

WEEK 3

DATA VISUALIZATION BASICS

TODAY'S TOPICS

- Why data visualization
- Visual encoding
- An introduction to ggplot2: grammar of graphics

WHAT IS DATA VISUALIZATION?

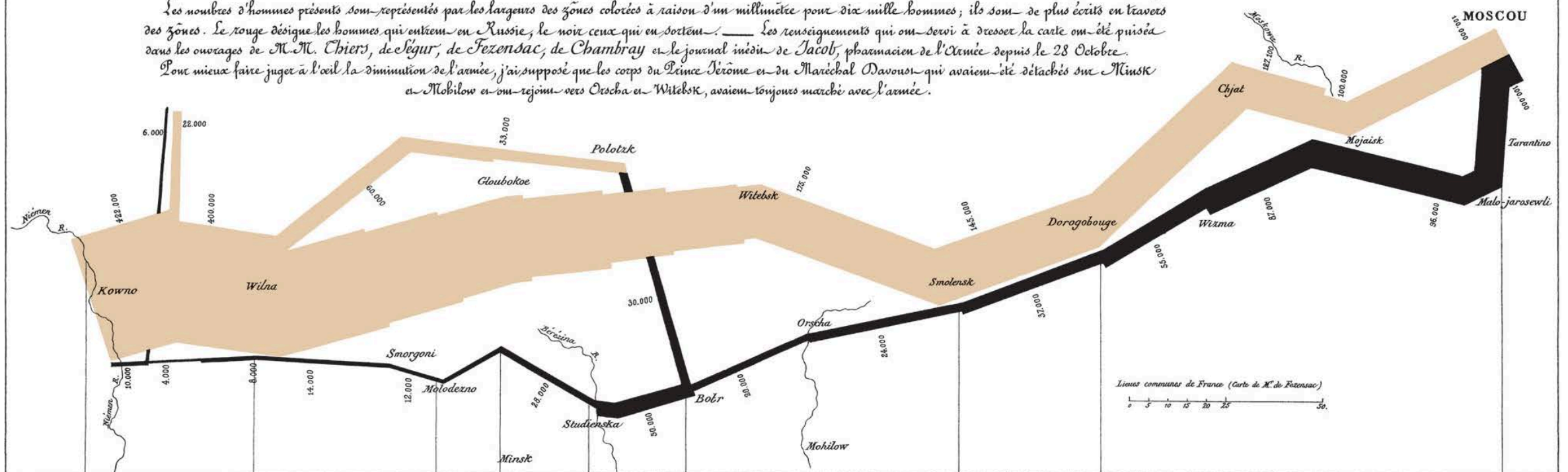
- Visual representation of information to help people make sense of complex phenomena through data.
- Designed to enable exploration, analysis, or communication.

WHAT MAKES A
VISUALIZATION EFFECTIVE?

Carte Figurative des pertes successives en hommes de l'Armée Française dans la campagne de Russie 1812-1813.

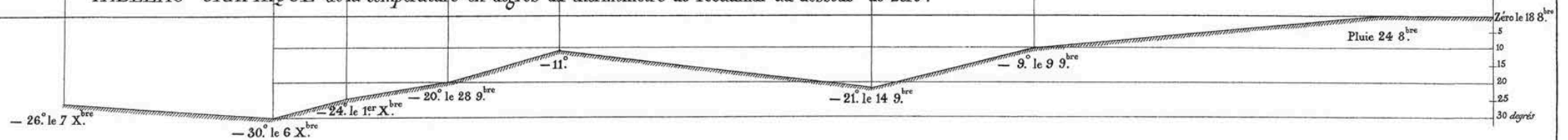
Dressée par M. Minard, Inspecteur Général des Ponts et Chaussées en retraite Paris, le 20 Novembre 1869.

Les nombres d'hommes présents sont représentés par les largeurs des zones colorées à raison d'un millimètre pour dix mille hommes; ils sont de plus écrits en travers des zones. Le rouge désigne les hommes qui entrent en Russie, le noir ceux qui en sortent. — Les renseignements qui ont servi à dresser la carte ont été puisés dans les ouvrages de M. M. Chiers, de Ségur, de Fezensac, de Chambray et le journal inédit de Jacob, pharmacien de l'Armée depuis le 28 Octobre. Pour mieux faire juger à l'œil la diminution de l'armée, j'ai supposé que les corps du Prince Jérôme et du Maréchal Davoust qui avaient été détachés sur Minsk et Mohilow et ont rejoint vers Orscha et Witebsk, avaient toujours marché avec l'armée.



TABEAU GRAPHIQUE de la température en degrés du thermomètre de Réaumur au dessous de zéro.

Les Cosaques passent au galop le Niemen gelé.



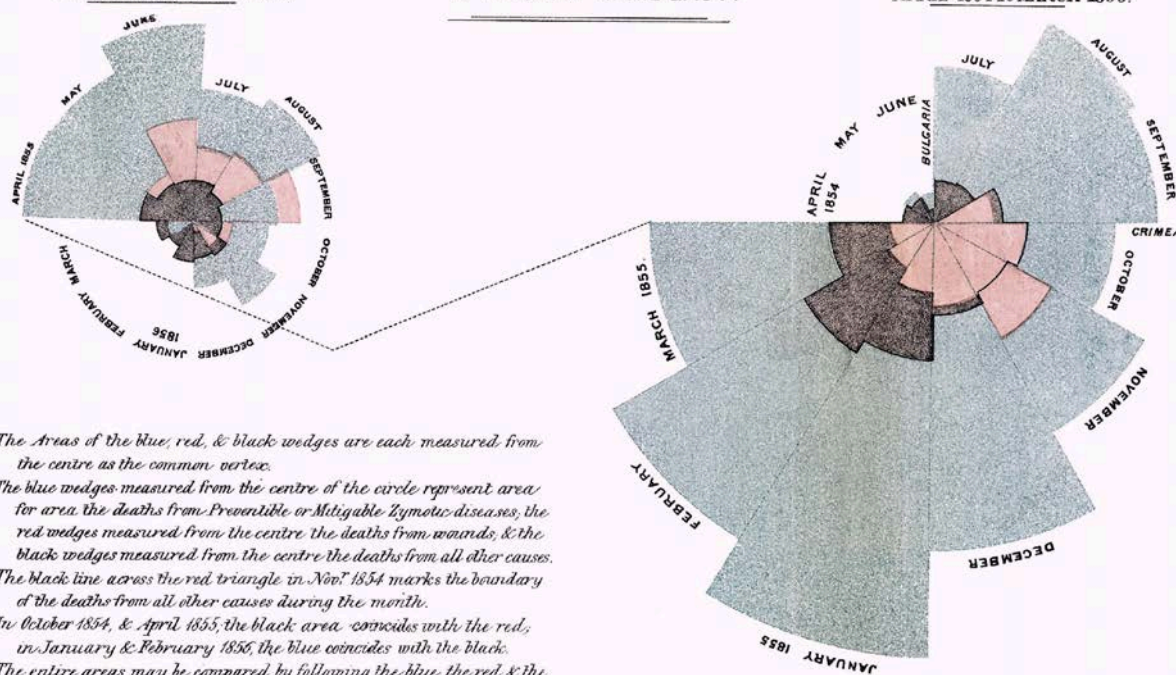
Autog. par Regnier, 8. Pas. 5^{me} Marie St G^{de} à Paris.

Imp. Lit. Regnier et Dourdet.

DIAGRAM OF THE CAUSES OF MORTALITY IN THE ARMY IN THE EAST.

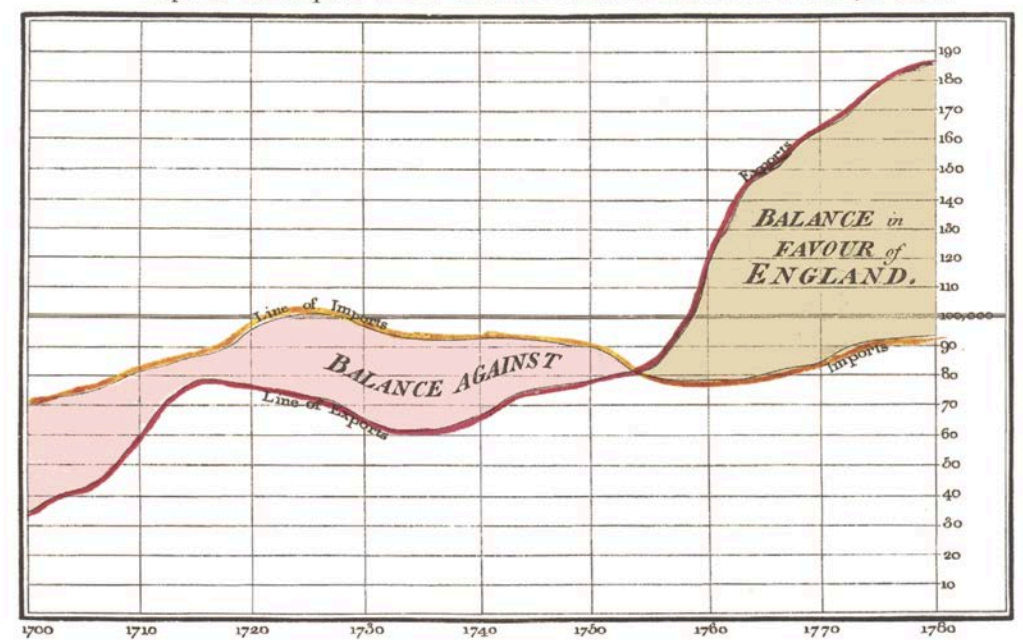
2. APRIL 1855 TO MARCH 1856.

1. APRIL 1854 TO MARCH 1855.



The Areas of the blue, red, & black wedges are each measured from the centre as the common vertex.
The blue wedges measured from the centre of the circle represent area for area the deaths from Preventable or Mitigable Zymotic diseases, the red wedges measured from the centre the deaths from wounds, & the black wedges measured from the centre the deaths from all other causes.
The black line across the red triangle in Nov. 1854 marks the boundary of the deaths from all other causes during the month.
In October 1854, & April 1855, the black area coincides with the red, in January & February 1855, the blue coincides with the black.
The entire areas may be compared by following the blue, the red & the black lines enclosing them.

Exports and Imports to and from DENMARK & NORWAY from 1700 to 1780.



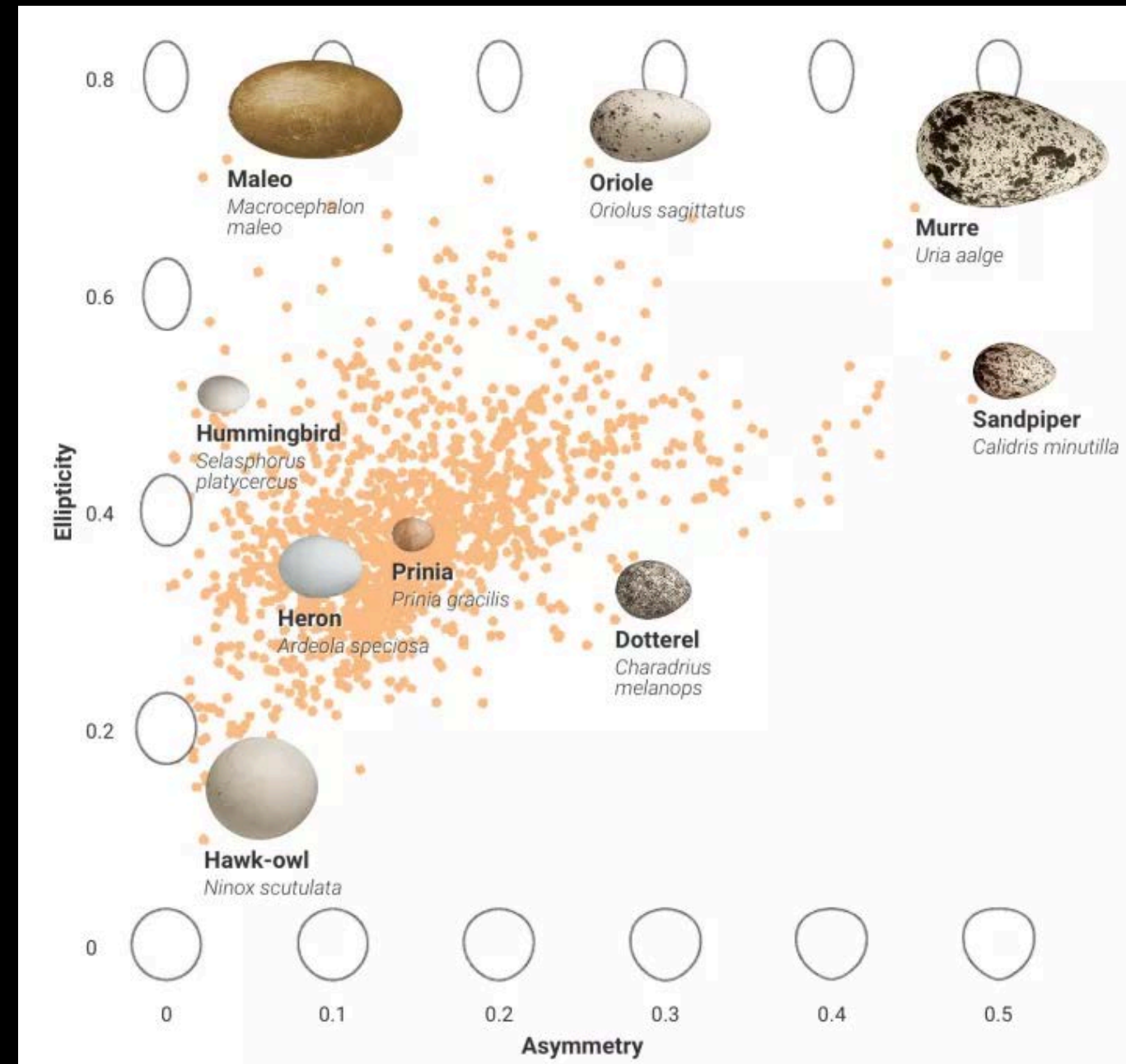
The Bottom line is divided into Years, the Right hand line into L10,000 each.
Published as the Act directs, 1st May 1786, by W. Playfair.
Noble sculpt 352 Strand, London.

