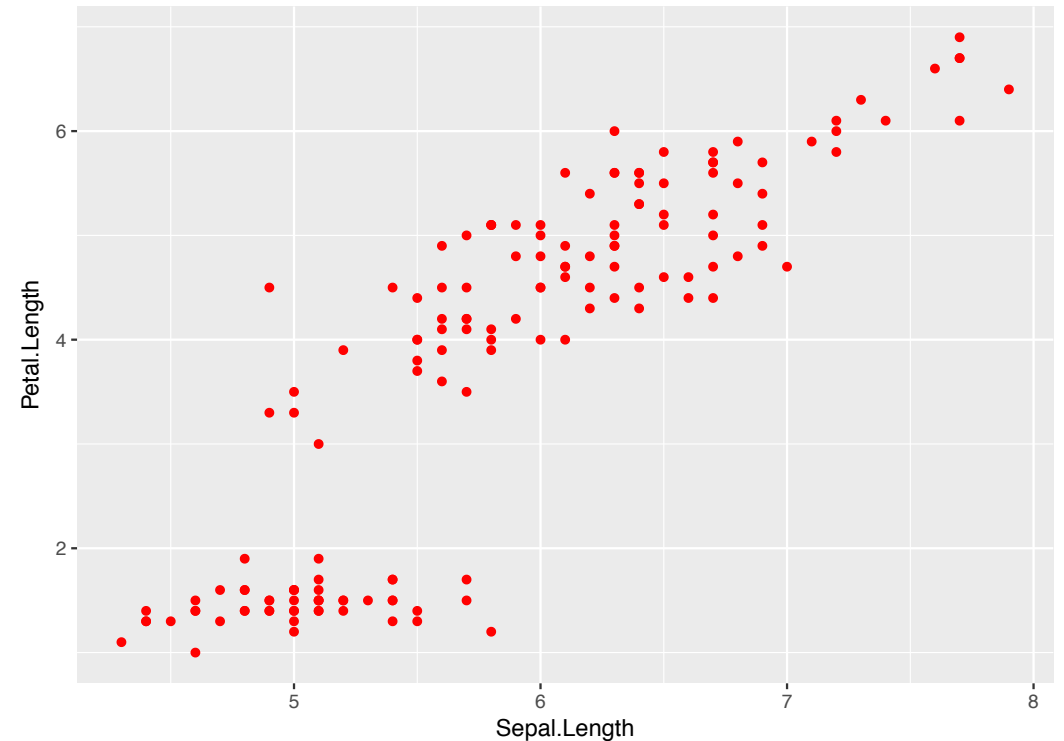
- Pass in the data frame as your first argument
- Aesthetics map the data onto plot characteristics, here x and y axes
- Display the data geometrically as points



# Example: scatterplot with color

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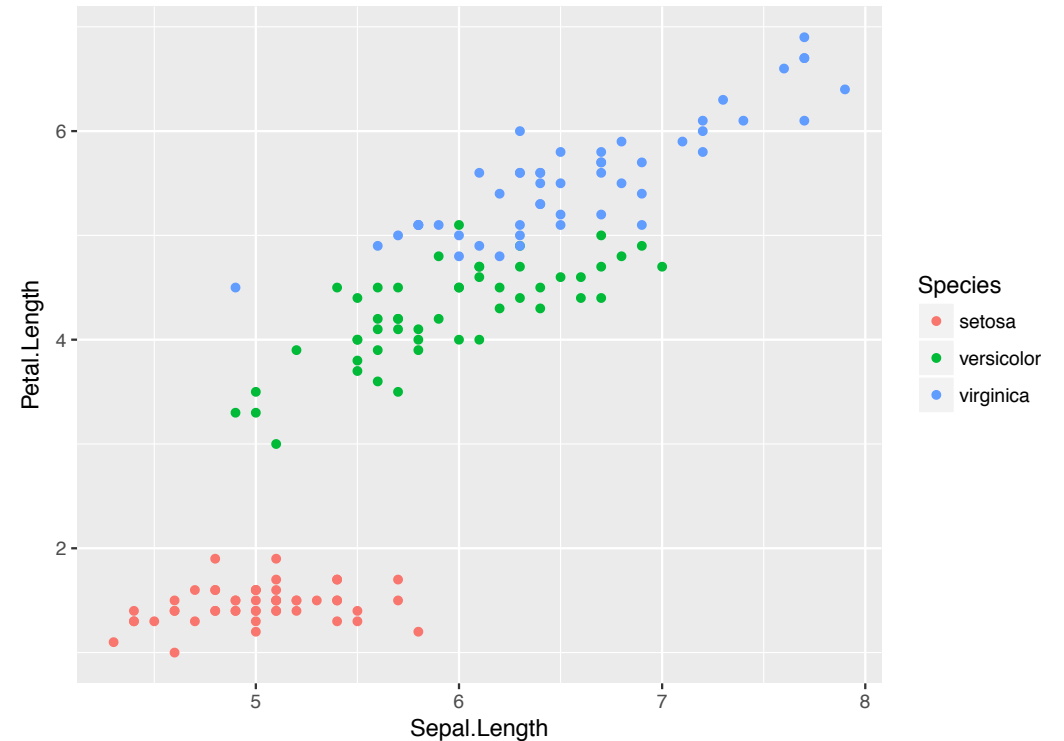
```
> ggplot(iris, aes(x = Sepal.Length, y = Petal.Length)) + geom_point(color = "red" )
```



# Example: scatterplot with aes color

```
> ggplot(iris, aes(x = Sepal.Length, y = Petal.Length, color = Species)) + geom_point()
```

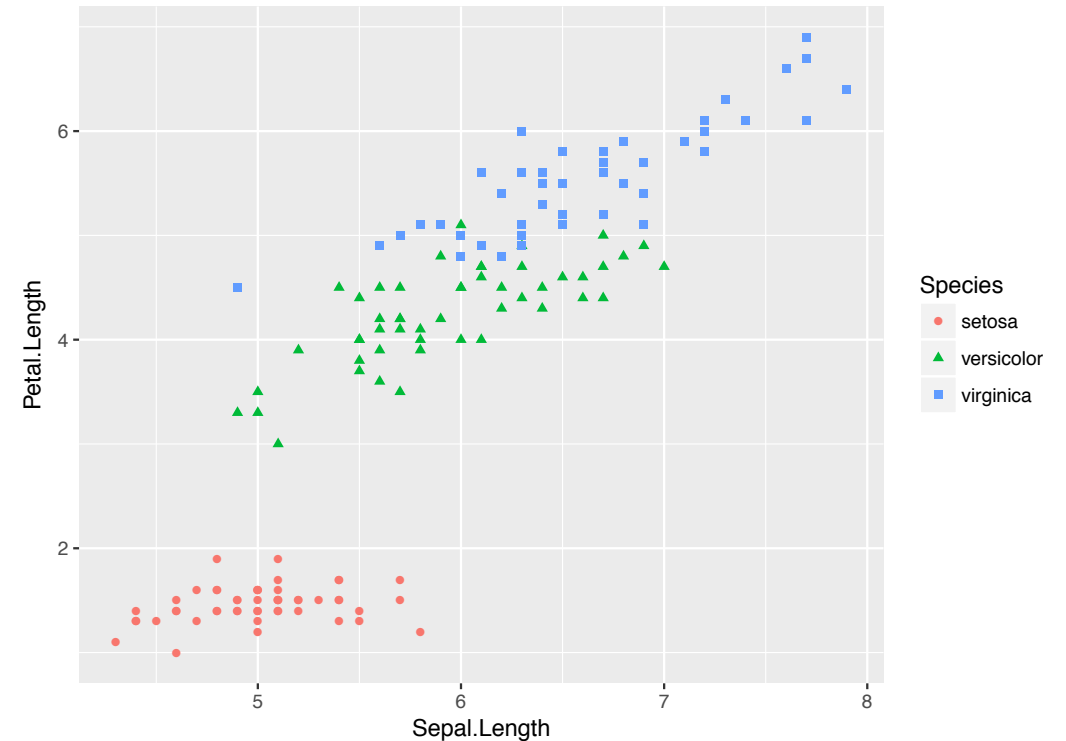
- Placing color inside aesesthetic maps it to the data.



