Scatter plots

INTRODUCTION TO DATA VISUALIZATION WITH GGPLOT2



Rick Scavetta Founder, Scavetta Academy



48 geometries

geom_*						
abline	contour	dotplot	jitter	pointrange	ribbon	spoke
area	count	errorbar	label	polygon	rug	step
bar	crossbar	errorbarh	line	qq	segment	text
bin2d	curve	freqpoly	linerange	qq_line	sf	tile
blank	density	hex	map	quantile	sf_label	violin
boxplot	density2d	histogram	path	raster	sf_text	vline
col	density_2d	hline	point	rect	smooth	

Common plot types

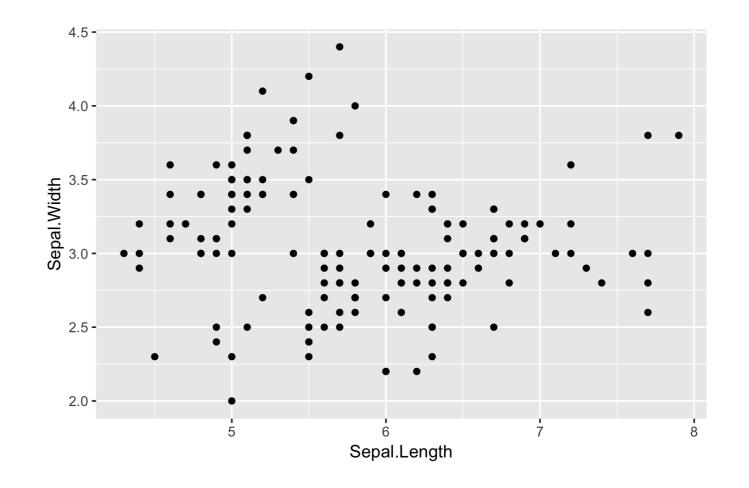
Plot type	Possible Geoms
Scatter plots	points, jitter, abline, smooth, count

Scatter plots

• Each geom can accept specific aesthetic mappings, e.g. geom_point():

Essential

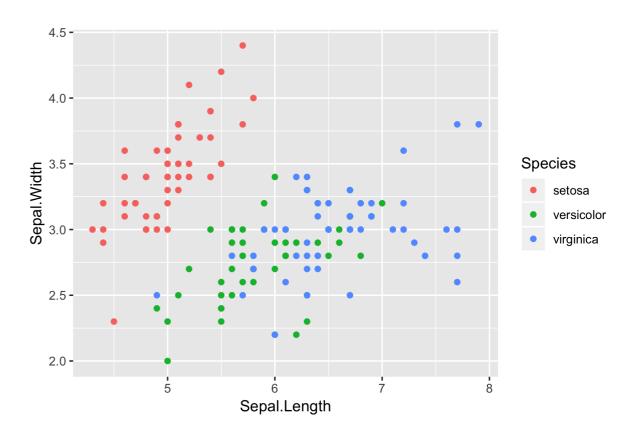
x,y



Scatter plots

• Each geom can accept specific aesthetic mappings, e.g. geom_point():

Essential	Optional
x,y	alpha, color, fill, shape, size, stroke

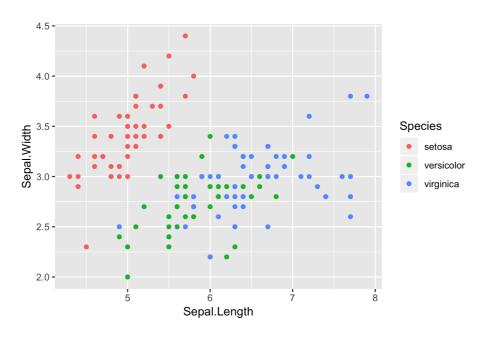


Geom-specific aesthetic mappings

```
# These result in the same plot!
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, col = Species)) +
  geom_point()

ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width)) +
  geom_point(aes(col = Species))
```

Control aesthetic mappings of each layer independently:



```
head(iris, 3) # Raw data
```

```
      Species Sepal.Length Sepal.Width Petal.Length Petal.Width

      1 setosa
      5.1
      3.5
      1.4
      0.2

      2 setosa
      4.9
      3.0
      1.4
      0.2

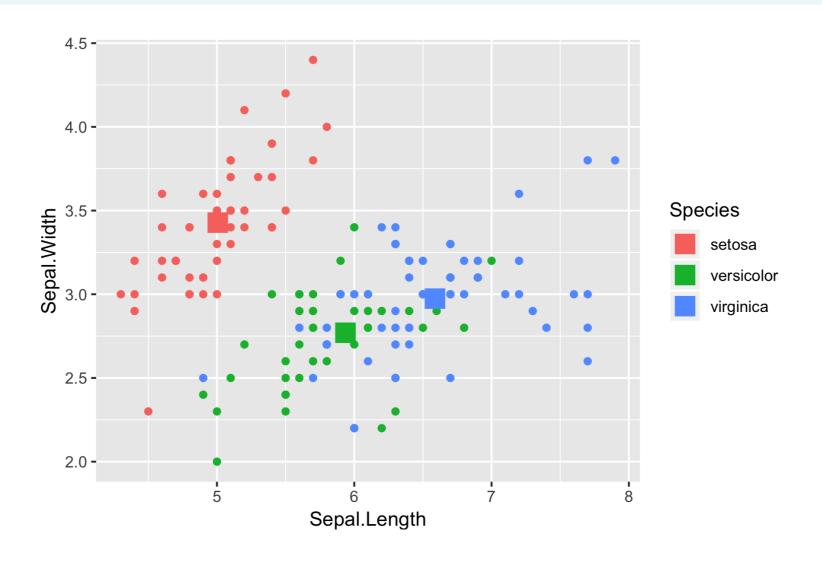
      3 setosa
      4.7
      3.2
      1.3
      0.2
```

```
iris %>%
  group_by(Species) %>%
  summarise_all(mean) -> iris.summary

iris.summary # Summary statistics
```

```
# A tibble: 3 x 5
         Sepal.Length Sepal.Width Petal.Length Petal.Width
 Species
               <dbl>
                        <dbl>
 <fct>
                                  <dbl>
                                           <dbl>
         5.01 3.43
                                           0.246
 setosa
                                  1.46
2 versicolor 5.94 2.77
                                   4.26 1.33
3 virginica
               6.59
                         2.97
                                   5.55
                                           2.03
```