WHAT MAKES A VISUALIZATION MEMORABLE?

- *Great data visualization tells a great story*
 - Convince us of something, compel us to action, enlighten us with new information, or force us to question our own preconceptions
- Great data visualization reaches people in a way that words or numbers can't

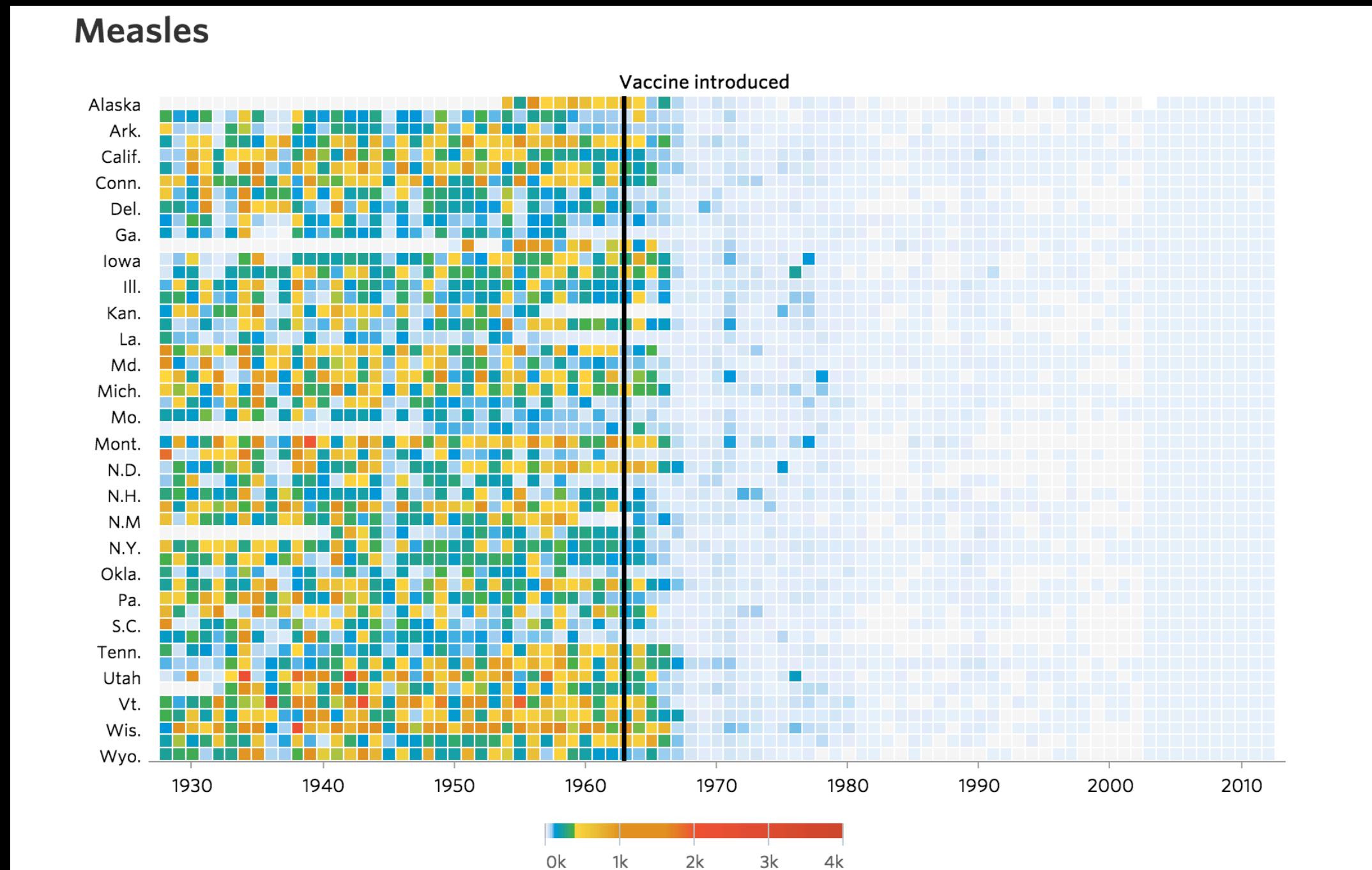
WHY DISPLAY DATA?

1. Initial examination and exploratory analysis
 - Check data (distribution, outliers, impossible values)
 - See data in context
 - Find patterns
 - Discover new questions



WHY DISPLAY DATA?

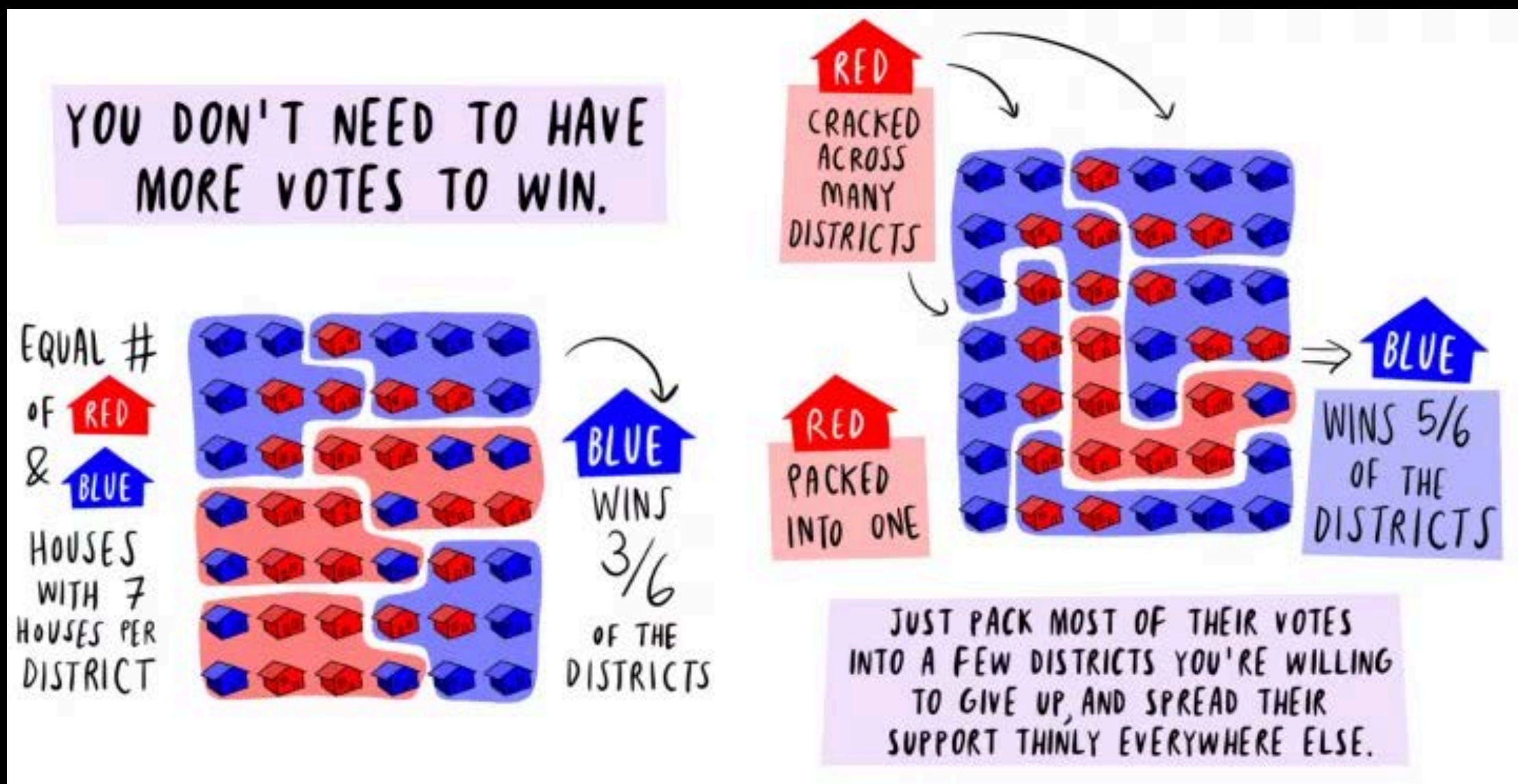
2. Answer questions & make decisions



WHY DISPLAY DATA?

3. Communicate meaning to others

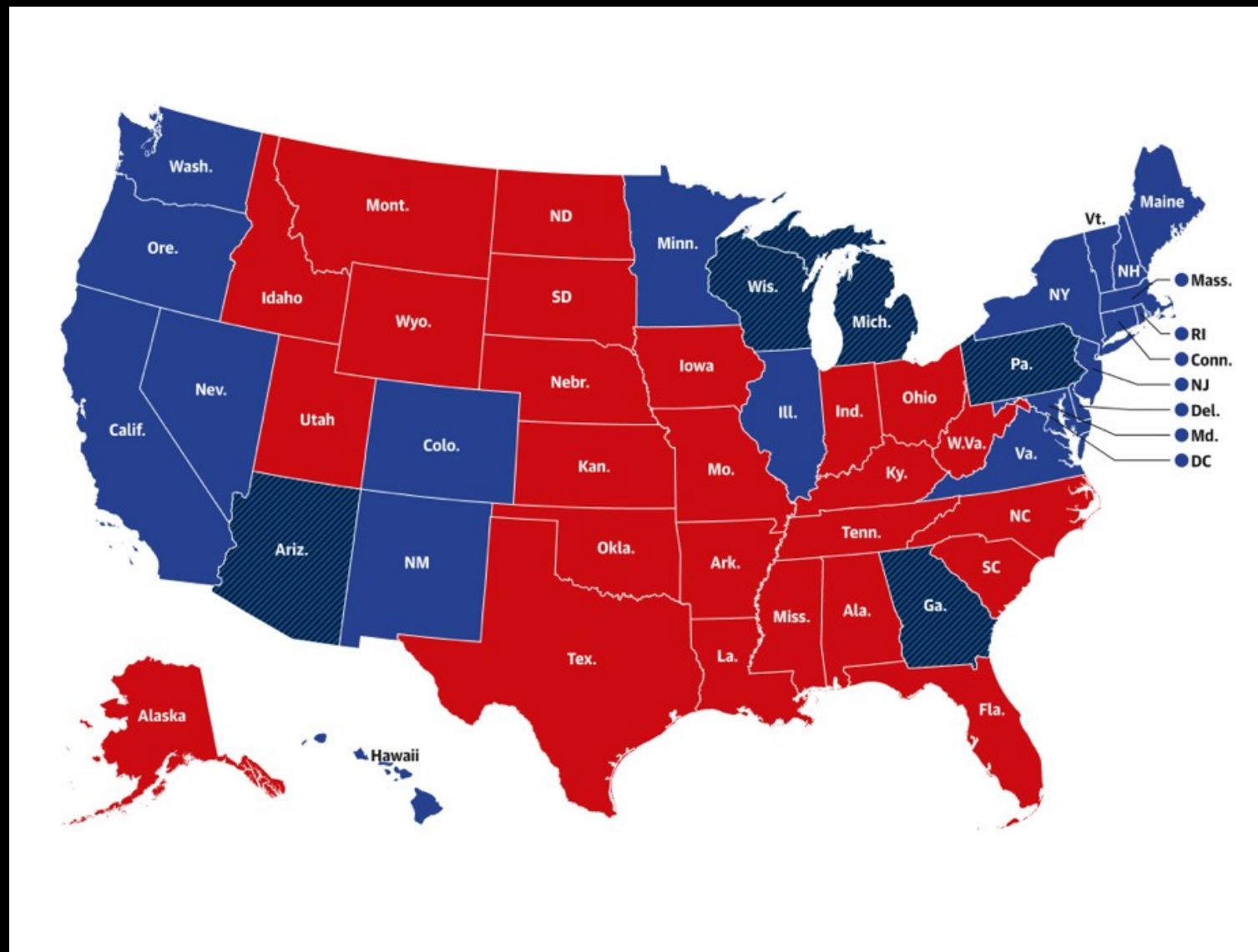
- Present argument, tell a story, inspire, archive