# Example: scatterplot with `aes` color, shape

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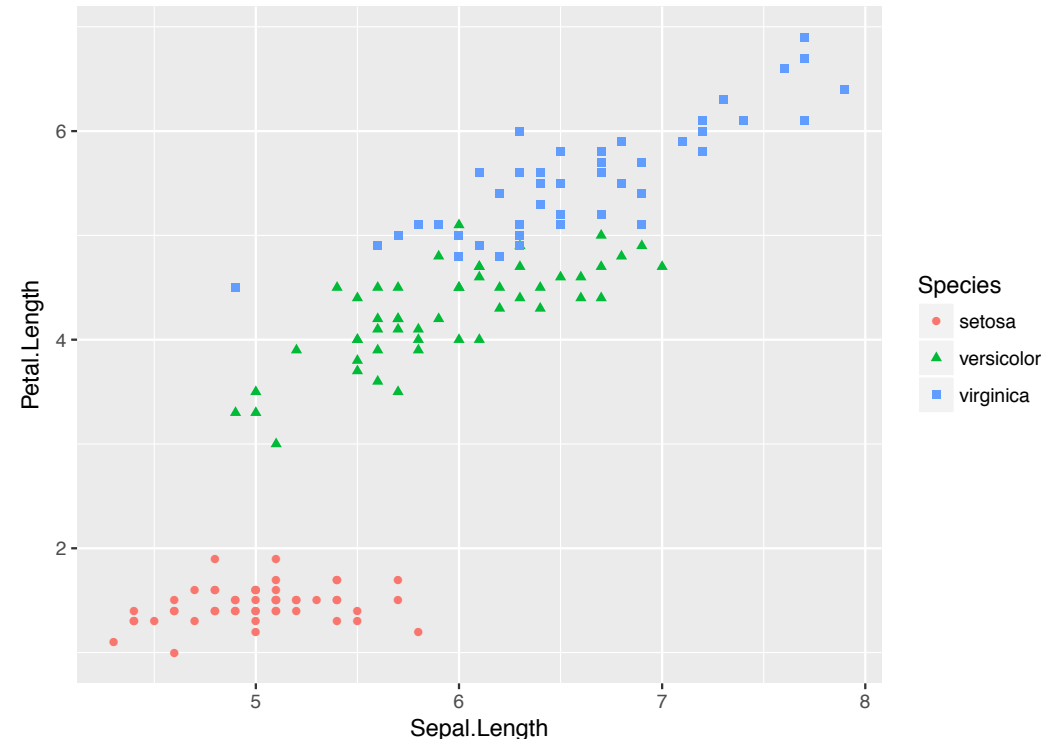
```
> ggplot(iris, aes(x = Sepal.Length, y = Petal.Length, color = Species, shape = Species))  
+ geom_point()
```



# Aesthetics may be placed inside the relevant geom

```
> ggplot(iris, aes(x = Sepal.Length, y = Petal.Length)) + geom_point(aes(color = Species, shape = Species))
```

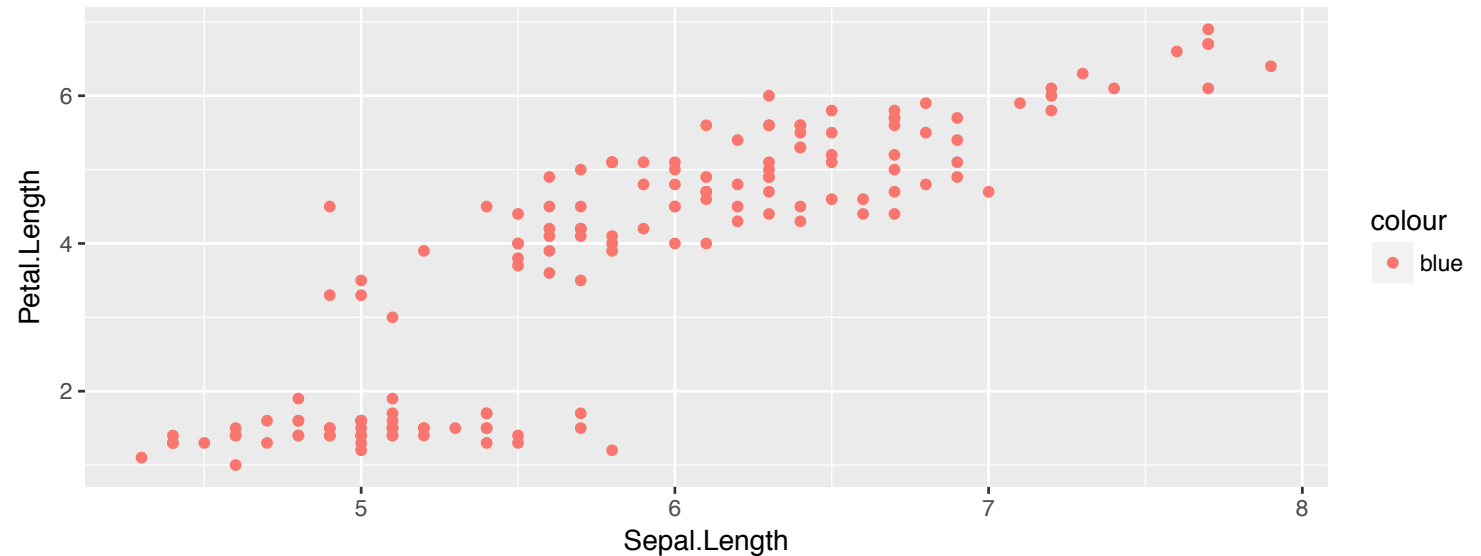
```
> ## Remember dplyr!  
> iris %>% ggplot(aes(x = Sepal.Length, y = Petal.Length)) + geom_point(aes(color = Species, shape = Species))
```



# Aesthetics are for mapping only

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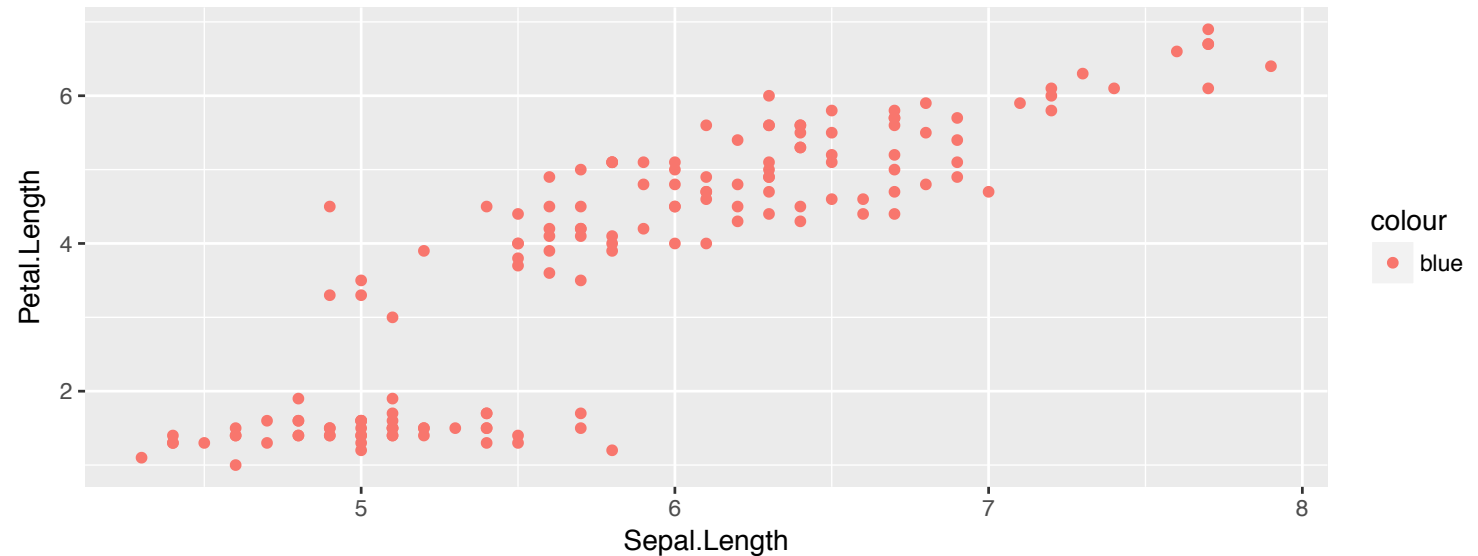
```
> ### Color all points blue?  
> ggplot(iris, aes(x = Sepal.Length, y = Petal.Length, color = "blue")) + geom_point()
```