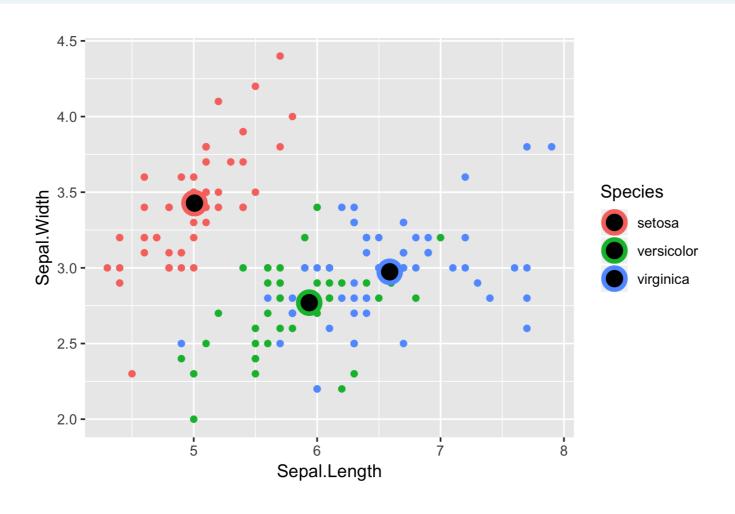
```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, col = Species)) +
    # Inherits both data and aes from ggplot()
    geom_point() +
    # Different data, but inherited aes
    geom_point(data = iris.summary, shape = 15, size = 5)
```



Shape attribute values

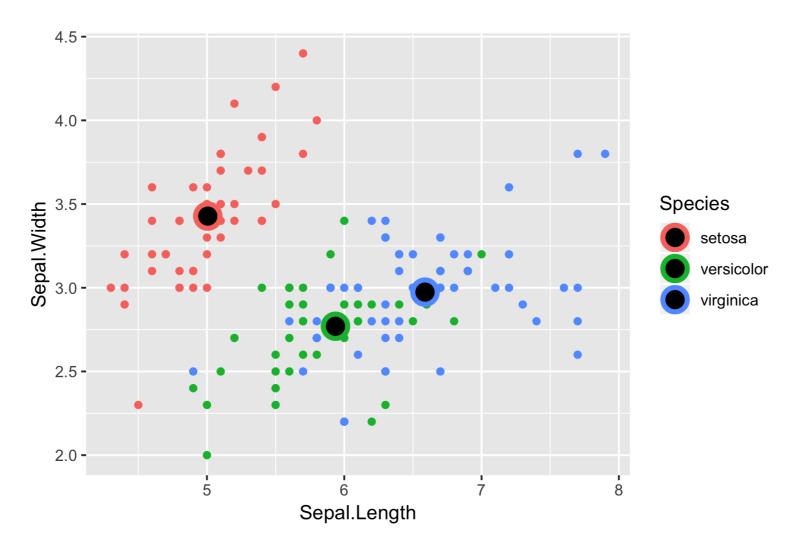
 \square 0 0 1 \triangle 2 + 3 \times \diamondsuit 5 ∇ 6 \boxtimes 7 + 8 \oplus \bigoplus 10 \bigotimes 11 \bigoplus 12 \bigotimes 13 \bigotimes **■** 15 **●** 16 **▲** 17 **♦** 18 **●**20
 21
 22
 23
 24
 25