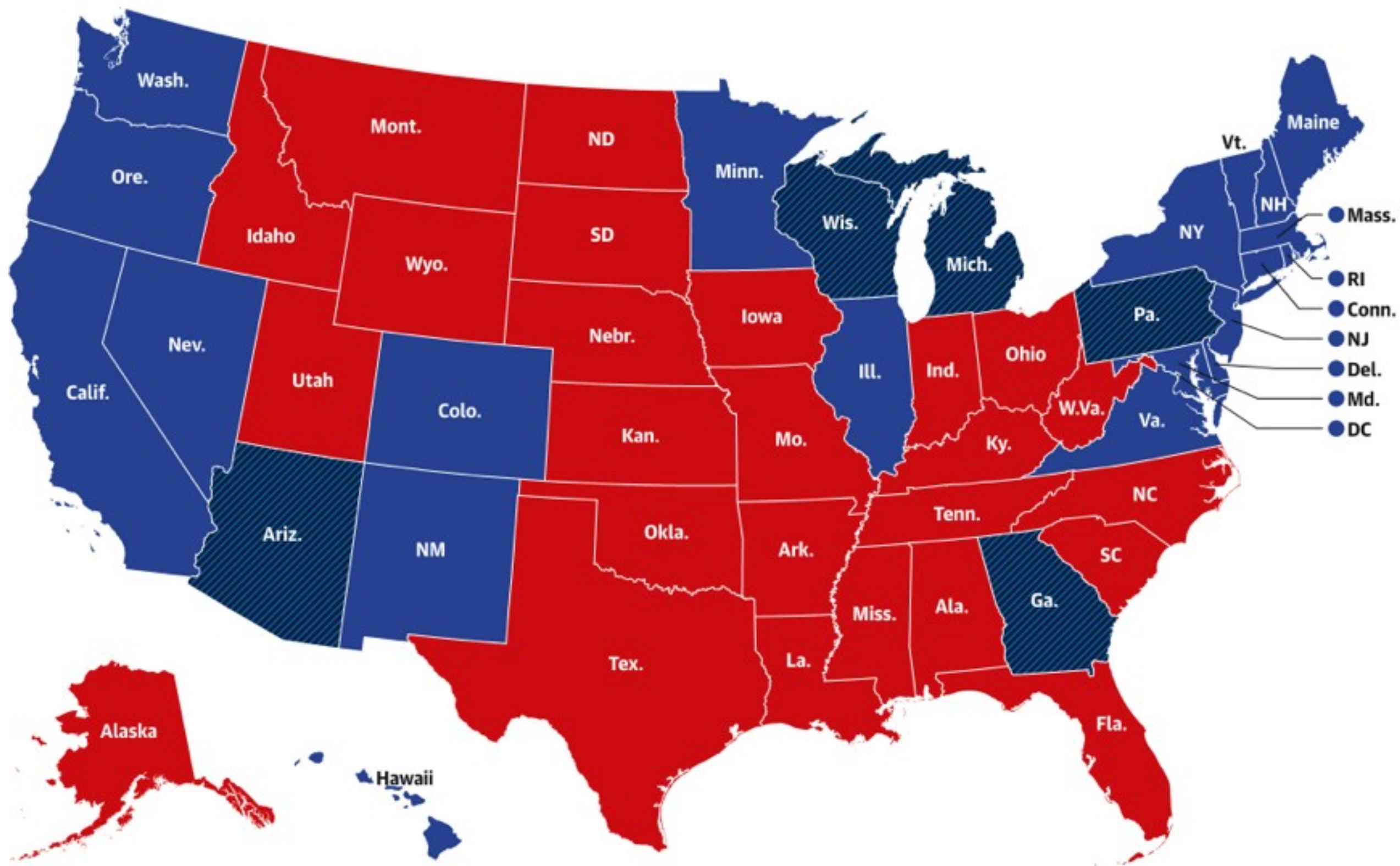


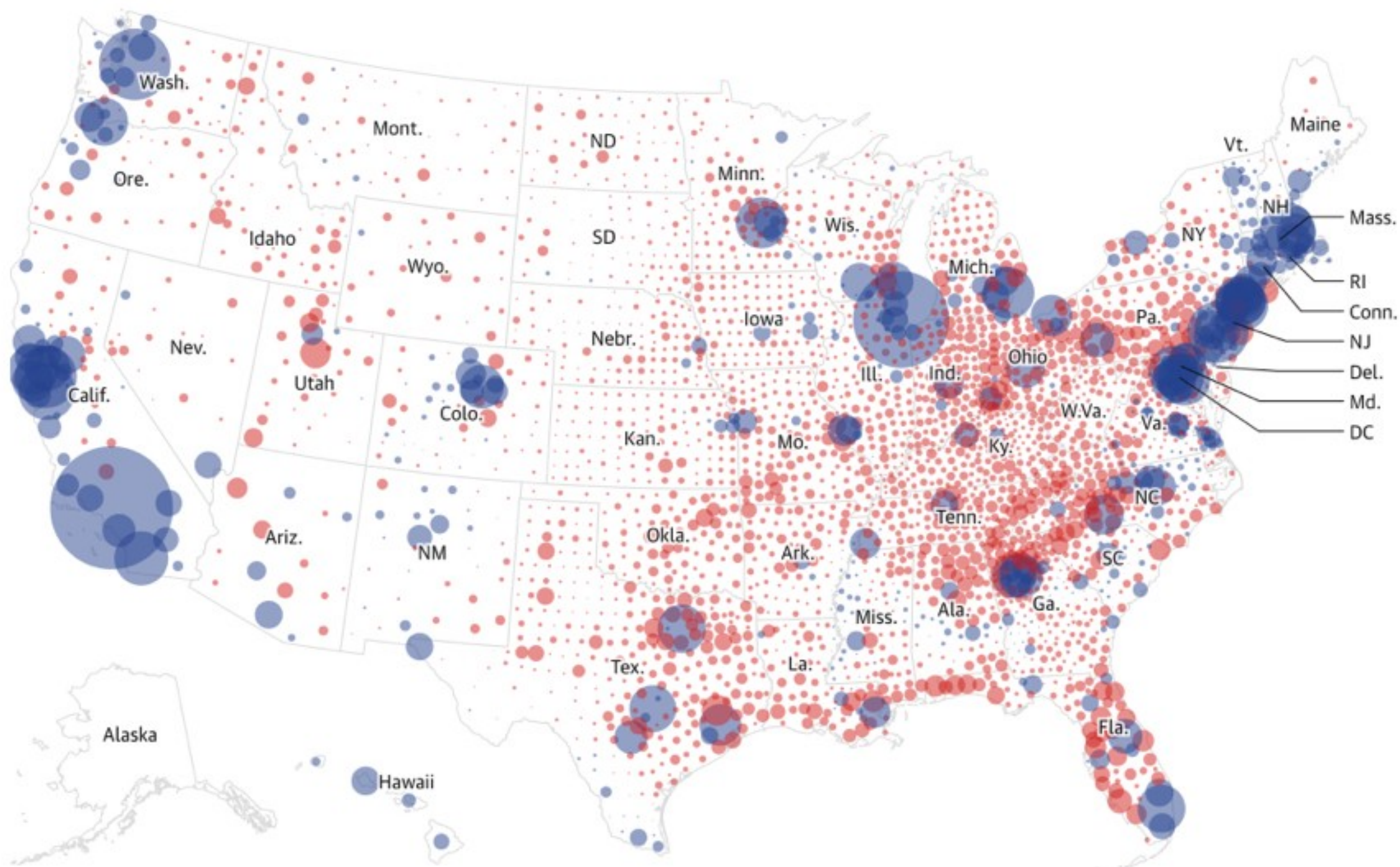
WHY DISPLAY DATA?

4. Improve understanding

- Represent something more clearly, honestly, etc.







2020 ELECTION MAP

👤 = 250,000 VOTES

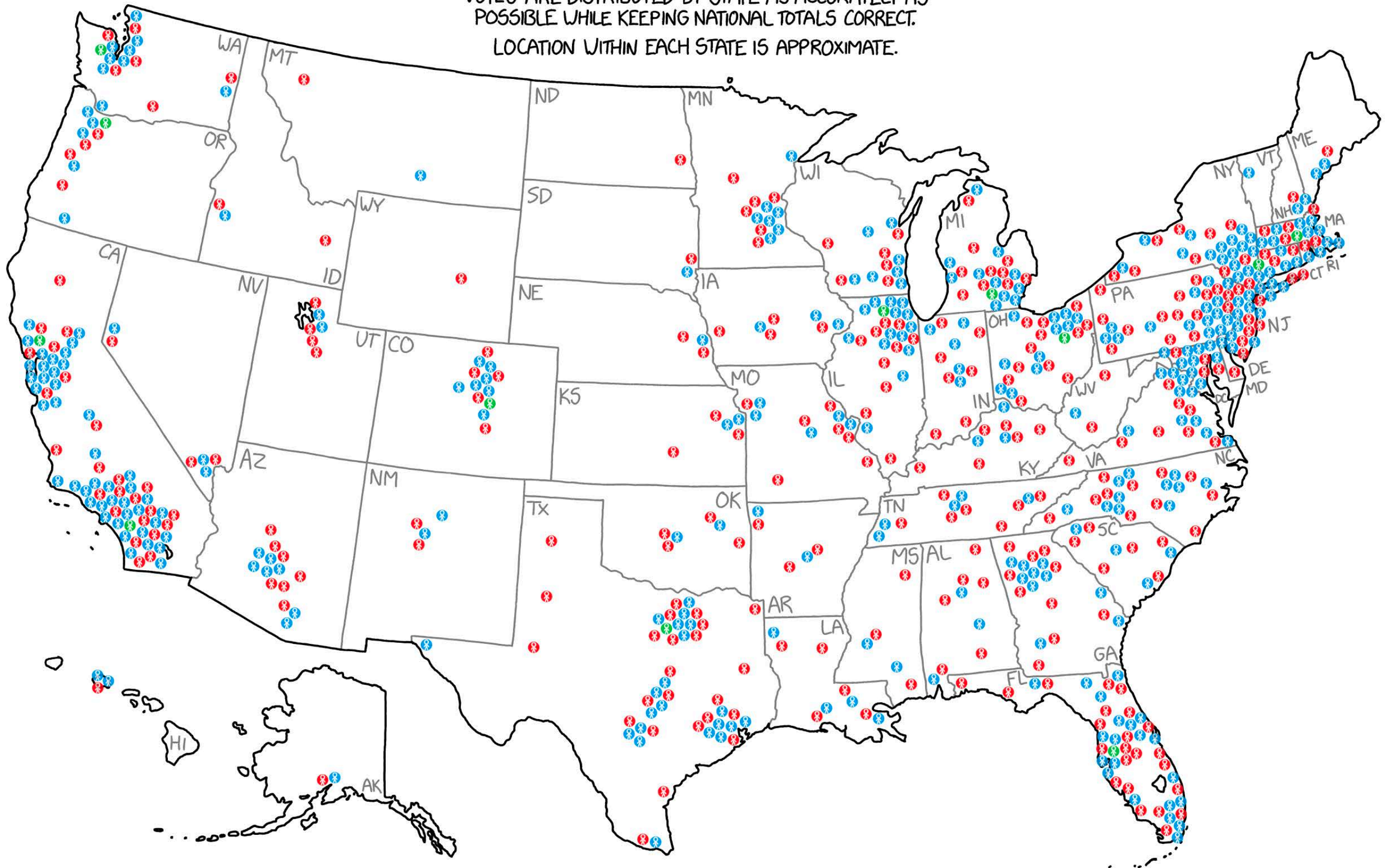
👤 BIDEN

👤 TRUMP

👤 OTHER

VOTES ARE DISTRIBUTED BY STATE AS ACCURATELY AS POSSIBLE WHILE KEEPING NATIONAL TOTALS CORRECT.