# Aesthetics are for mapping only

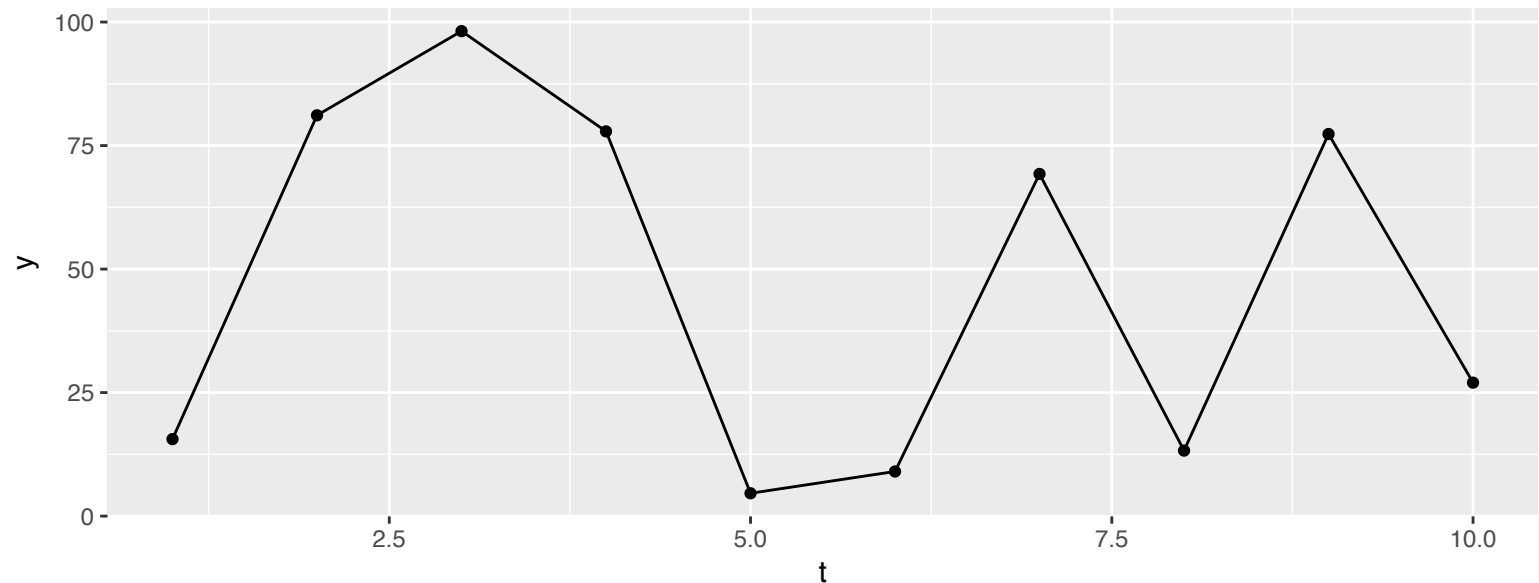
- ~~### Color all points blue?~~
- ~~ggplot(iris, aes(x = Sepal.Length, y = Petal.Length, color = "blue")) + geom\_point()~~
- > ### Correctly color all points blue
- > ggplot(iris, aes(x = Sepal.Length, y = Petal.Length)) + geom\_point(color = "blue")



# Example: multiple geoms

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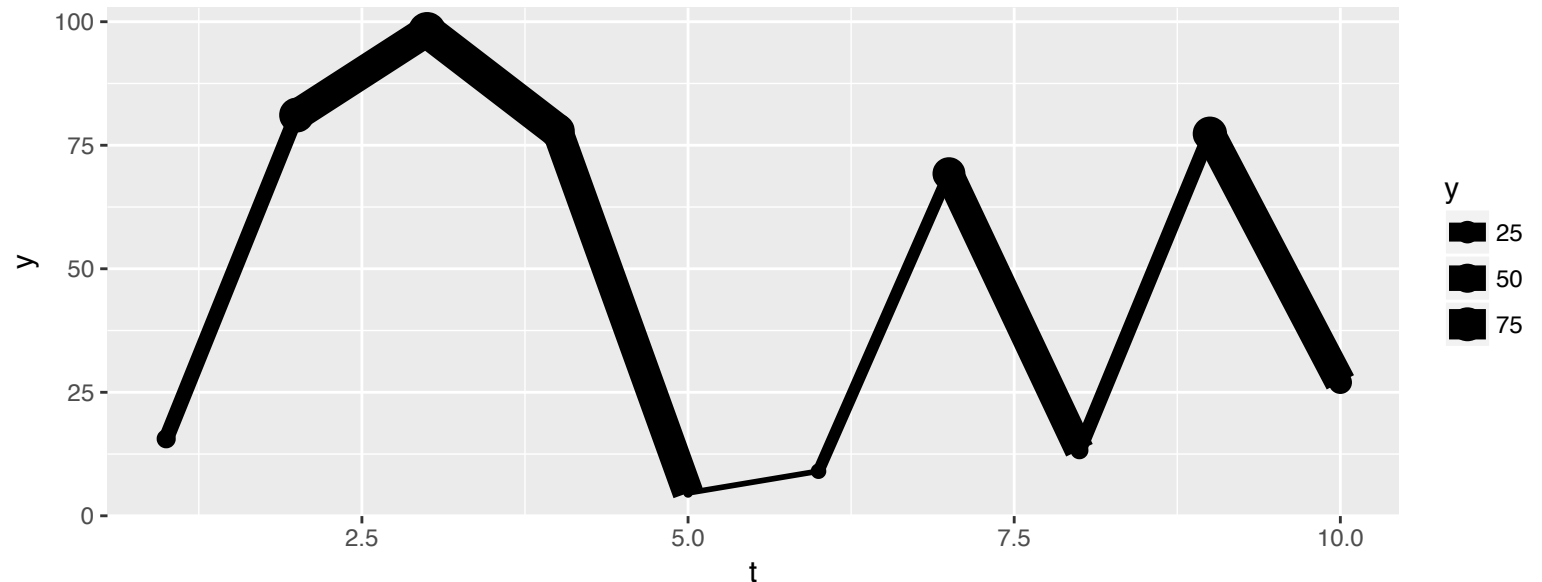
```
> ### Use some fake data:  
> fake.data <- data.frame(t = 1:10, y = runif(10, 1, 100))  
  
> ggplot(fake.data, aes(x = t, y = y)) + geom_point() + geom_line()
```