Example



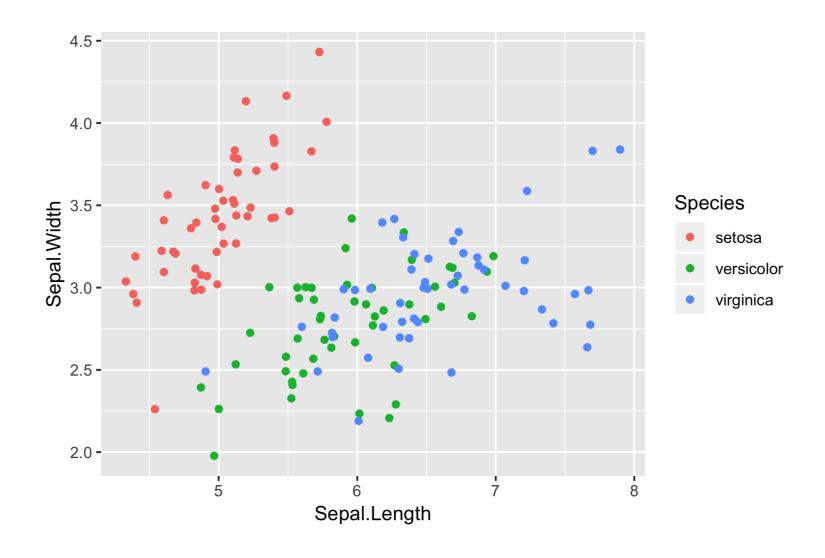
On-the-fly stats by ggplot2

- See the second course for the stats layer.
- Note: Avoid plotting only the mean without a measure of spread, e.g. the standard deviation.



position = "jitter"

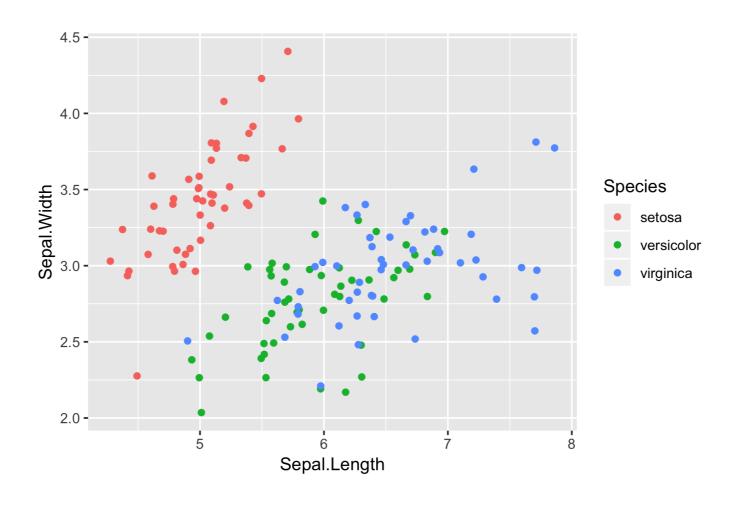
```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, col = Species)) +
  geom_point(position = "jitter")
```



geom_jitter()

A short-cut to geom_point(position = "jitter")

```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, col = Species)) +
  geom_jitter()
```

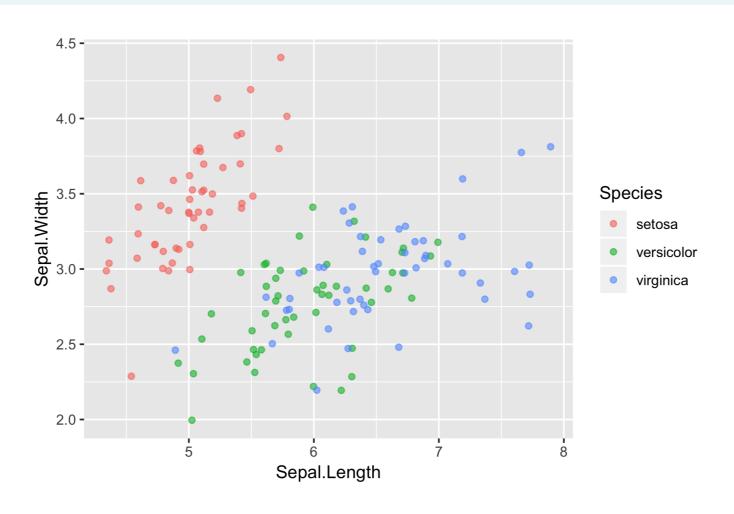




Don't forget to adjust alpha

Combine jittering with alpha-blending if necessary

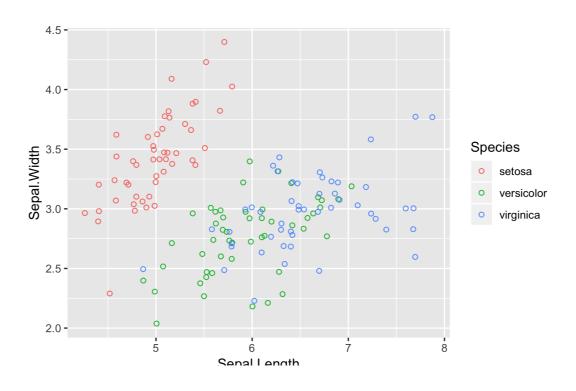
```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, col = Species)) + <math>geom\_jitter(alpha = 0.6)
```



Hollow circles also help

- shape = 1 is a. hollow circle.
- Not necessary to also use alpha-blending.

```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, col = Species)) + <math>geom_jitter(shape = 1)
```



Let's practice!