LOCATION WITHIN EACH STATE IS APPROXIMATE.



WHY WE VISUALIZE DATA

WHY WE VISUALIZE DATA

- Information shaped as a graphic functions as a **cognitive aid**
- You can take something extremely complex, or something concealed in the data, and make it understandable and salient

CORE PRINCIPLES OF GOOD GRAPHICS

- They are **honest** representations of the data
- They show the right amount of data
- They attract readers' attention
- They don't frustrate readers

Some degree of
subjectivity &
judgement

Fig. 1.1 in Claus Wilke's *Data Visualization*

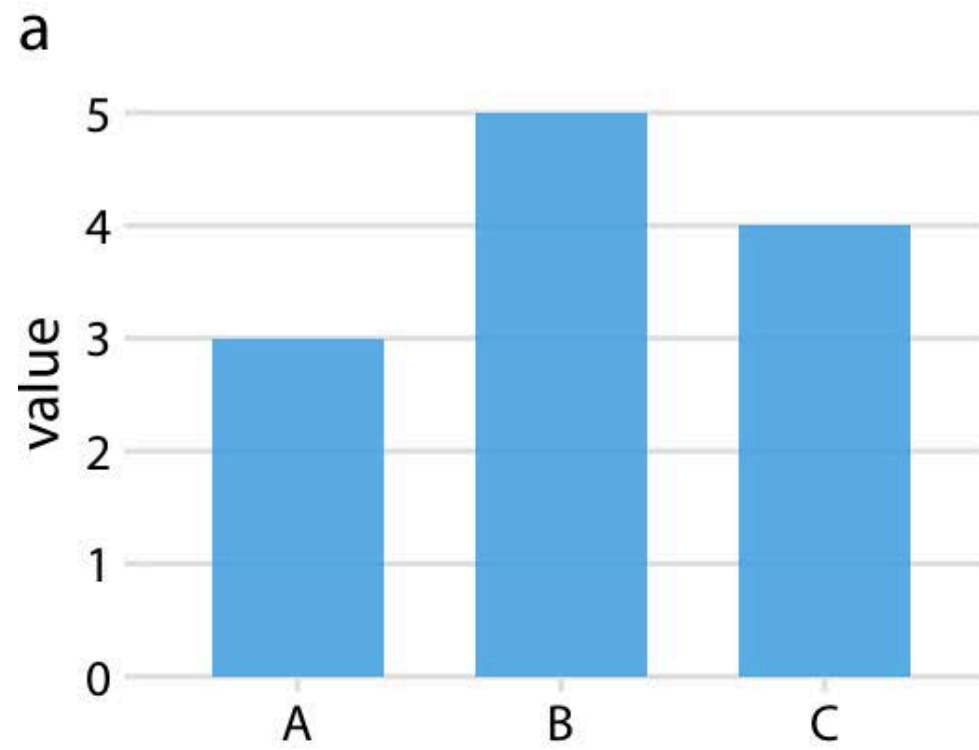
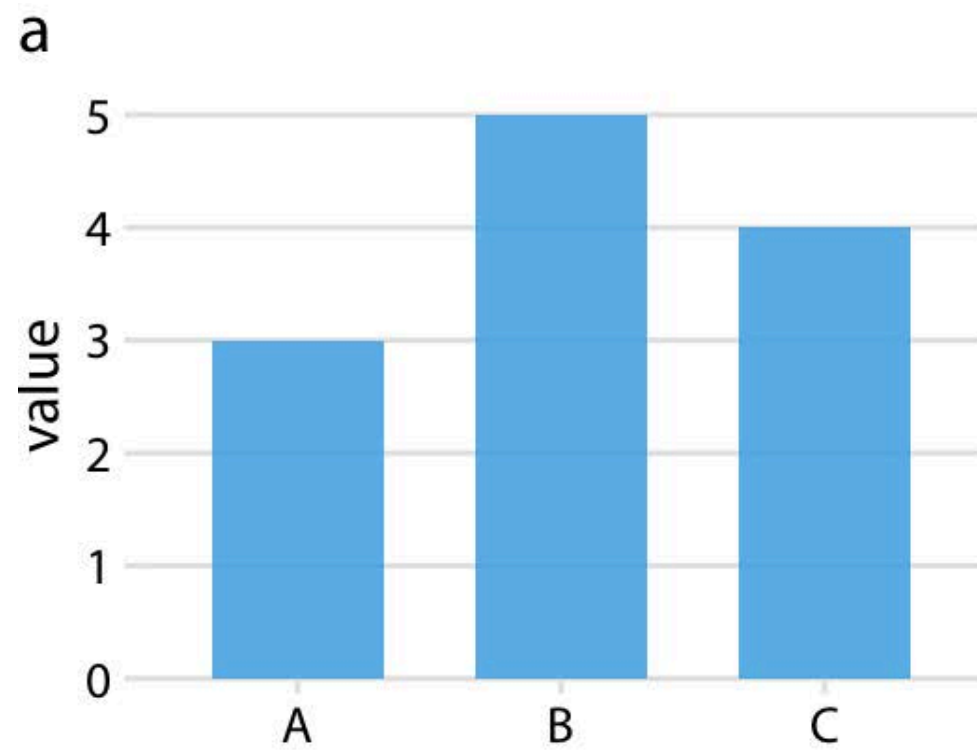


Fig. 1.1 in Claus Wilke's *Data Visualization*



Aesthetic problems or "bad taste"

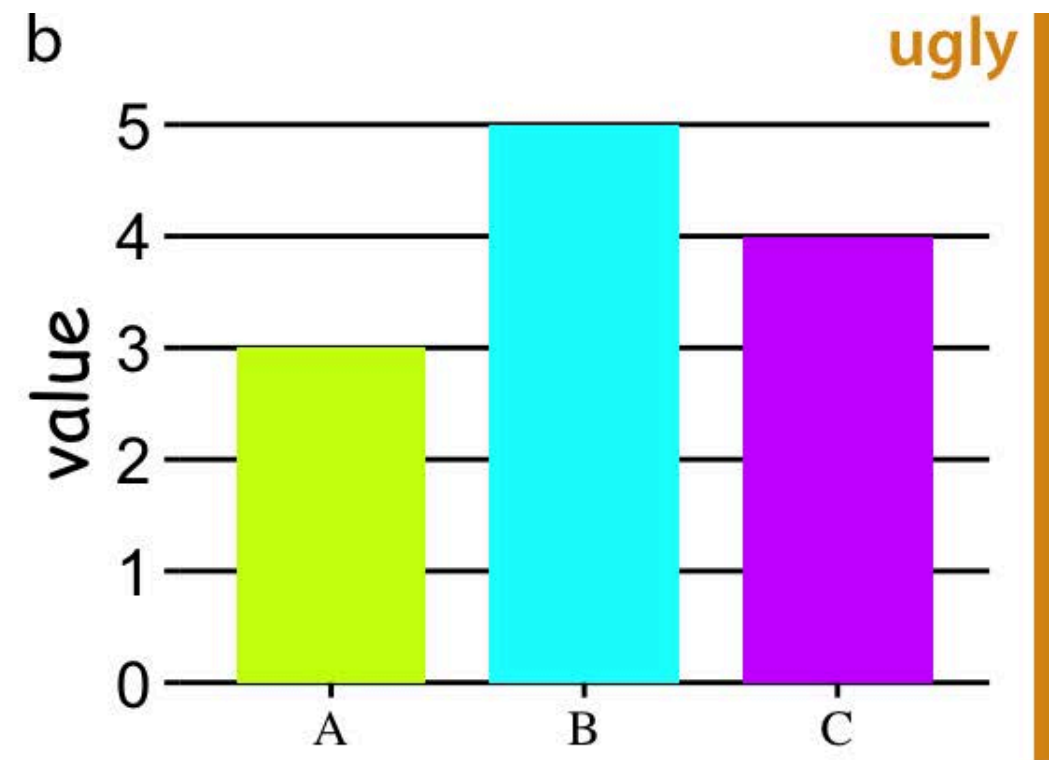
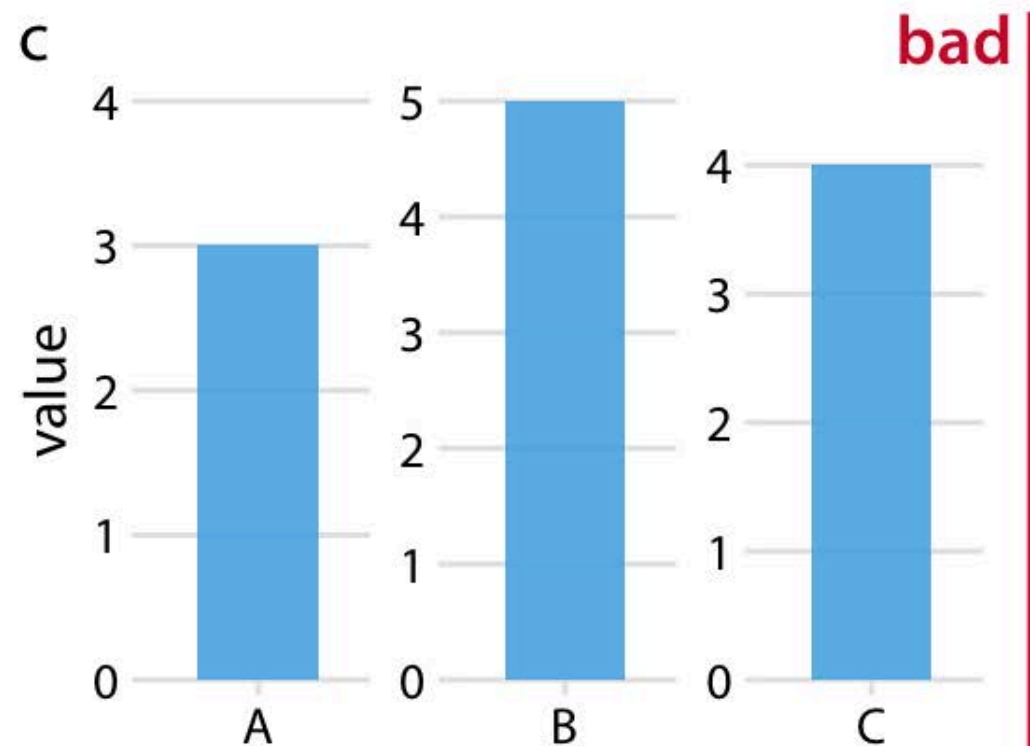
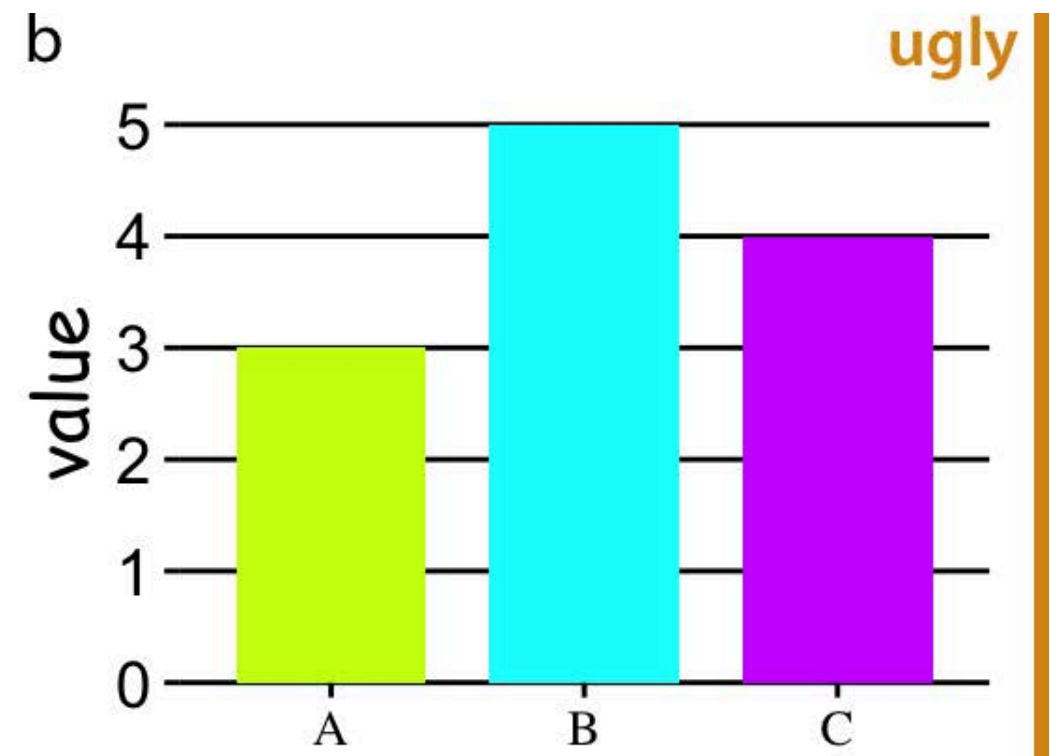
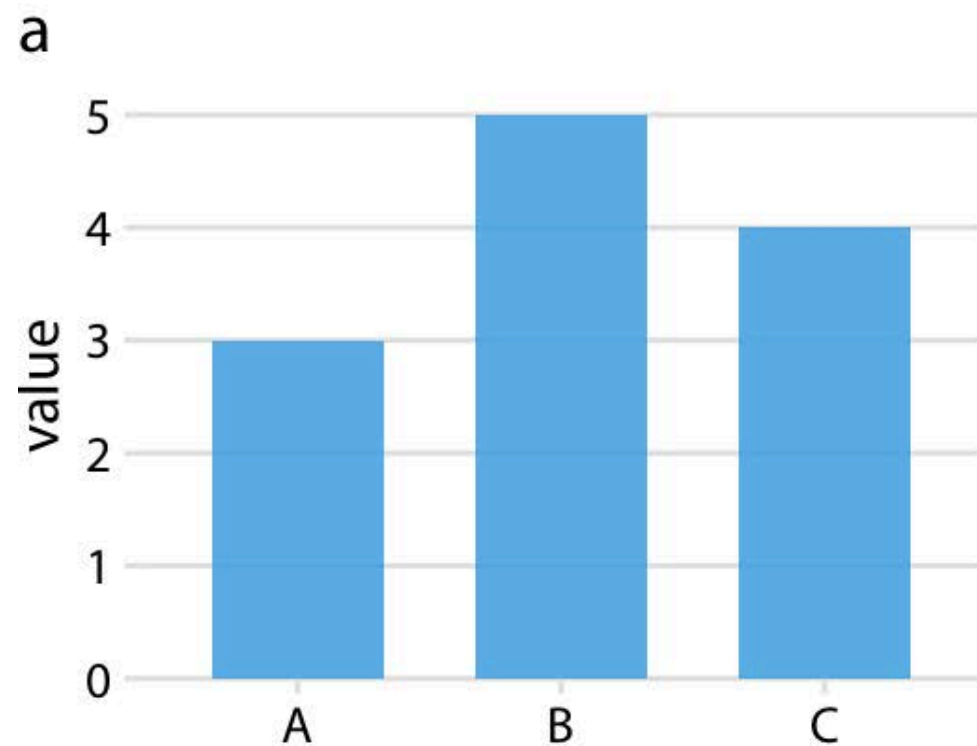