# Make sure aesthetic mappings are properly applied

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```
> ggplot(fake.data, aes(x = t, y = y, size = y)) + geom_point() + geom_line()
```

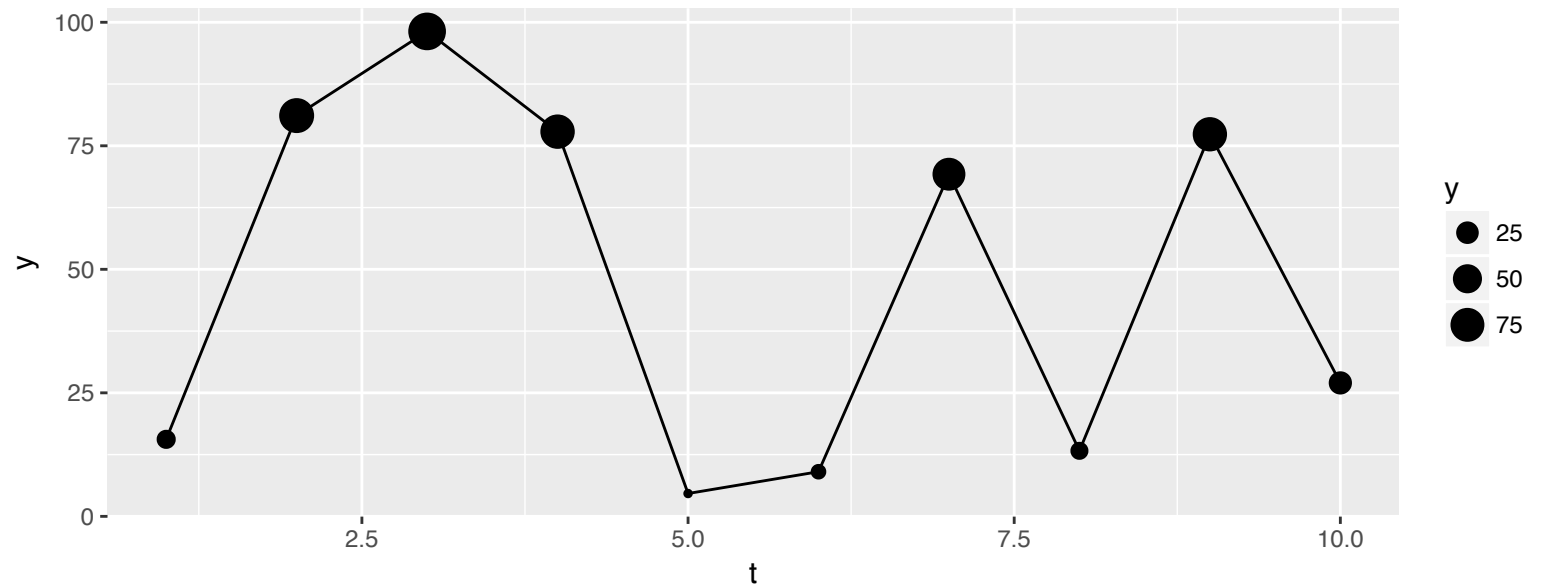


# Make sure aesthetic mappings are properly applied

---

~~> ggplot(fake.data, aes(x = t, y = y, size = y)) + geom\_point() + geom\_line()~~

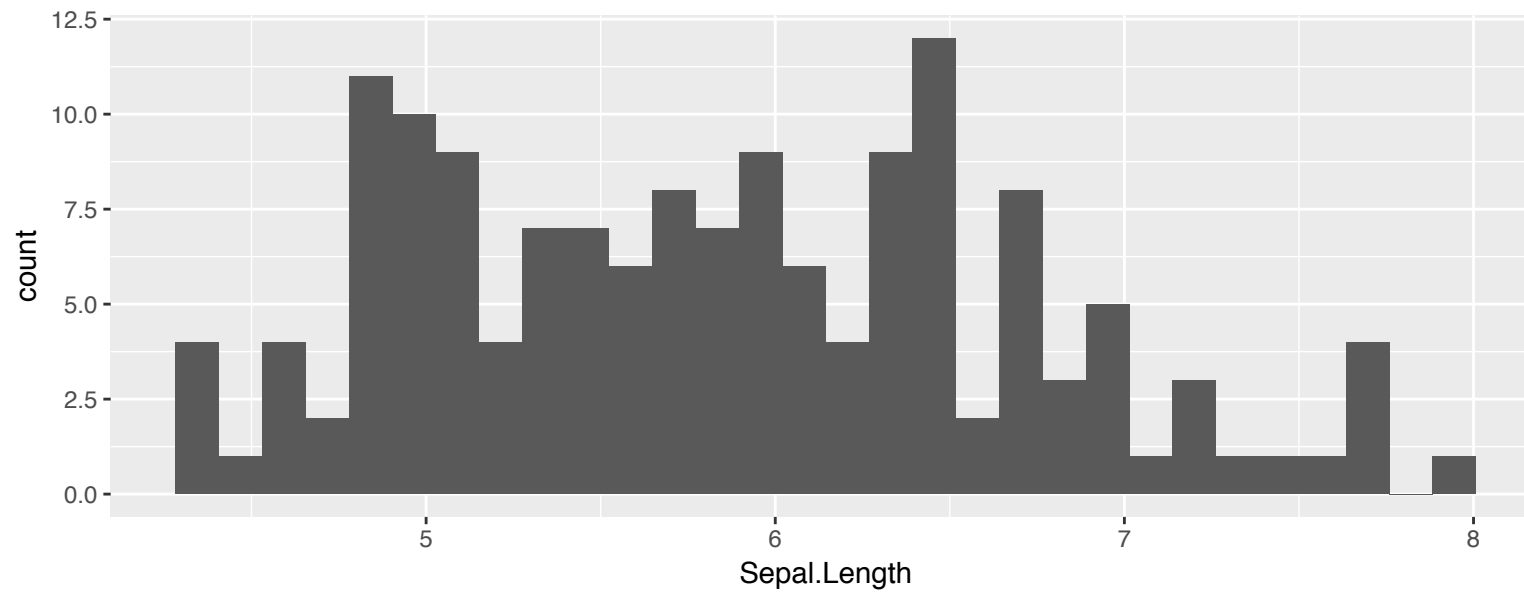
> ggplot(fake.data, aes(x = t, y = y)) + geom\_point( aes(size=y) ) + geom\_line()



# Histograms

---

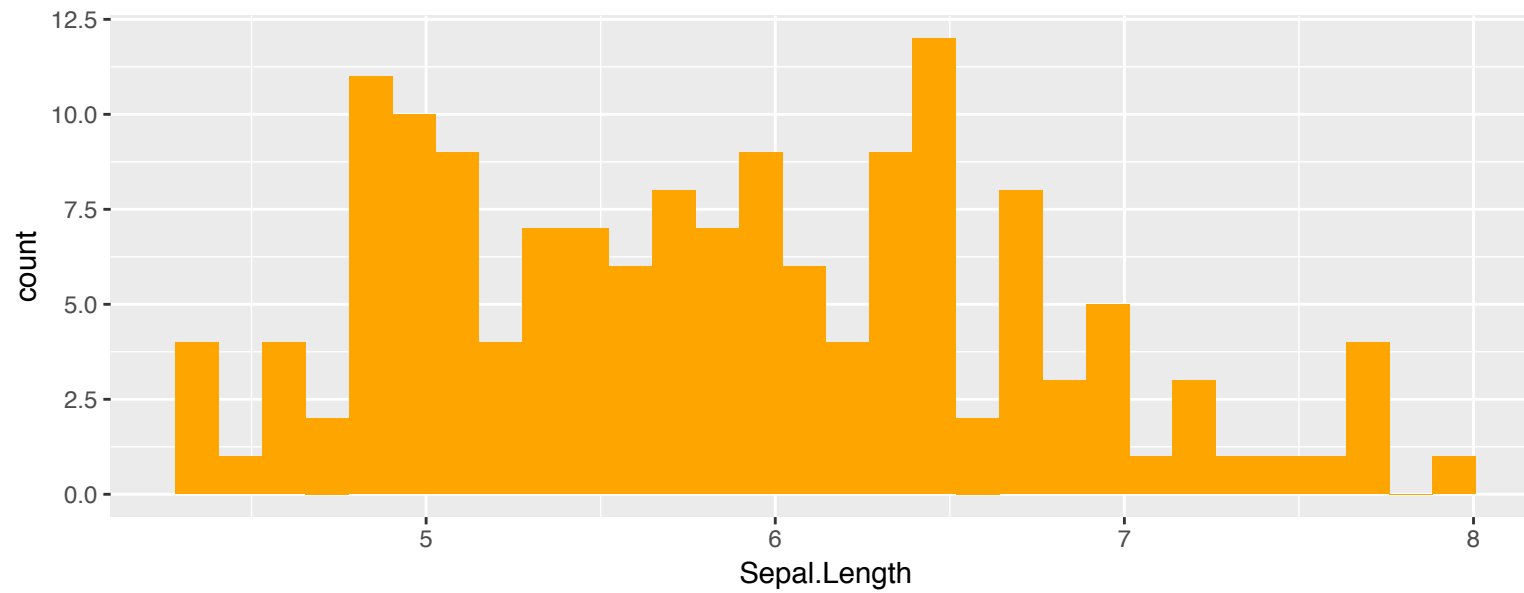
```
> ggplot(iris, aes(x = Sepal.Length)) + geom_histogram()
```



# Histograms

---

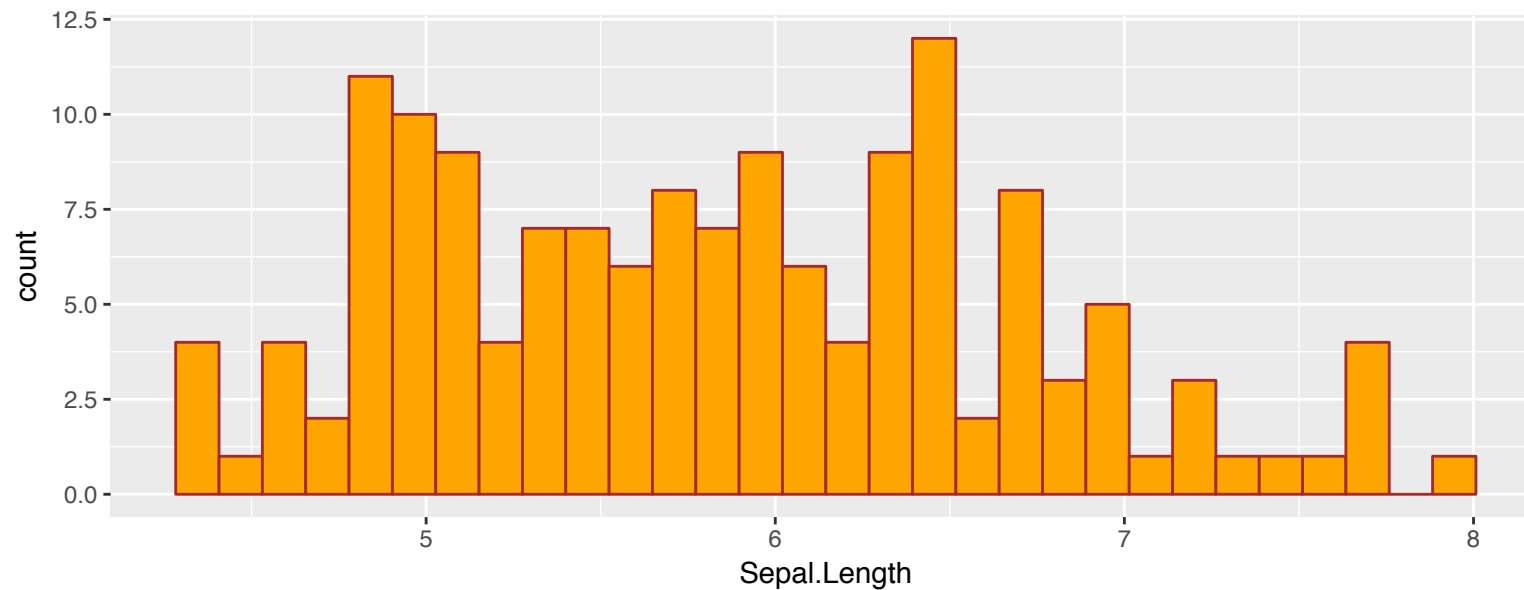
```
> ggplot(iris, aes(x = Sepal.Length)) + geom_histogram( fill = "orange" )
```



# Histograms

---

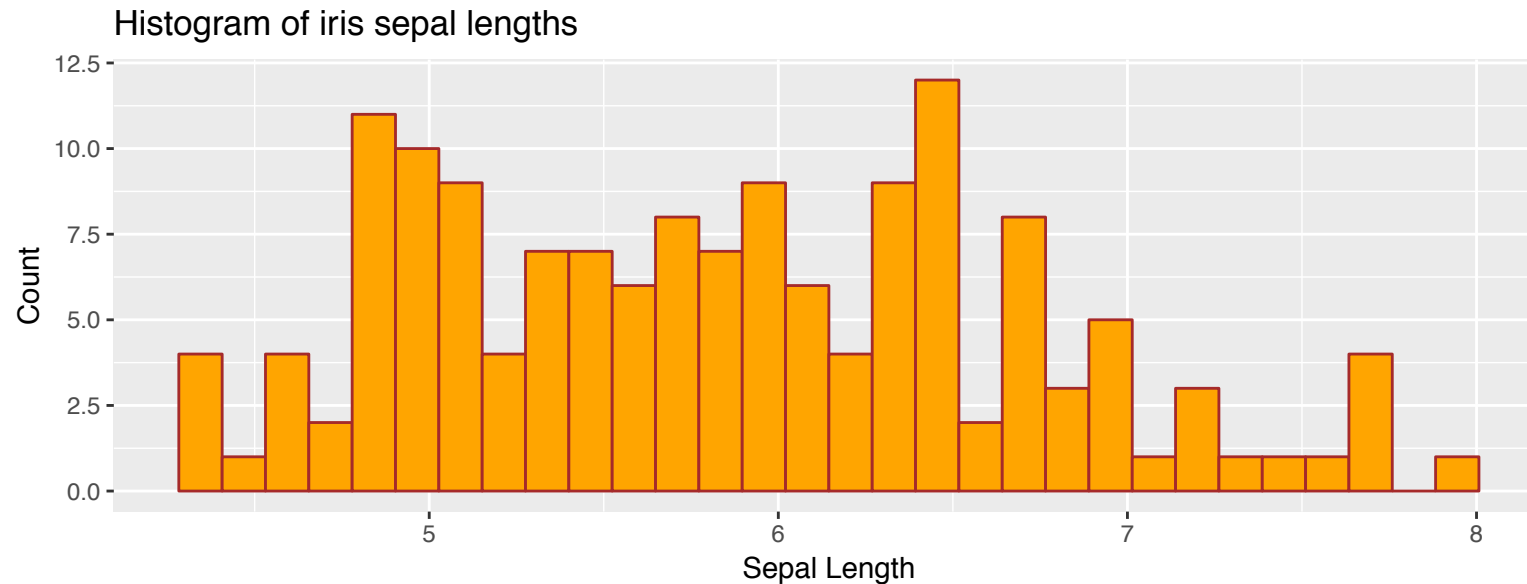
```
> ggplot(iris, aes(x = Sepal.Length)) + geom_histogram( fill = "orange", color = "brown" )
```



# Histograms

---