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Histograms

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Common plot types

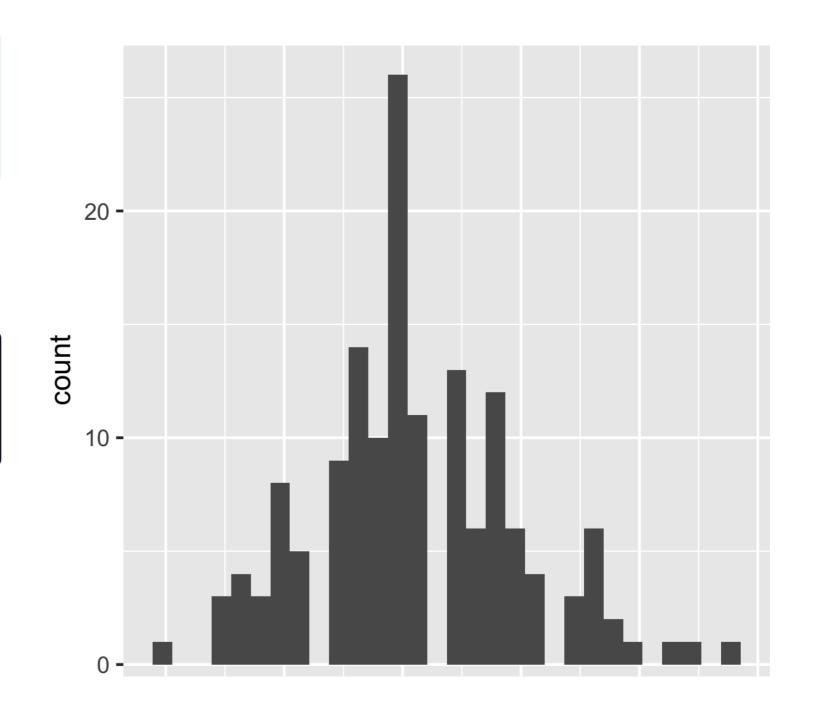
Plot type	Possible Geoms		
Scatter plots	points, jitter, abline, smooth, count		
Bar plots	histogram, bar, col, errorbar		
Line plots	line, path		

Histograms

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

- A plot of binned values
 - i.e. a statistical function

```
`stat_bin()` using `bins = 30`.
Pick better value with `binwidth`.
```



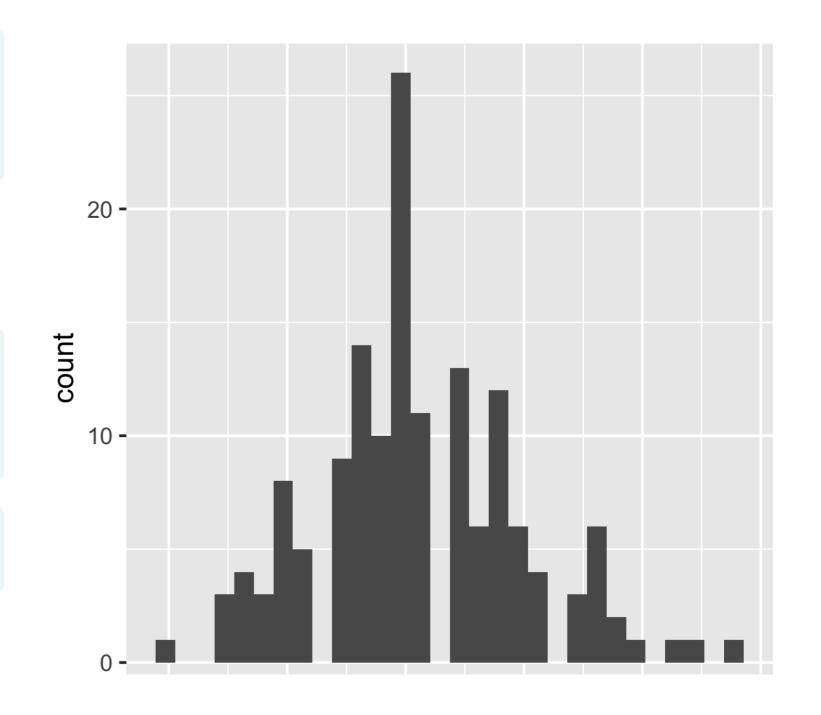
Default of 30 even bins

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

- A plot of binned values
 - i.e. a statistical function

```
# Default bin width:
diff(range(iris$Sepal.Width))/30
```