Fig. 1.1 in Claus Wilke's *Data Visualization*

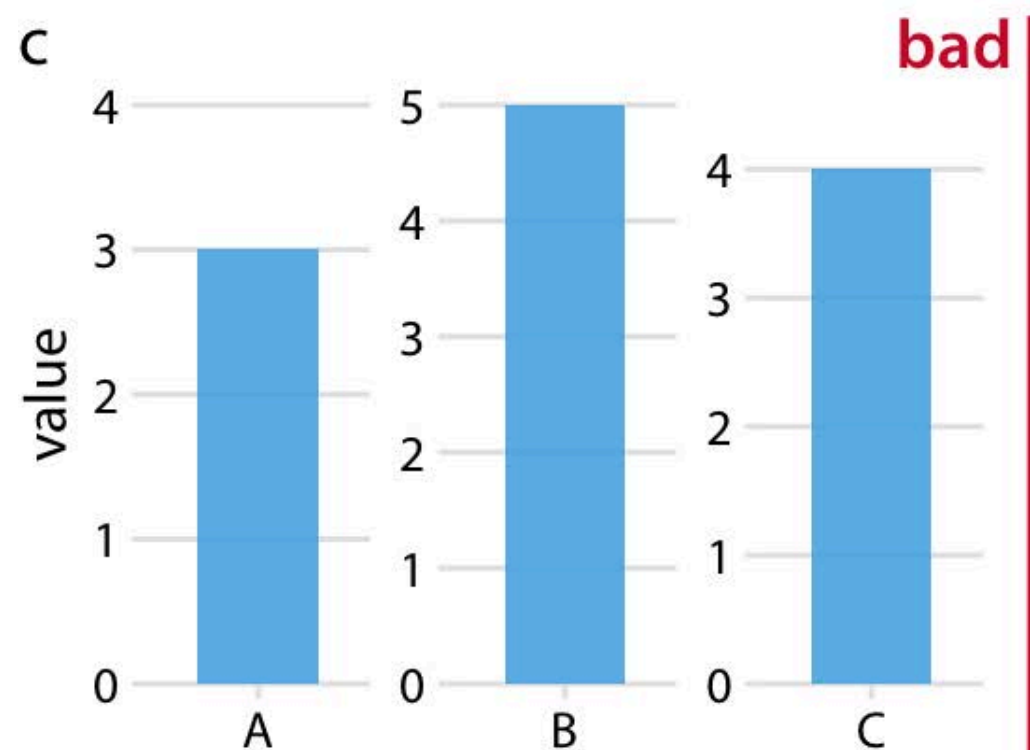
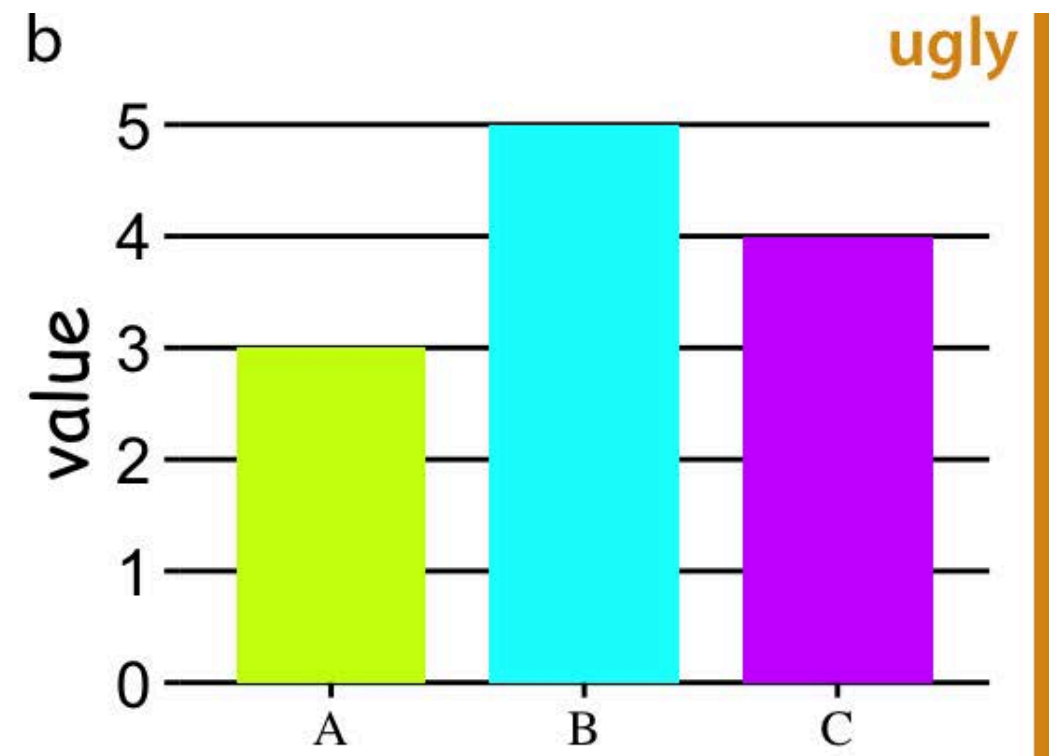
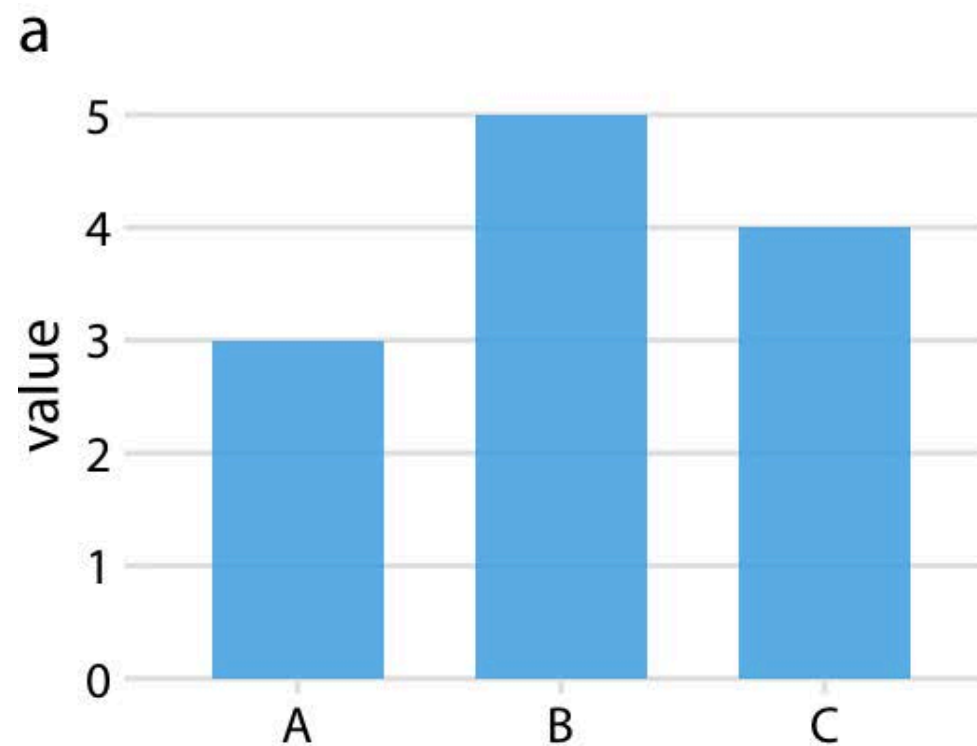
Aesthetic problems or "bad taste"