```
> ggplot(iris, aes(x = Sepal.Length)) + geom_histogram( fill = "orange", color = "brown" )  
+ xlab("Sepal Length") + ylab("Count") + ggtitle("Histogram of iris sepal lengths")
```

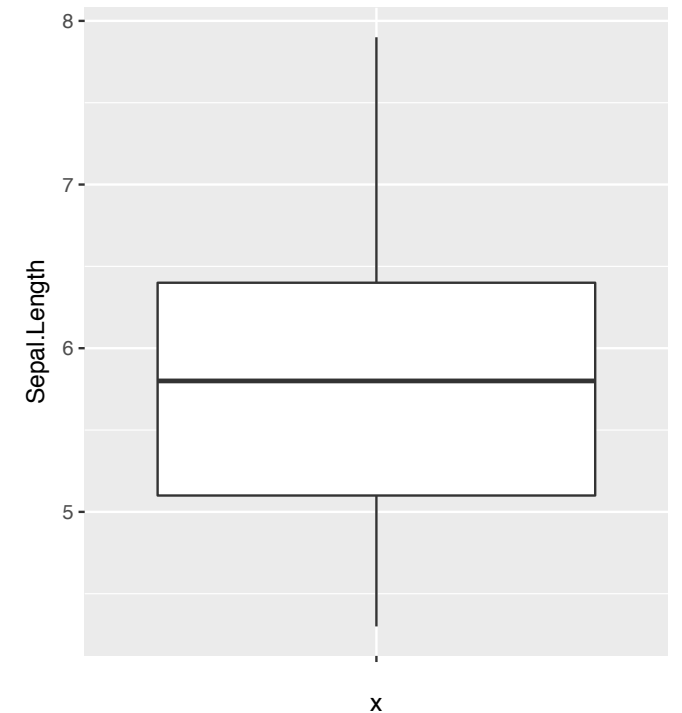




# Boxplots

---

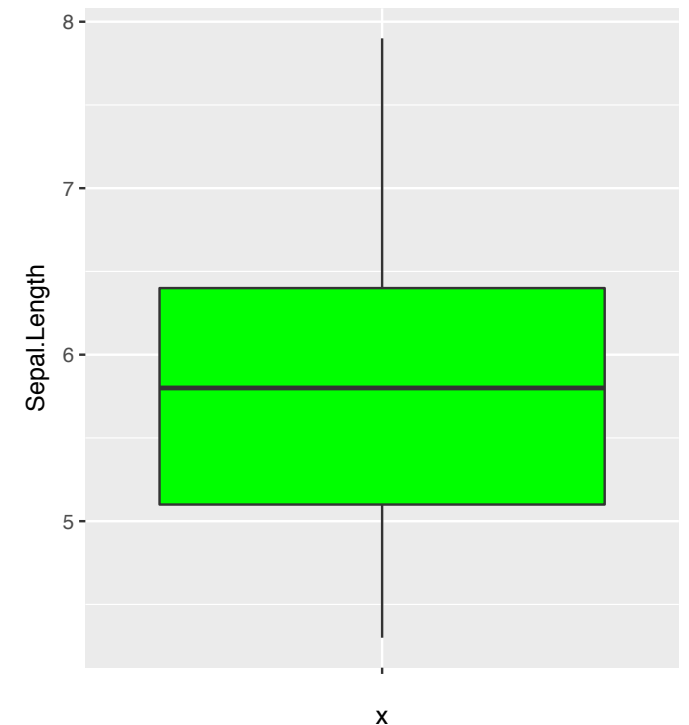
```
> ggplot(iris, aes(x = "", y = Sepal.Length)) + geom_boxplot()
```



# Boxplots

---

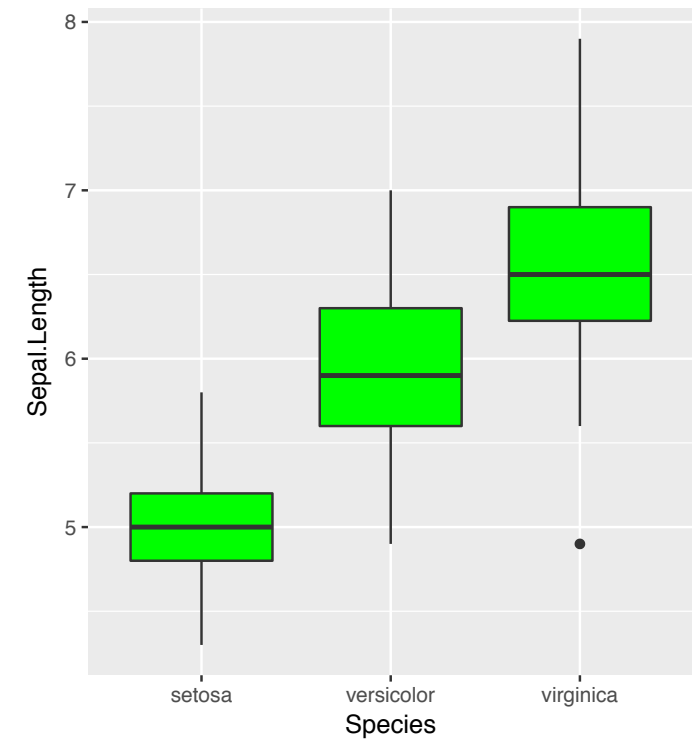
```
> ggplot(iris, aes(x = "", y = Sepal.Length)) + geom_boxplot(fill = "green")
```



# Boxplots

---

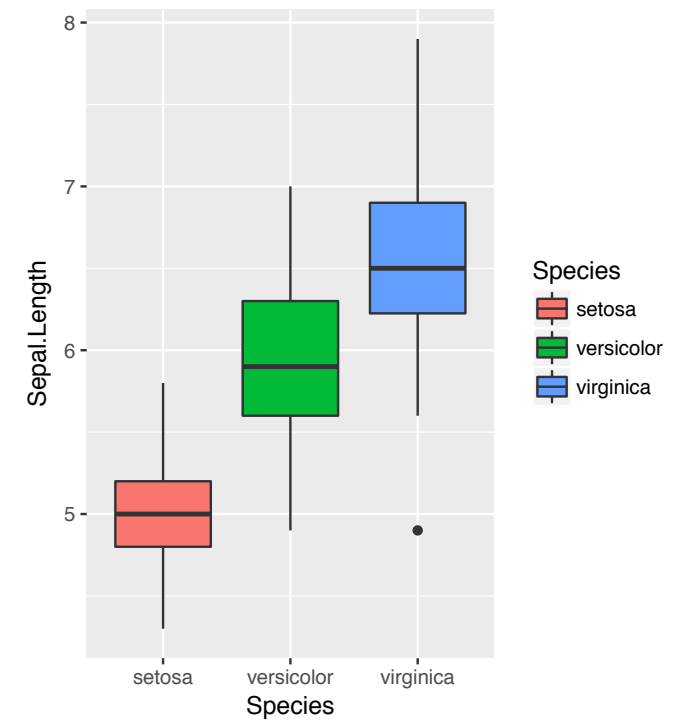
```
> ggplot(iris, aes(x = Species, y = Sepal.Length)) + geom_boxplot(fill = "green")
```



# Boxplots

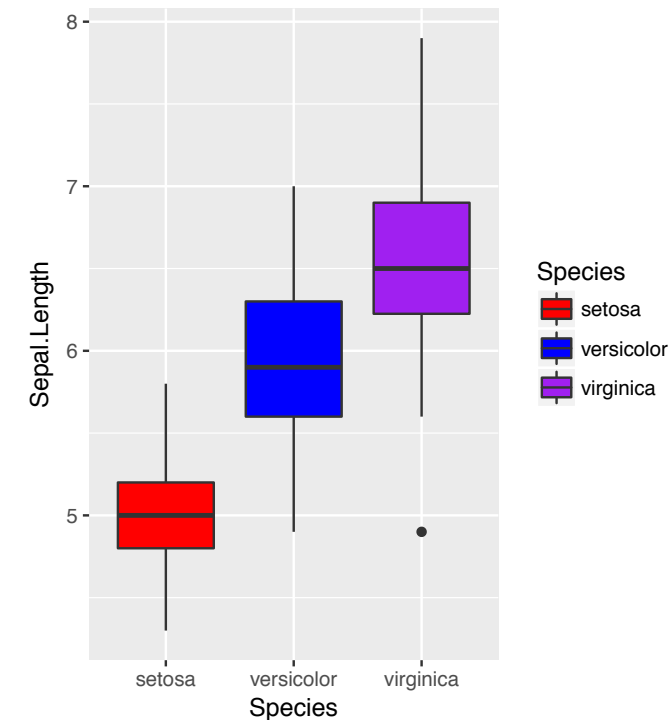
---