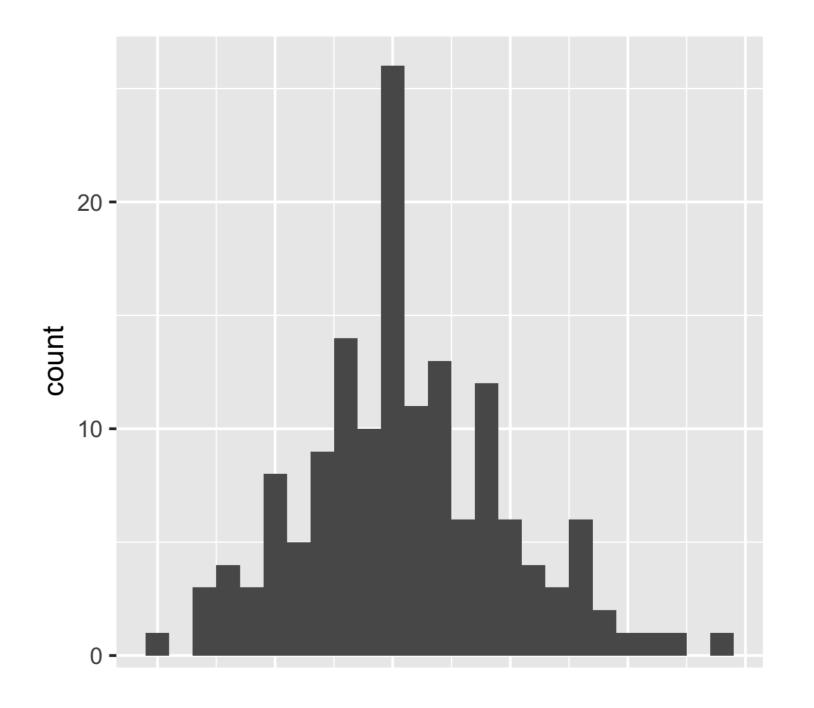
```
[1] 0.08
```



Intuitive and meaningful bin widths

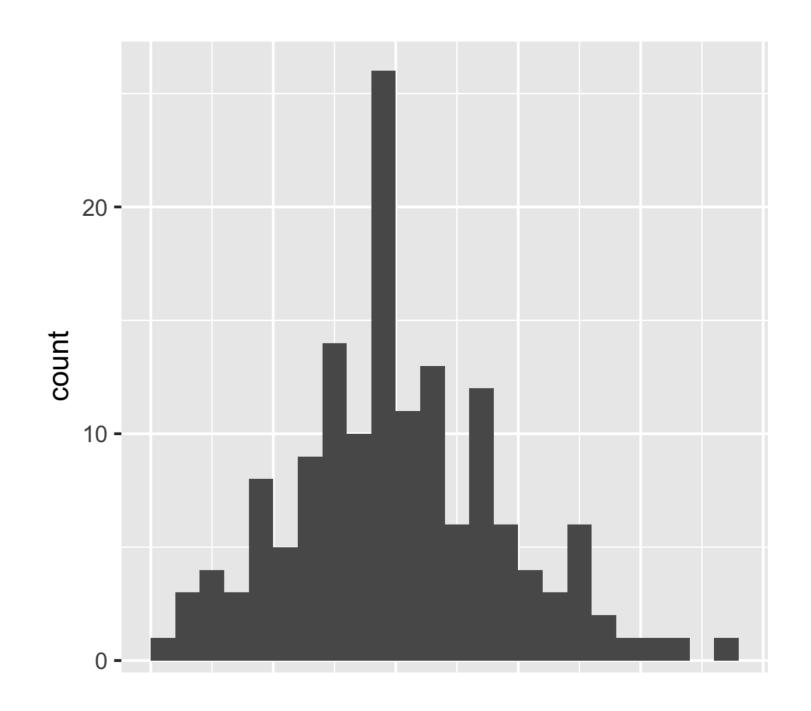
```
ggplot(iris, aes(x = Sepal.Width)) + geom_histogram(binwidth = 0.1)
```

- Always set a meaningful bin widths for your data.
- No spaces between bars.

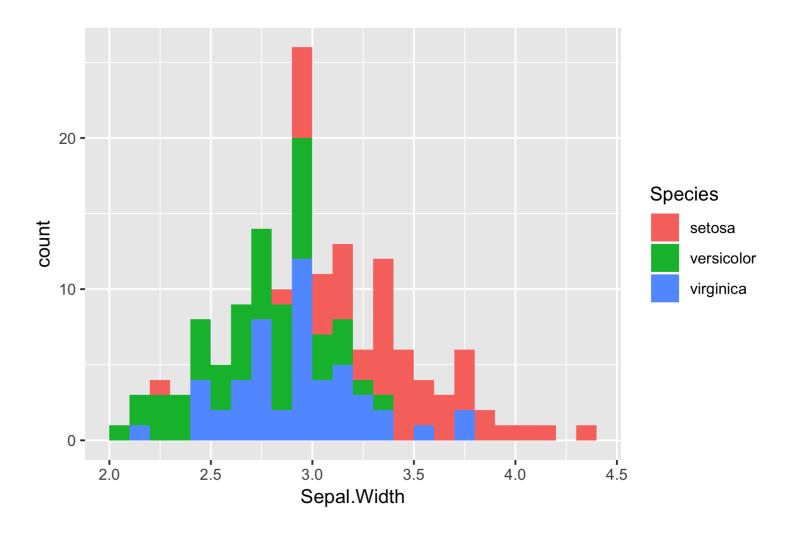


Re-position tick marks

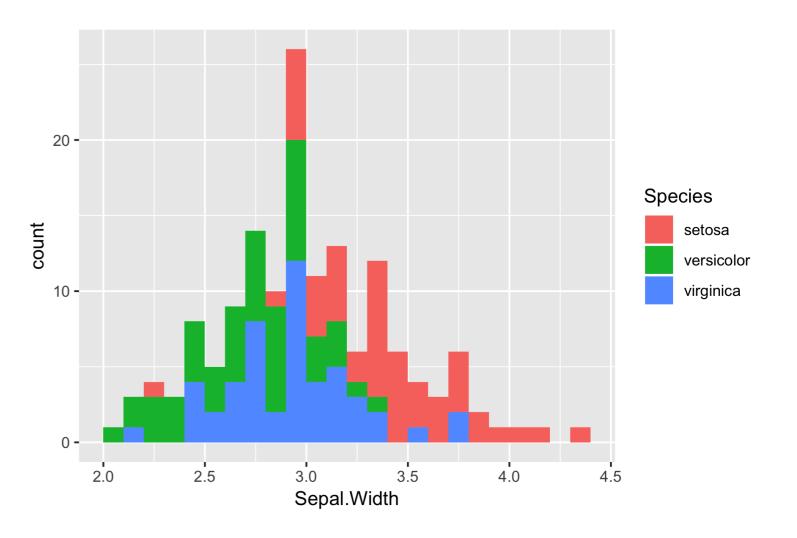
- Always set a meaningful bin widths for your data.
- No spaces between bars.
- X axis labels are between bars.