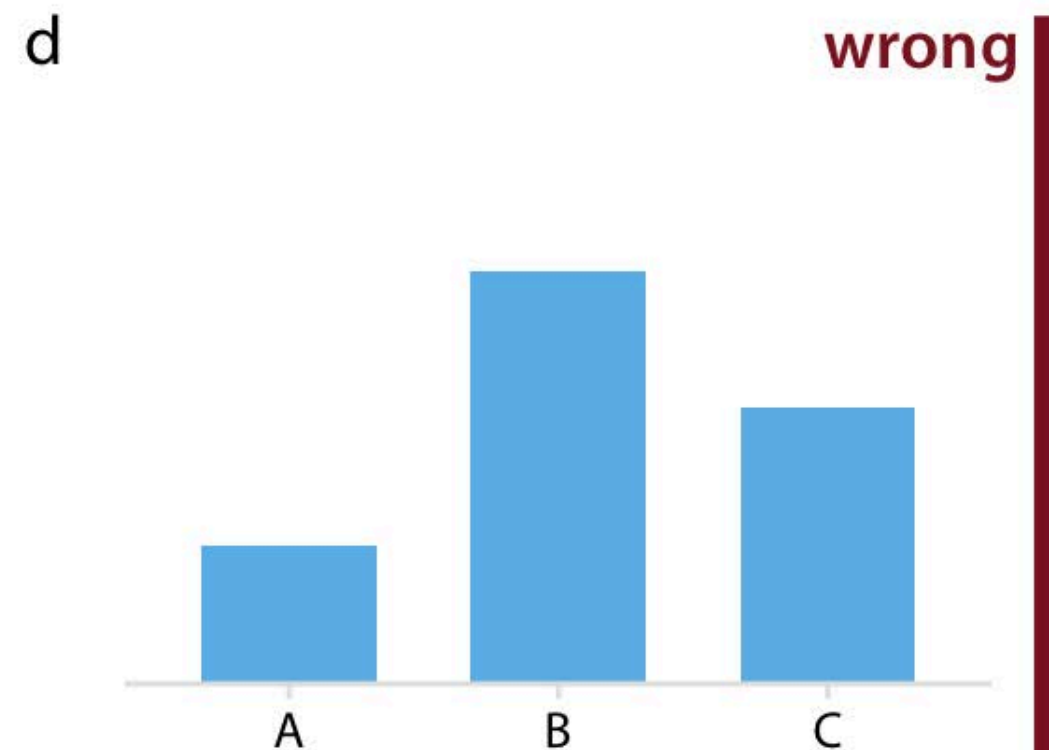
Perceptual problems

Fig. 1.1 in Claus Wilke's *Data Visualization*

Aesthetic problems or "bad taste"



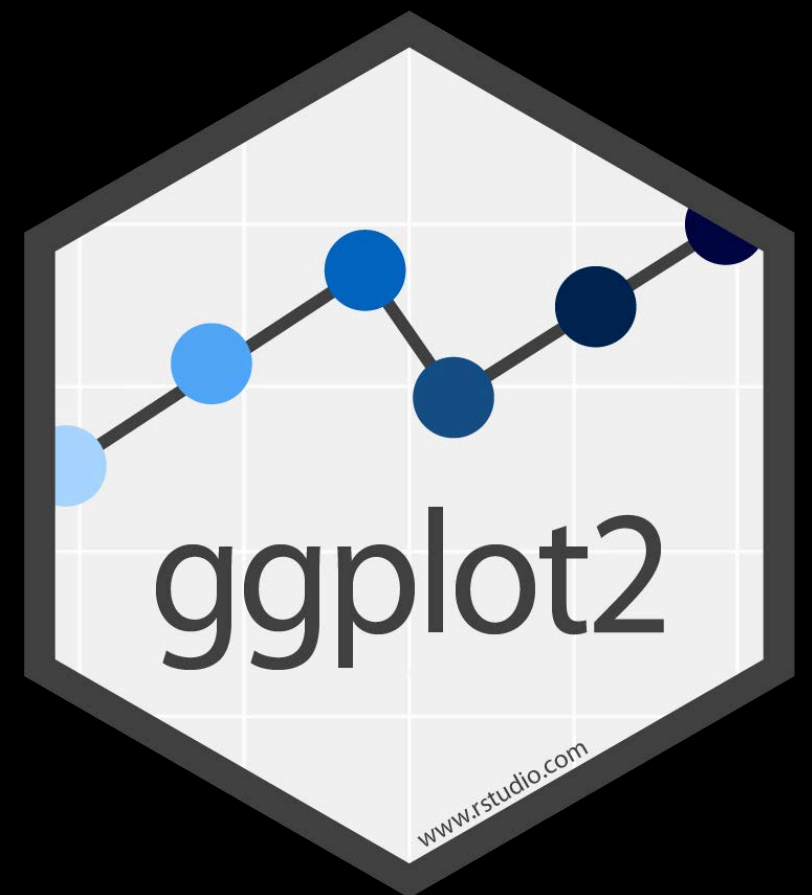
Perceptual problems



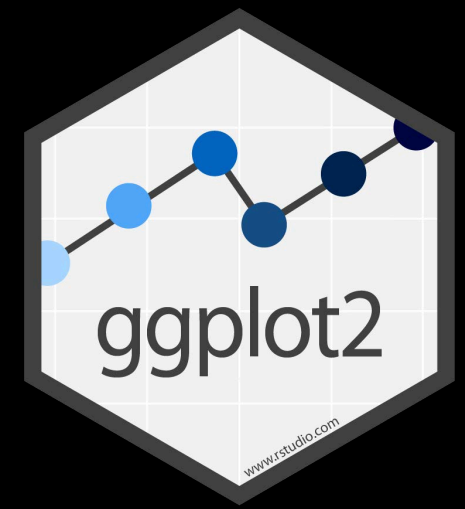
Numerical problems or "bad data"

GGPLOT2

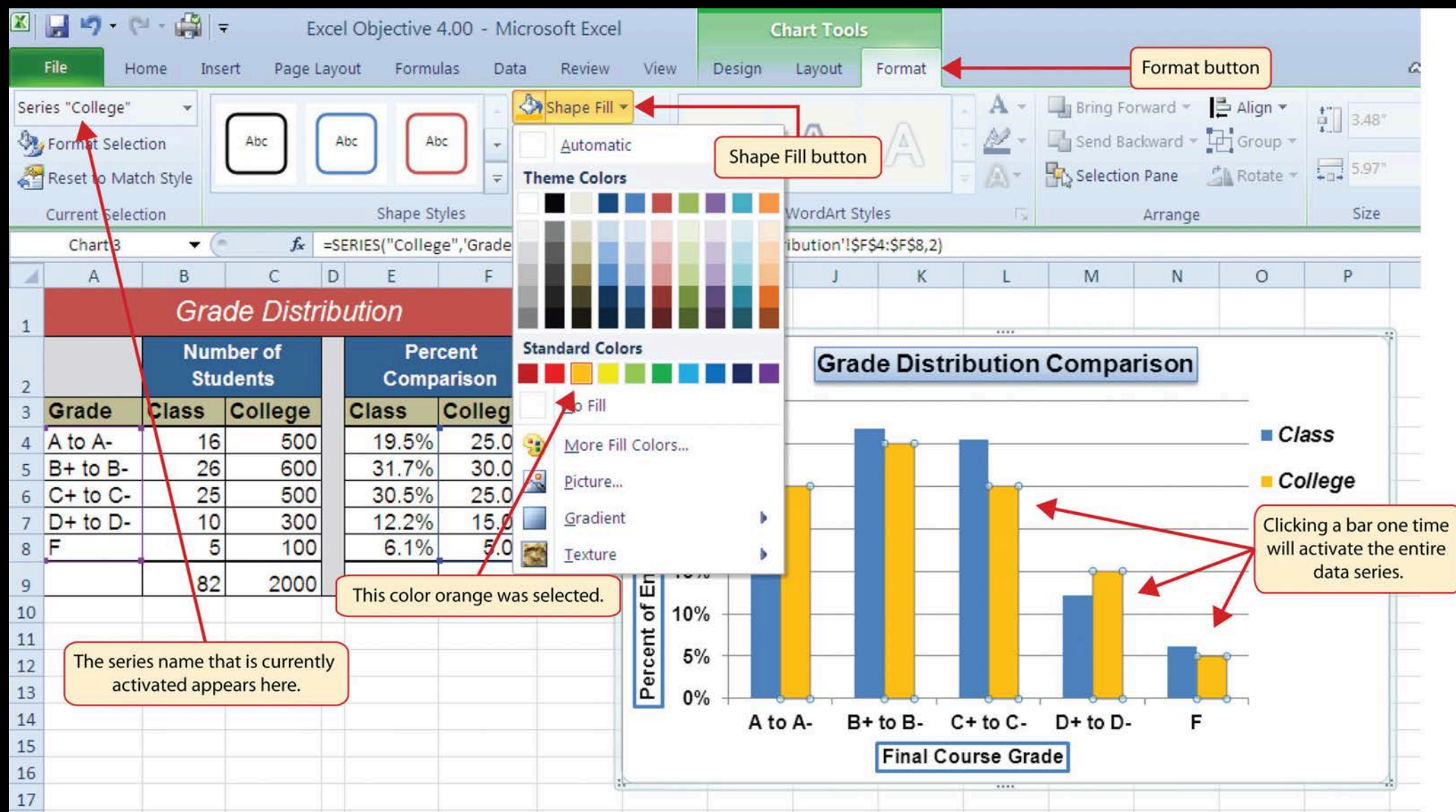
A GRAMMAR OF GRAPHICS



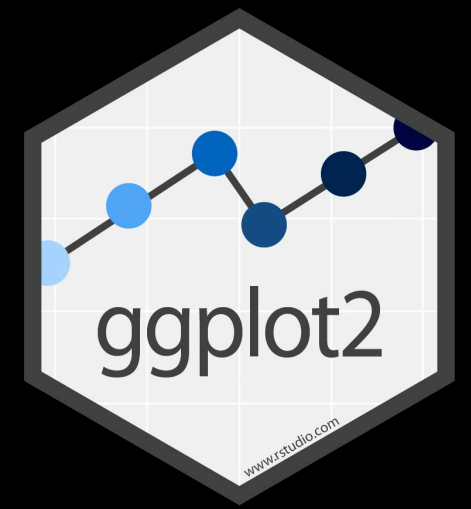
GRAMMAR OF GRAPHICS



- Traditional plotting: you *are* a painter
- Manually place/modify graphical elements

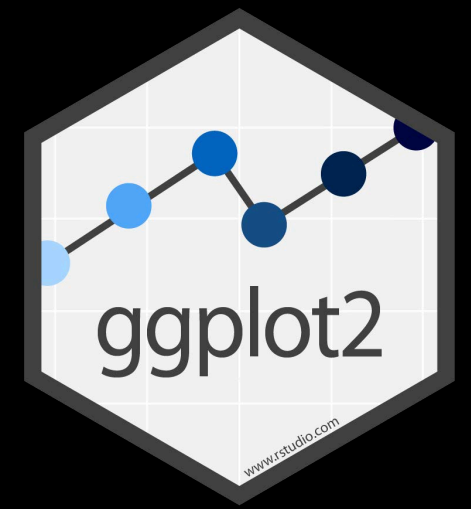


GRAMMAR OF GRAPHICS



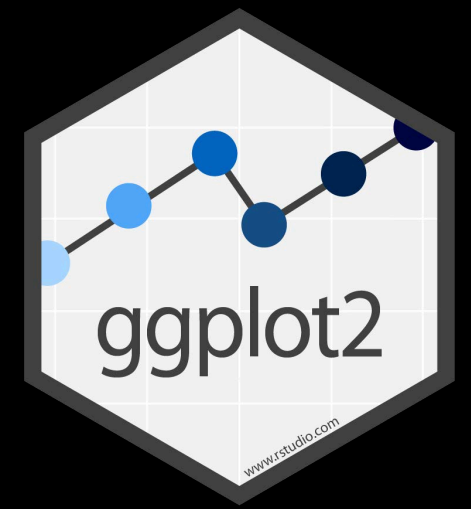
- Traditional plotting: you *are* a painter
 - Manually place/modify graphical elements

GRAMMAR OF GRAPHICS

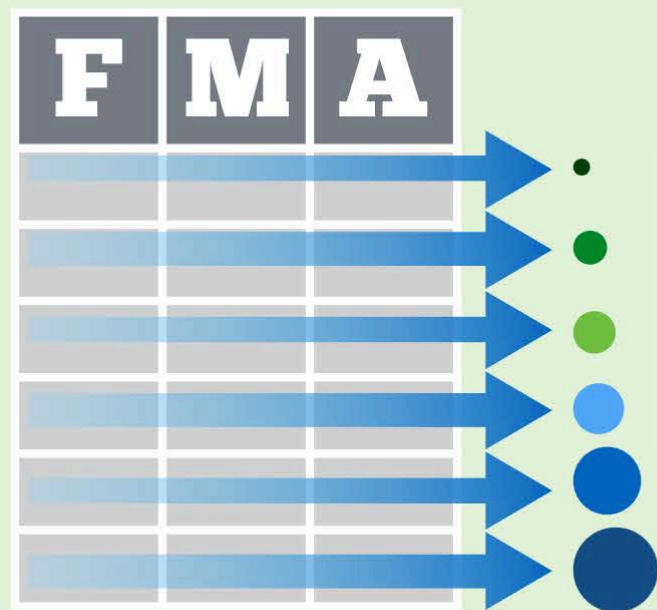


- ggplot2: you *employ* a painter
 - Describe conceptually how to visualize the data