```
> ggplot(iris, aes(x = Species, y = Sepal.Length, fill = Species)) + geom_boxplot()
```



# Boxplots: Customizing the fill mappings

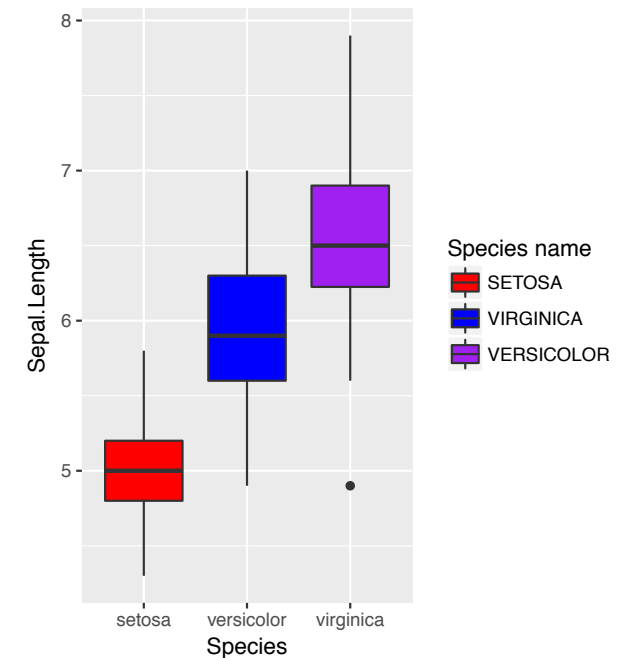
```
> ggplot(iris, aes(x = Species, y = Sepal.Length, fill = Species)) + geom_boxplot() +  
  scale_fill_manual(values=c("red", "blue", "purple"))
```



# scale\_fill\_manual() also tweaks legend

---

```
> ggplot(iris, aes(x = Species, y = Sepal.Length, fill = Species)) + geom_boxplot() +  
  scale_fill_manual(values=c("red", "blue", "purple"), name = "Species name",  
  labels=c("SETOSA", "VIRGINICA", "VERSICOLOR"))
```



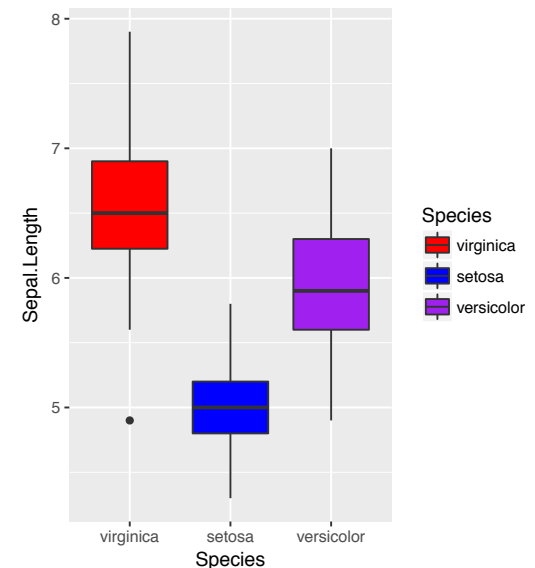
# Changing the order

---

```
> ### Ordering depends on factor levels
> levels(iris$Species)
[1] "setosa"      "versicolor" "virginica"

> ### Change order of levels
> iris$Species <- factor(iris$Species, levels=c("virginica", "setosa", "versicolor"))
[1] "virginica"  "setosa"     "versicolor"

> ### Replot
> ggplot(iris, aes(x = Species, y = Sepal.Length, fill = Species)) +
  geom_boxplot() +
  scale_fill_manual(values=c("red", "blue", "purple"))
```



# Grouped boxplots

---

This will apply to violin plots as well.