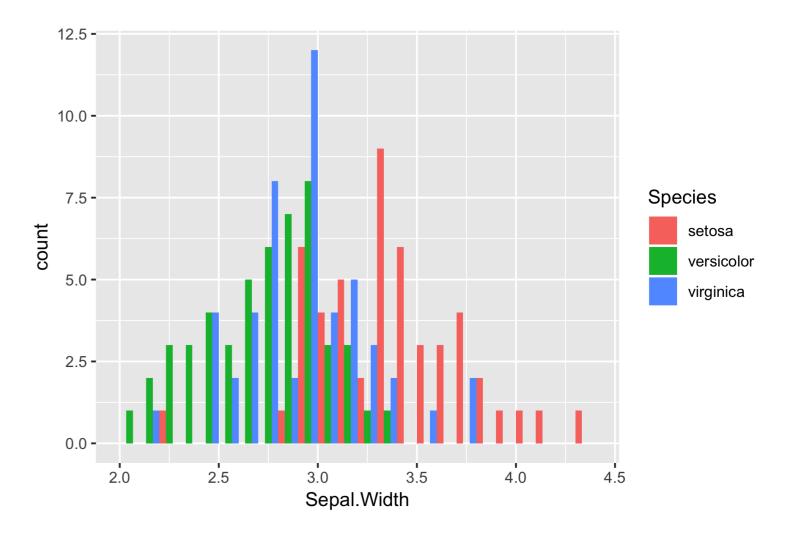
Different Species



Default position is "stack"

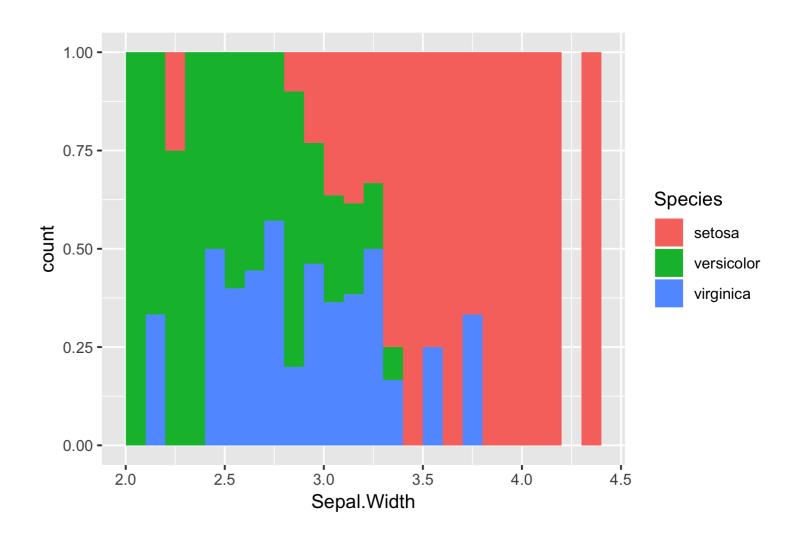


position = "dodge"



position = "fill"

```
ggplot(iris, aes(x = Sepal.Width, fill = S geom_histogram(binwidth = .1, center = 0
```



Final Slide

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Bar plots

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Bar Plots, with a categorical X-axis

Use geom_bar() or geom_col()

Geom	Stat	Action
<pre>geom_bar()</pre>	"count"	Counts the number of cases at each x position
geom_col()	"identity"	Plot actual values

- All positions from before are available
- Two types
 - Absolute counts
 - Distributions

Bar Plots, with a categorical X-axis

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Bar Plots, with a categorical X-axis

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- Two types
 - Absolute counts
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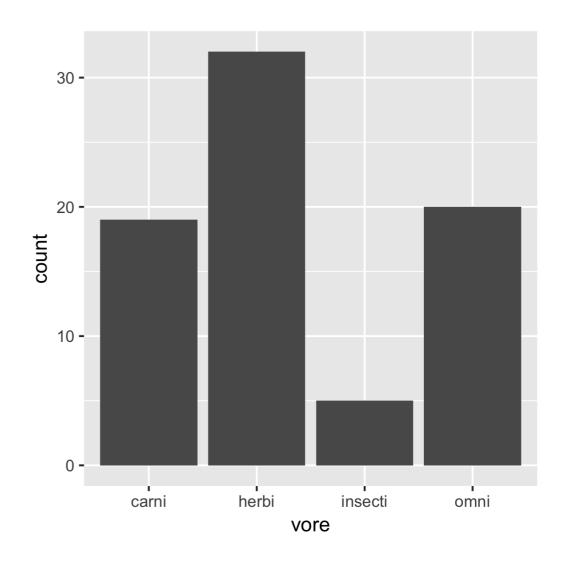
Habits of mammals

```
str(sleep)
```

```
'data.frame': 76 obs. of 3 variables:
$ vore : Factor w/ 4 levels "carni", "herbi", ...: 1 4 2 4 2 2 1 1 2 2 ...
$ total: num 12.1 17 14.4 14.9 4 14.4 8.7 10.1 3 5.3 ...
$ rem : num NA 1.8 2.4 2.3 0.7 2.2 1.4 2.9 NA 0.6 ...
```