GRAMMAR OF GRAPHICS



- ggplot2: you *employ* a painter
- Describe conceptually how to visualize the data



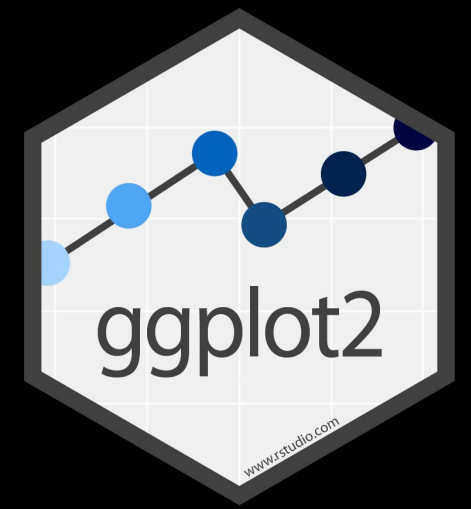
data

geom

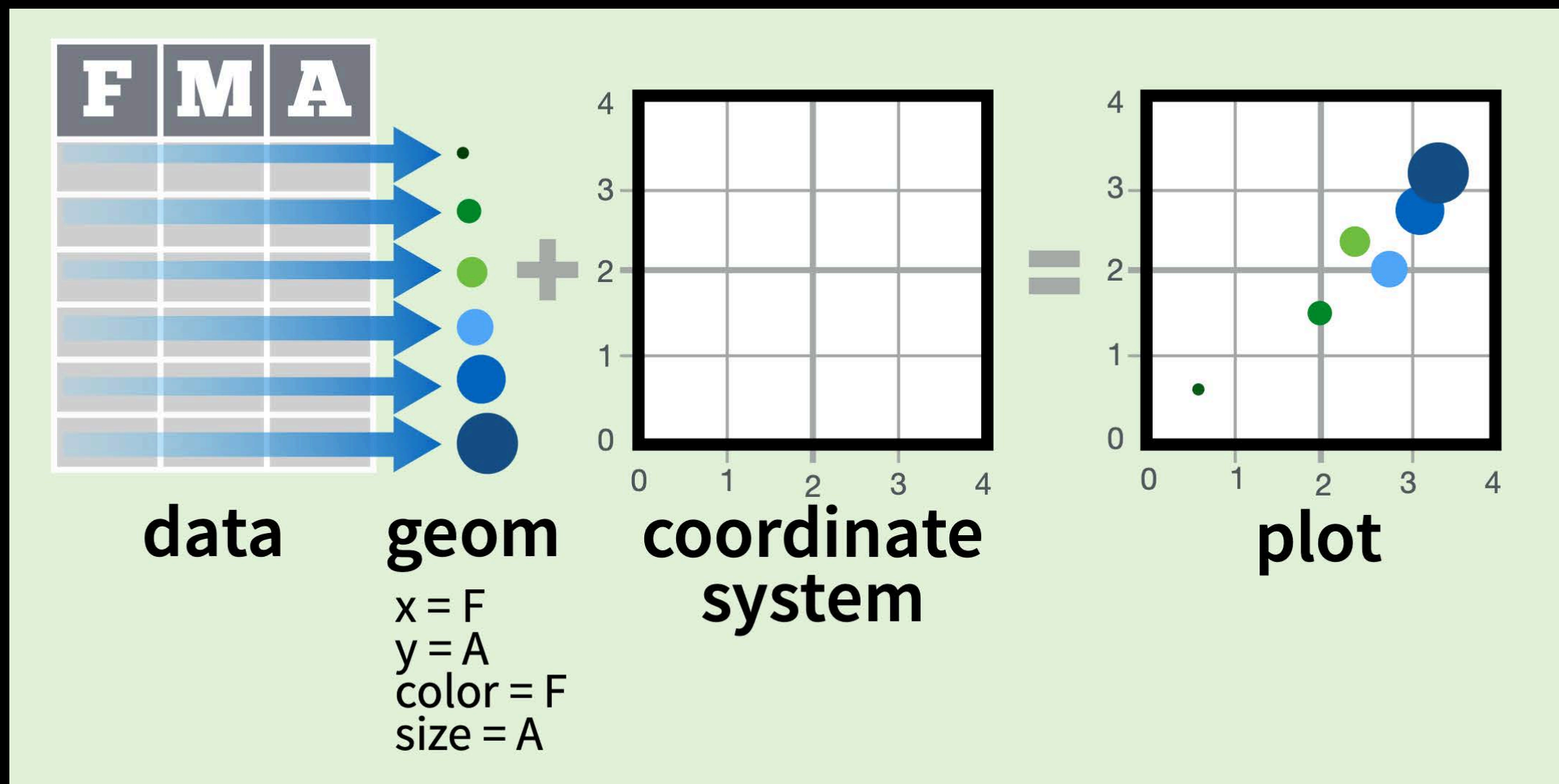
x = F
y = A
color = F
size = A

Data values *mapped* to
visual elements of the plot

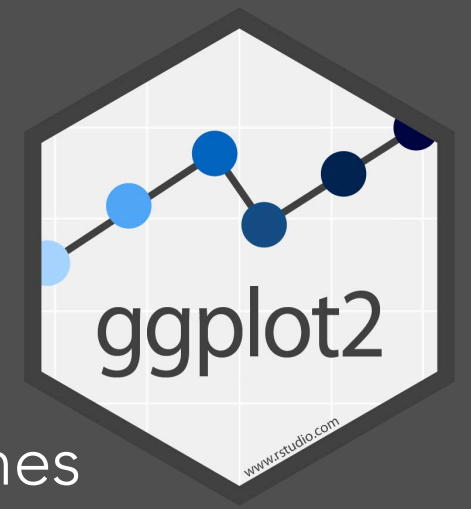
GRAMMAR OF GRAPHICS



- ggplot2: you *employ* a painter
- Describe conceptually how to visualize the data



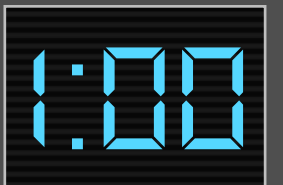
YOUR TURN



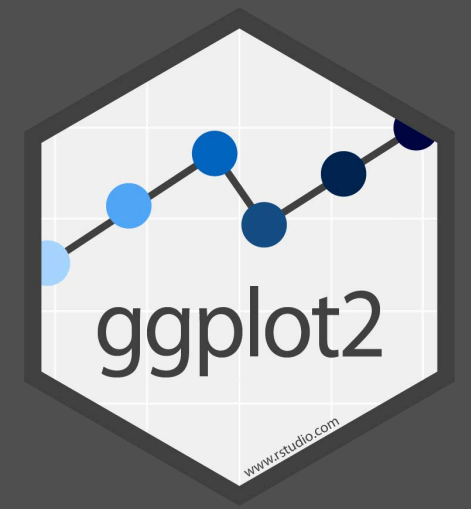
- Let's create a plot in ggplot2 using the `storms` dataset that comes with `tidyverse`

name	year	month	day	hour	lat	long	status	category	wind	pressure	ts_diameter	hu_diameter
Amy	1975	6	27	0	27.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	6	28.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	12	29.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	18	30.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	28	0	31.5	-78.8	tropical depression	-1	25	1012	NA	NA
Amy	1975	6	28	6	32.4	-78.7	tropical depression	-1	25	1012	NA	NA

- First, think about the variables:
 - What relationship do you expect between wind speed and pressure?
 - Is storm status determined by wind speed or air pressure? (no Googling!)



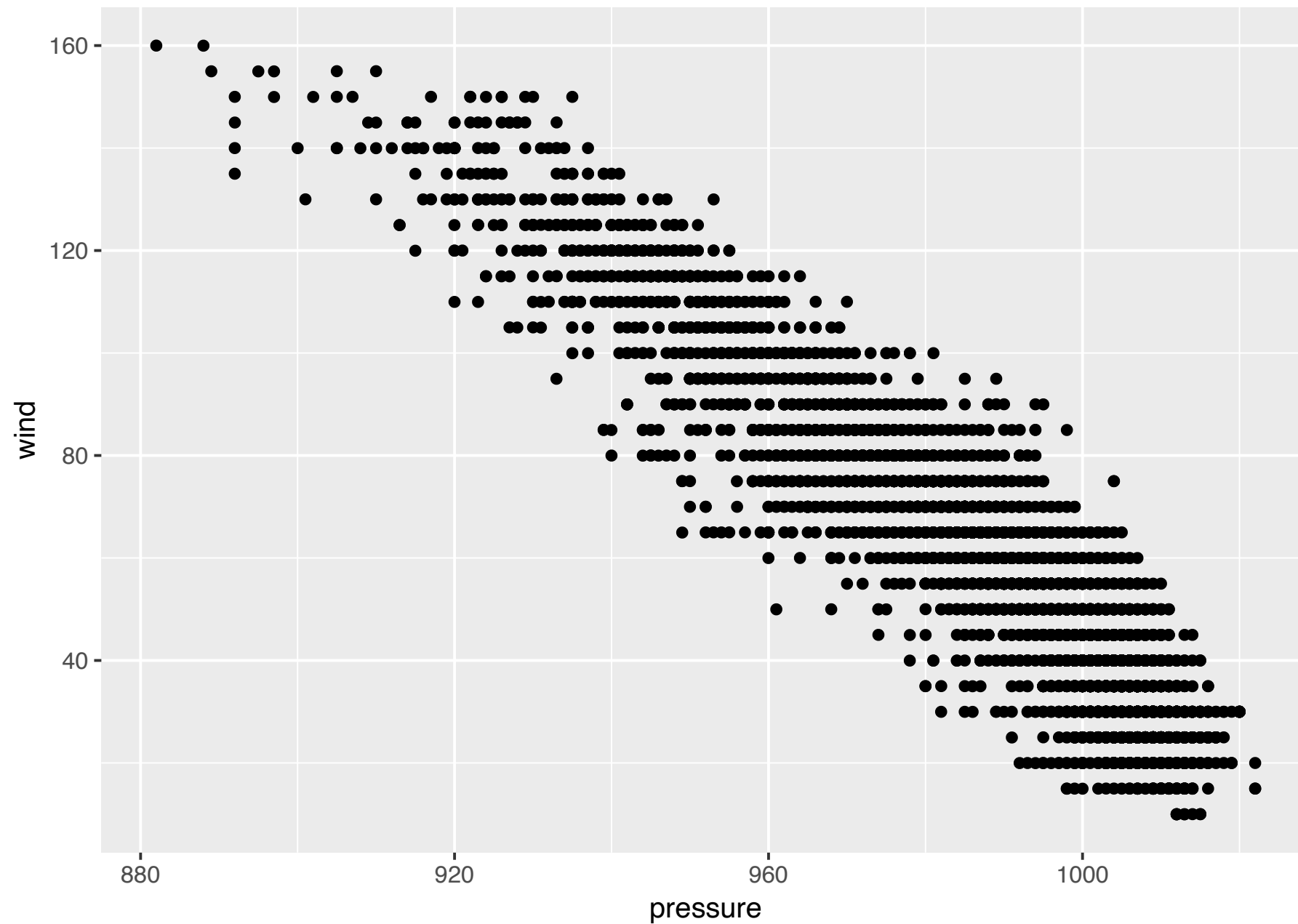
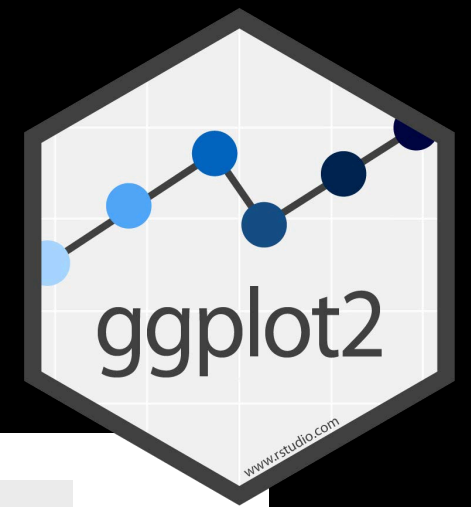
YOUR TURN