```
> ## Create another categorical variable for grouping purposes
> iris %>%
  group_by(Species) %>%
  mutate(size = ifelse( Sepal.Width > median(Sepal.Width) , "big" , "small" )) -> iris2
                        Condition                Value if TRUE      Value if FALSE
> head(iris2)
Source: local data frame [150 x 6]
Groups: Species [3]
```

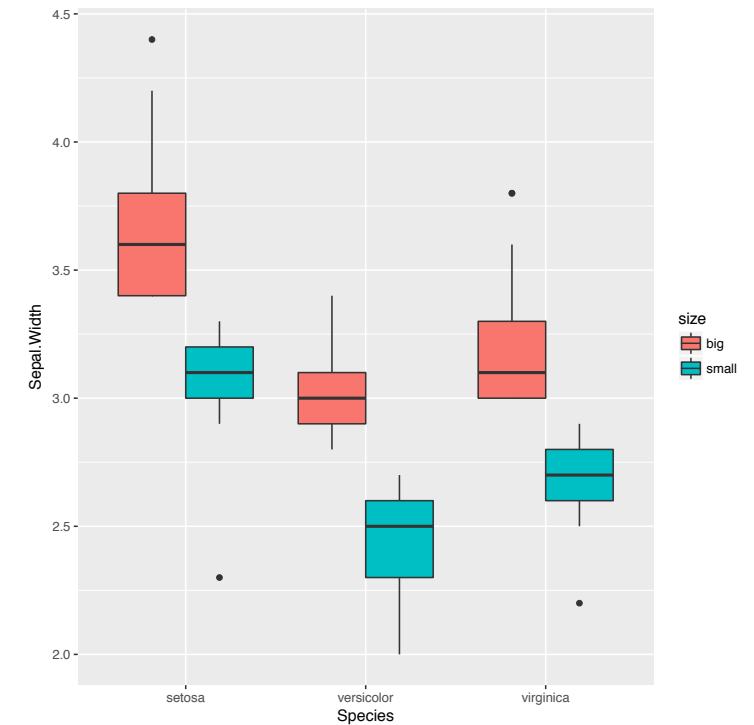
	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species	size
	<dbl>	<dbl>	<dbl>	<dbl>	<fctr>	<chr>
1	5.1	3.5	1.4	0.2	setosa	big
2	4.9	3.0	1.4	0.2	setosa	small
3	4.7	3.2	1.3	0.2	setosa	small
4	4.6	3.1	1.5	0.2	setosa	small
5	5.0	3.6	1.4	0.2	setosa	big
6	5.4	3.9	1.7	0.4	setosa	big



# Grouped boxplots

---

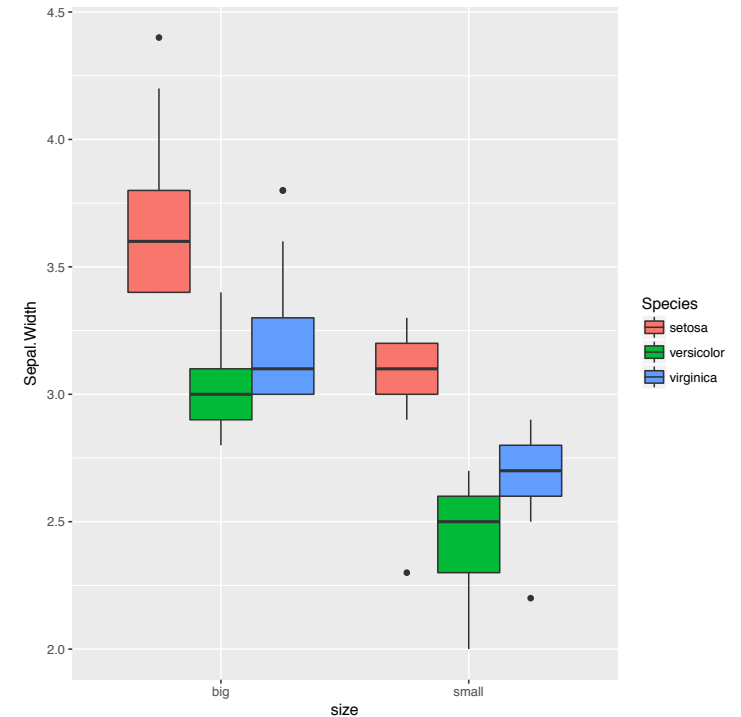
```
> ggplot(iris2, aes( x = Species, fill=size, y=Sepal.Width)) + geom_boxplot()
```



# Grouped boxplots

---

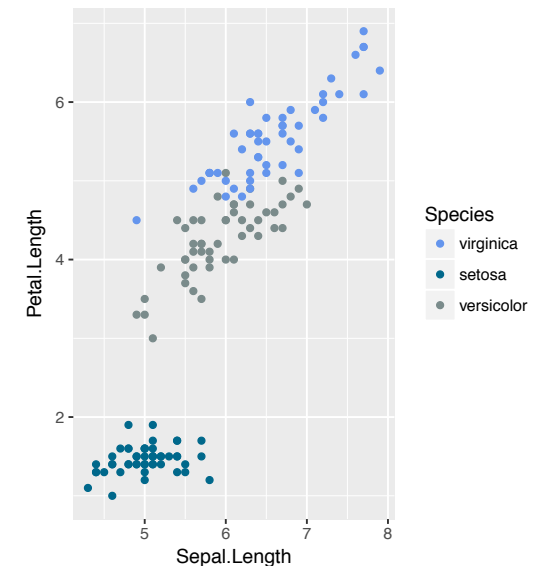
```
> ggplot(iris2, aes( x = size, fill = Species, y=Sepal.Width)) + geom_boxplot()
```



# Detour: scale\_color\_manual() customizes color

---

```
> ggplot(iris, aes(x = Sepal.Length, y = Petal.Length)) +  
  geom_point(aes(color = Species)) +  
  scale_color_manual(values=c("cornflowerblue", "deepskyblue4", "lightcyan4"))
```



# Detour round 2: `scale_<fill/color>_??`

---

There are **many** scales to use besides default and custom.

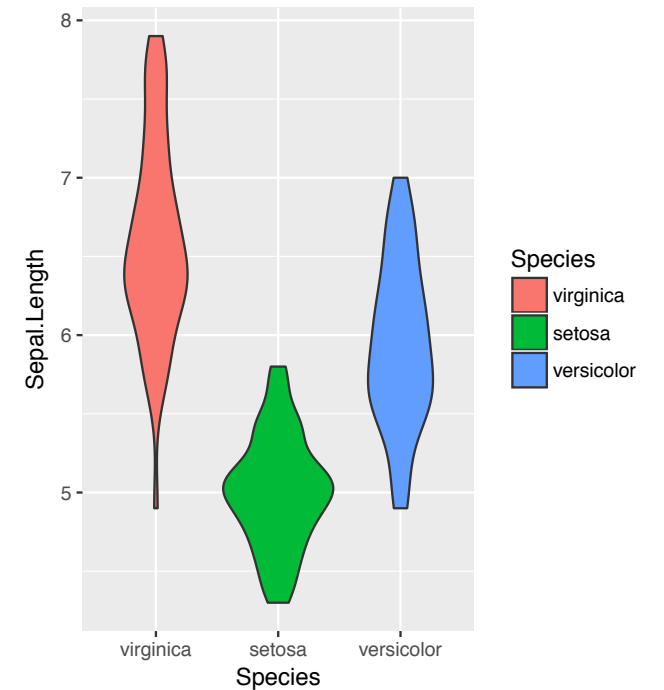
- `scale_<fill/color>_brewer()` uses pre-made color schemes from [colorbrewer.org](http://colorbrewer.org)
- `scale_color_gradient()` can take a low and high to fill along a spectrum

See here: <http://ggplot2.tidyverse.org/reference/#scales>

# Violin plot

---

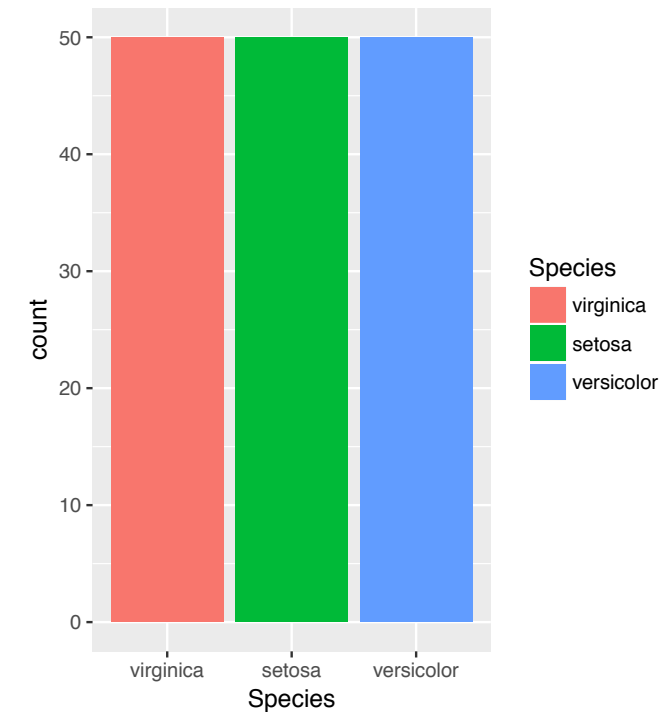
```
> ggplot(iris, aes(x = Species, y = Sepal.Length, fill = Species)) + geom_violin()
```



# Bar plot

---

```
> ggplot(iris, aes(x = Species, fill = Species)) + geom_bar()
```



# Stacked/grouped bar plot

---