Bar plot

```
ggplot(sleep, aes(vore)) +
  geom_bar()
```

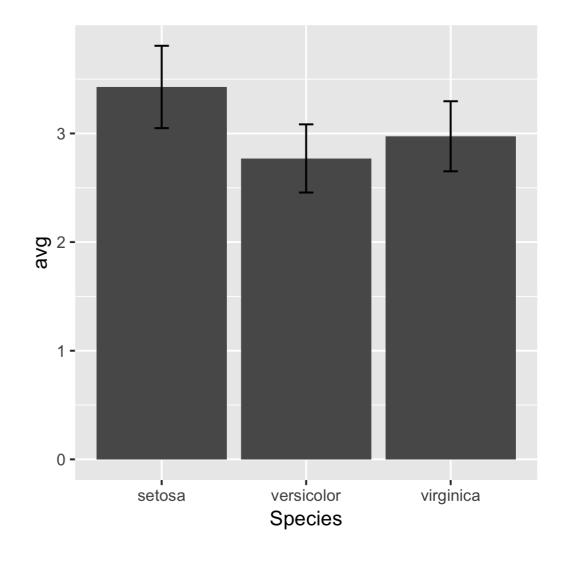


Plotting distributions instead of absolute counts

iris_summ_long

Species	avg	stdev
setosa	3.43	0.38
versicolor	2.77	0.31
virginica	2.97	0.32

Plotting distributions



Let's practice!

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Line plots

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Common plot types

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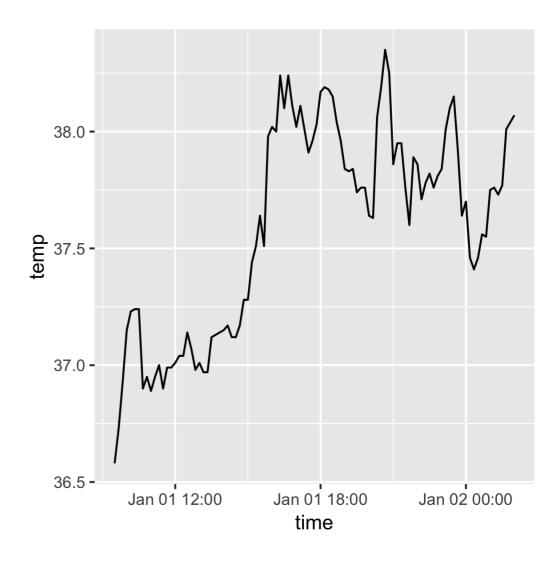
Beaver

str(beaver)

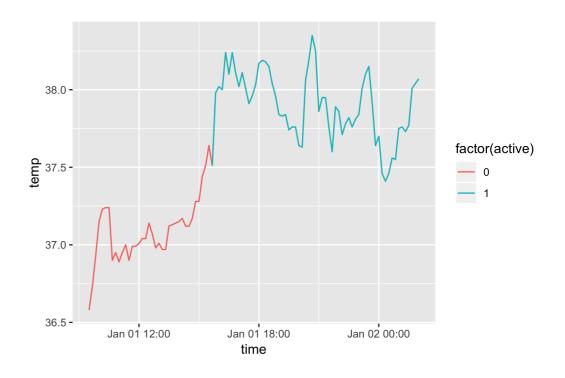
```
'data.frame': 101 obs. of 3 variables:
$ time : POSIXct, format: "2000-01-01 09:30:00" "2000-01-01 09:40:00" "2000-01-01 09:50  
$ temp : num 36.6 36.7 36.9 37.1 37.2 ...
$ active: Factor w/ 2 levels "0","1": 1 1 1 1 1 1 1 1 1 ...
```

Beaver

```
ggplot(beaver, aes(x = time, y = temp)) +
  geom_line()
```



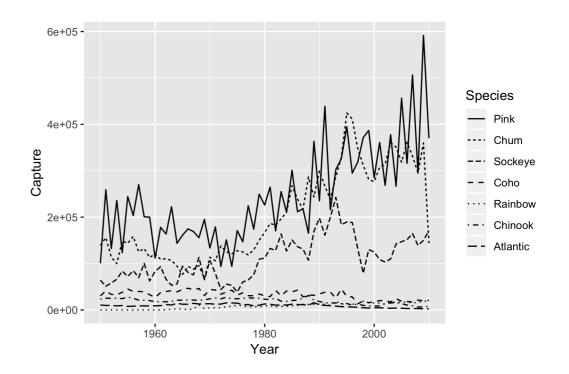
Beaver



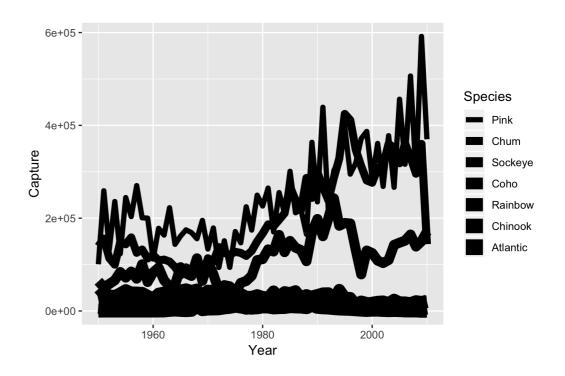
The fish catch dataset

str(fish)