```
> head(iris2)
```

```
Source: local data frame [150 x 6]
```

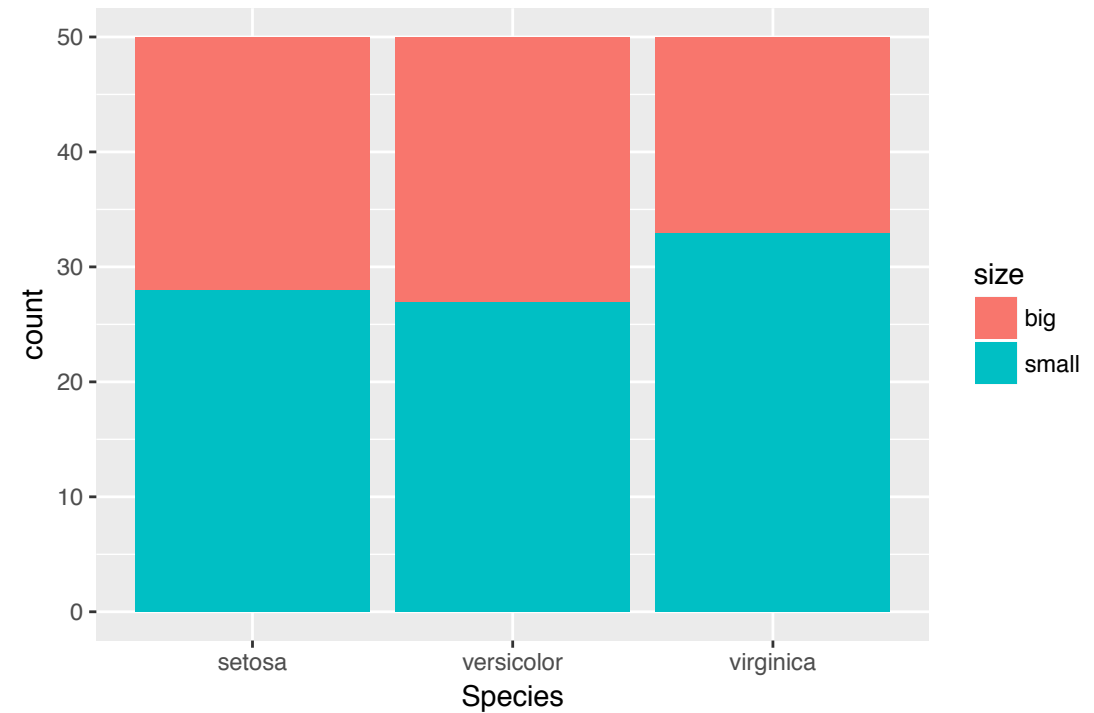
```
Groups: Species [3]
```

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species	size
	<dbl>	<dbl>	<dbl>	<dbl>	<fctr>	<chr>
1	5.1	3.5	1.4	0.2	setosa	big
2	4.9	3.0	1.4	0.2	setosa	small
3	4.7	3.2	1.3	0.2	setosa	small
4	4.6	3.1	1.5	0.2	setosa	small
5	5.0	3.6	1.4	0.2	setosa	big
6	5.4	3.9	1.7	0.4	setosa	big

# Stacked/grouped bar plot

---

```
> ggplot(iris, aes(x = Species, fill = size)) + geom_bar()
```

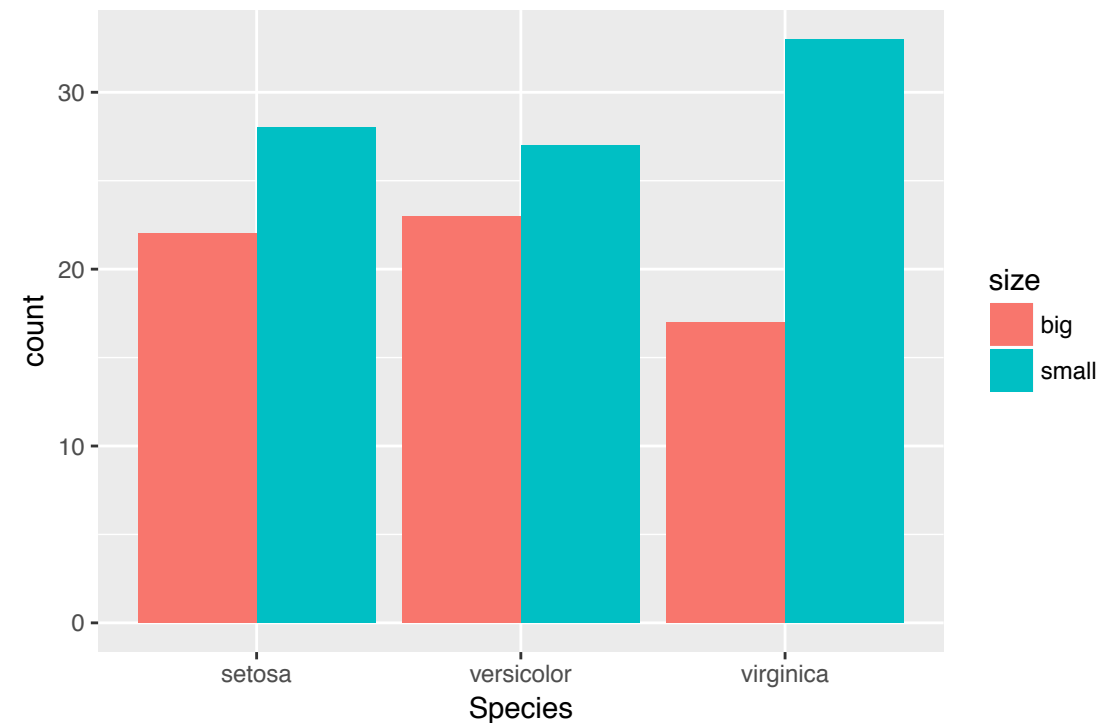




# Stacked/grouped bar plot

---

```
> ggplot(iris, aes(x = Species, fill = size)) + geom_bar( position = "dodge" )
```

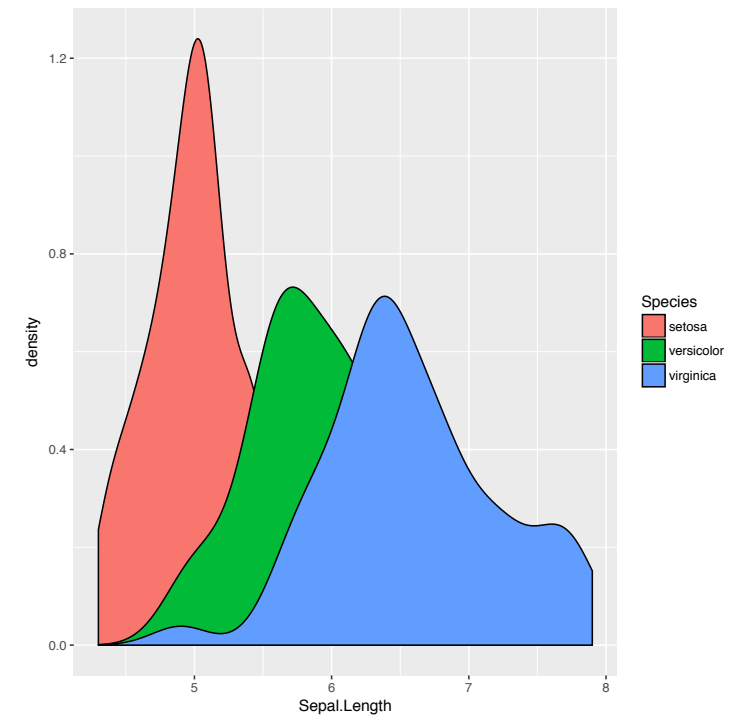


# Density plot

---

```
> ggplot(iris, aes(x = Sepal.Length, fill = Species)) + geom_density()
```

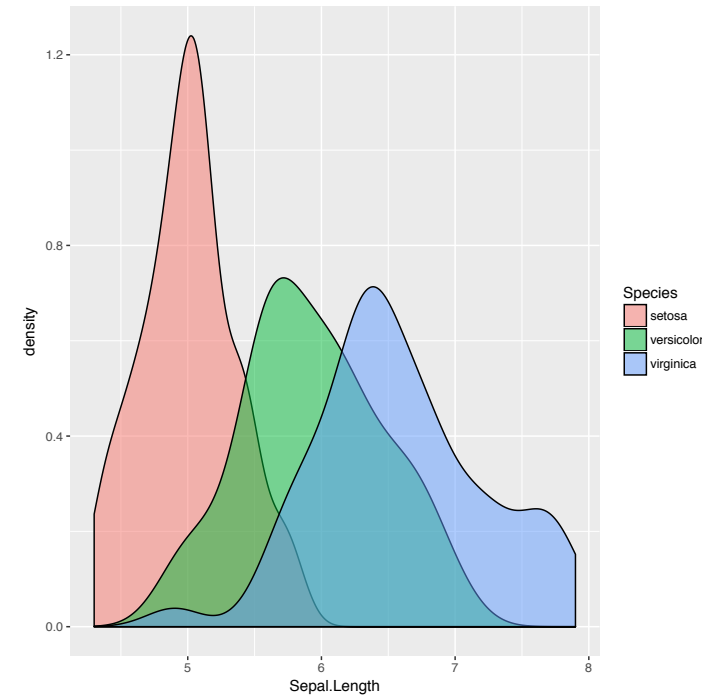
What does the tail of the setosa distribution look like?



# Density plot

---

```
> ggplot(iris, aes(x = Sepal.Length, fill = Species)) + geom_density( alpha = 0.5 )
```



# Themes

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Gray background and grid not working for you? Me neither.

- Built-in other themes: <http://ggplot2.tidyverse.org/reference/ggtheme.html>
- Customize your theme: <http://ggplot2.tidyverse.org/reference/theme.html>
- Use somebody else's themes:
  - <https://cran.r-project.org/web/packages/ggthemes/vignettes/ggthemes.html>
  - <https://cran.r-project.org/web/packages/cowplot/vignettes/introduction.html>