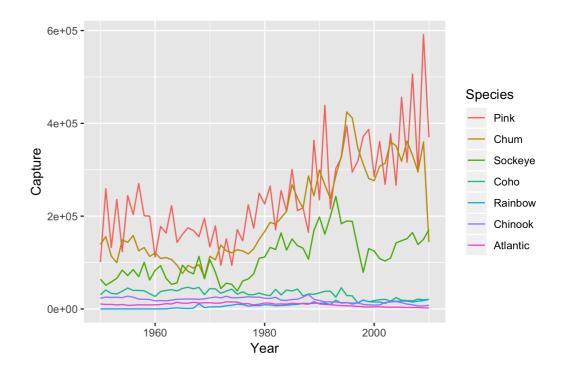
Linetype aesthetic



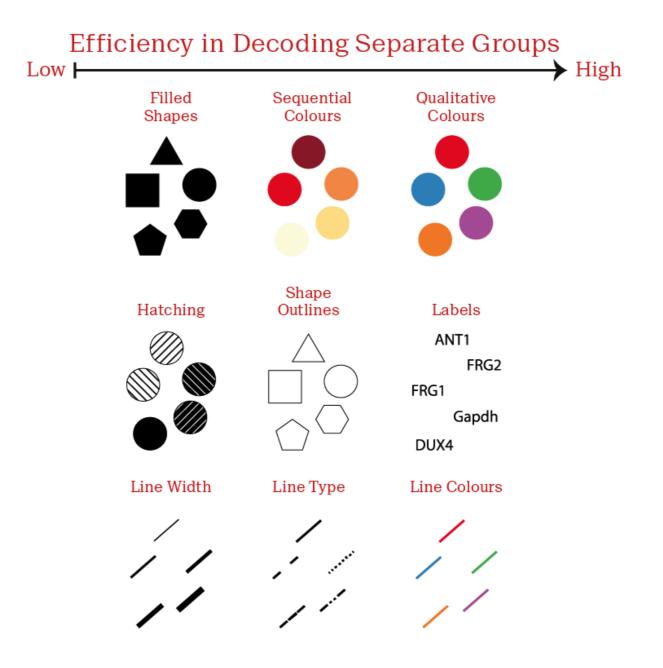
Size aesthetic



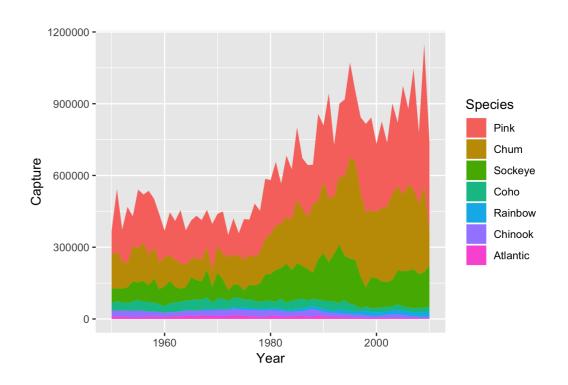
Color aesthetic



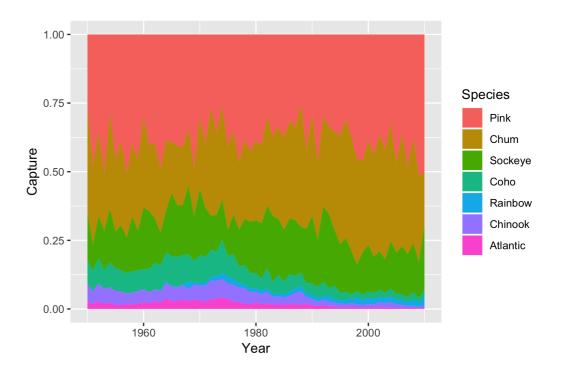
Aesthetics for categorical variables



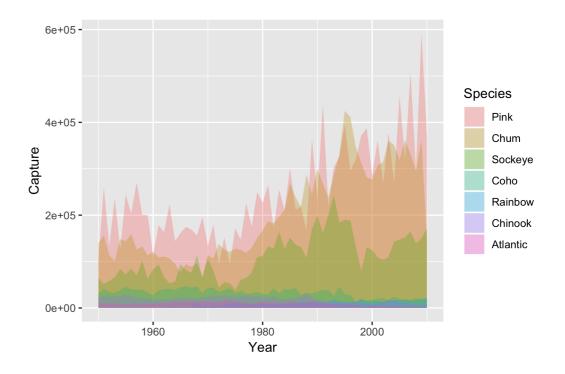
Fill aesthetic with geom_area()



Using position = "fill"



geom_ribbon()



Let's practice!

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