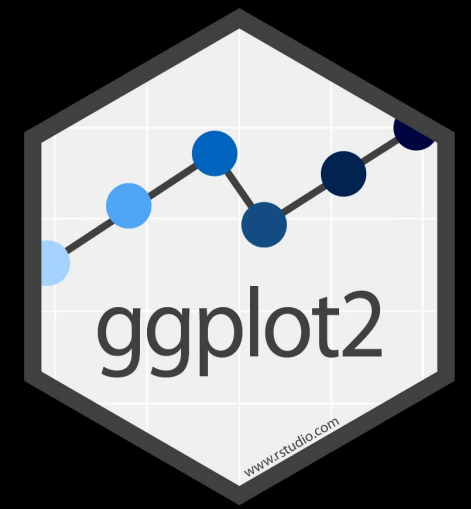
- Create a new R markdown and clear out the extra stuff
- Create a new R chunk and load the **tidyverse** package.
- Run this code to make a graph. Pay strict attention to spelling, capitalization, and parentheses!

```
ggplot(data = storms) +  
  geom_point(mapping = aes(x = pressure, y = wind))
```

1:00



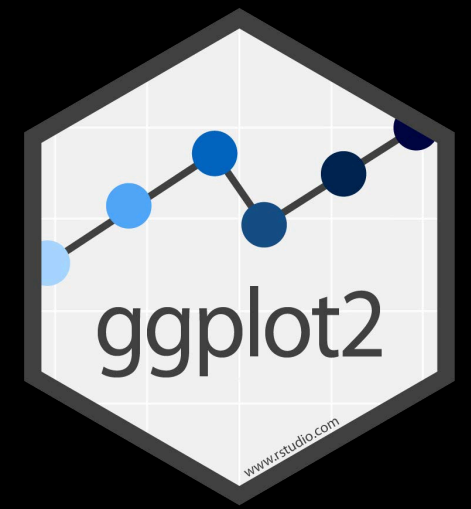
```
ggplot(data = storms) +  
  geom_point(mapping = aes(x = pressure, y = wind))
```



1. "Initialize" a plot with `ggplot()`
2. Add layers with `geom_` functions

Important tip: always put the + at the end of a line, never at the start

```
ggplot(data = storms) +  
  geom_point(mapping = aes(x = pressure, y = wind))
```



initialize

data

+ before new line

```
ggplot(data = storms) +  
  geom_point(mapping = aes(x = pressure, y = wind))
```

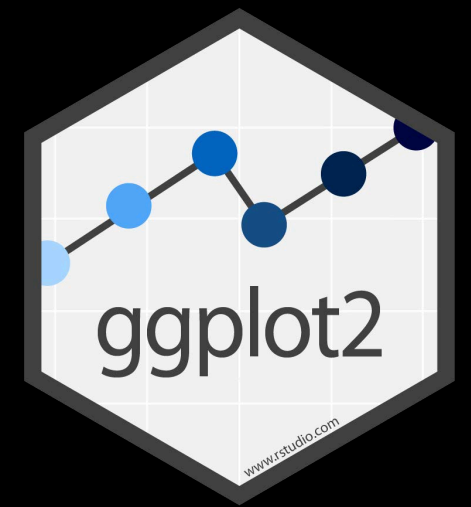
type of layer

aes()

x variable

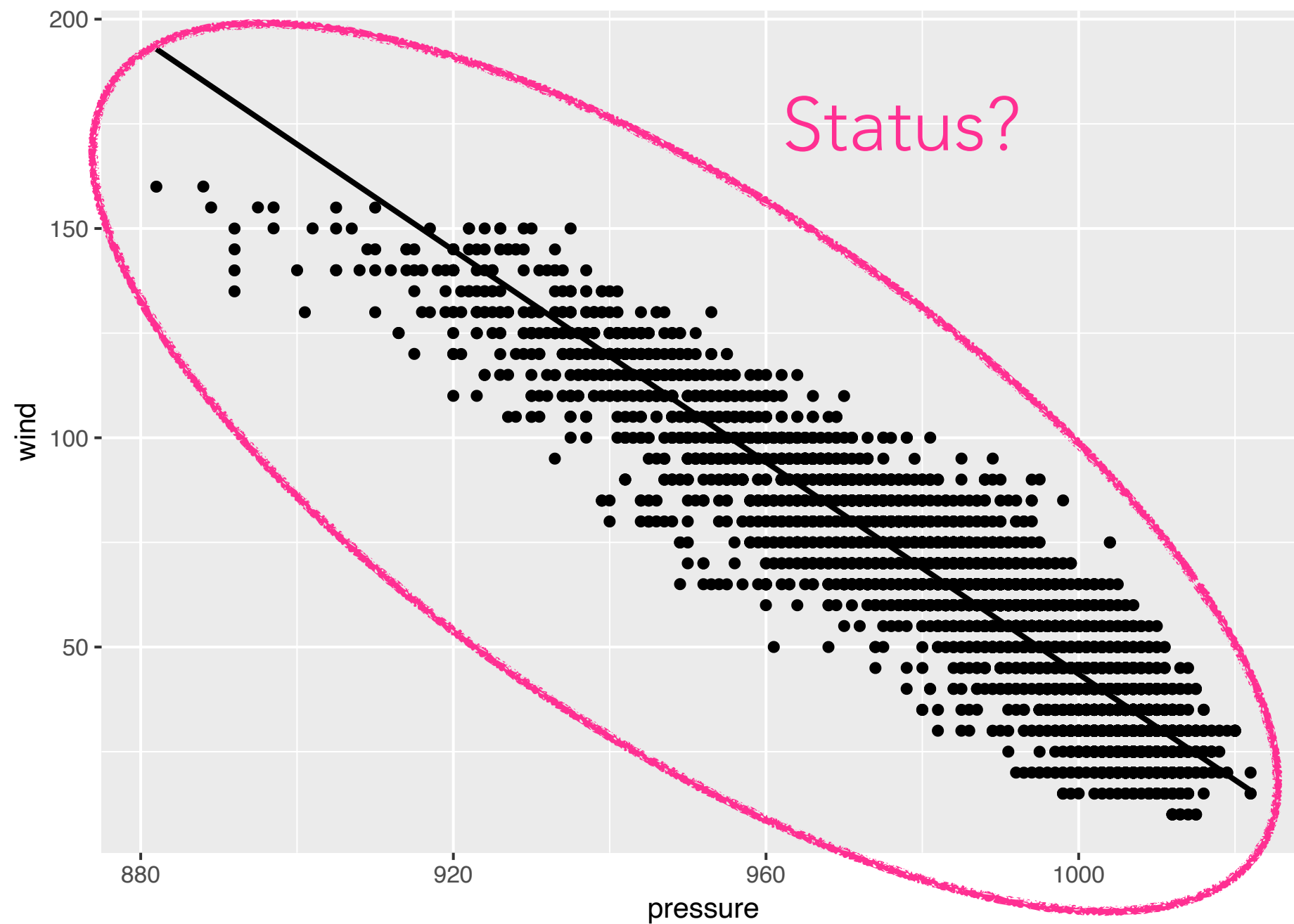
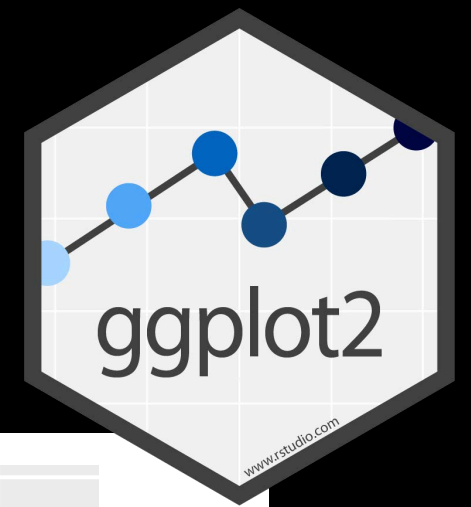
y variable

A TEMPLATE



```
ggplot(data = <DATA>) +  
  <GEOM_FUNCTION>(mapping = aes(<MAPPINGS>))
```

```
ggplot(data = storms) +  
  geom_point(mapping = aes(x = pressure, y = wind))
```

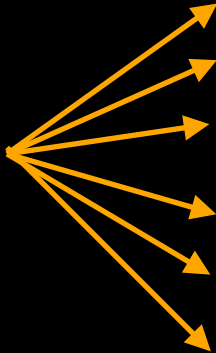
```
ggplot(data = storms) +  
  geom_point(mapping = aes(x = pressure, y = wind))
```

VISUAL ENCODING

VISUAL ENCODING

- Every visualization can be described as a set of mappings:
 - From data *items* to visual *marks*.
 - From data *variables* to visual *channels*.

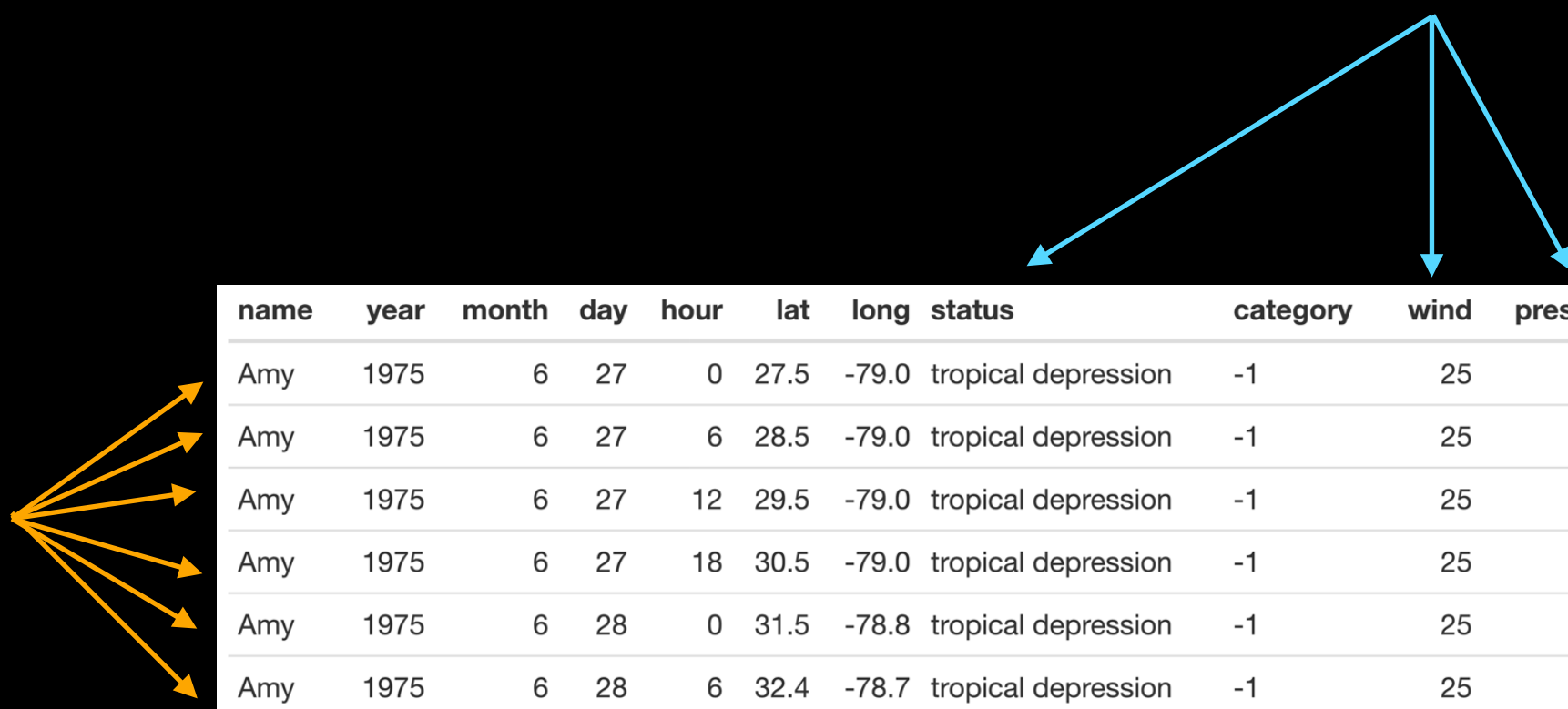
Items
(cases)



name	year	month	day	hour	lat	long	status	category	wind	pressure	ts_diameter	hu_diameter
Amy	1975	6	27	0	27.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	6	28.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	12	29.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	18	30.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	28	0	31.5	-78.8	tropical depression	-1	25	1012	NA	NA
Amy	1975	6	28	6	32.4	-78.7	tropical depression	-1	25	1012	NA	NA

Variables
(attributes)

Items
(cases)



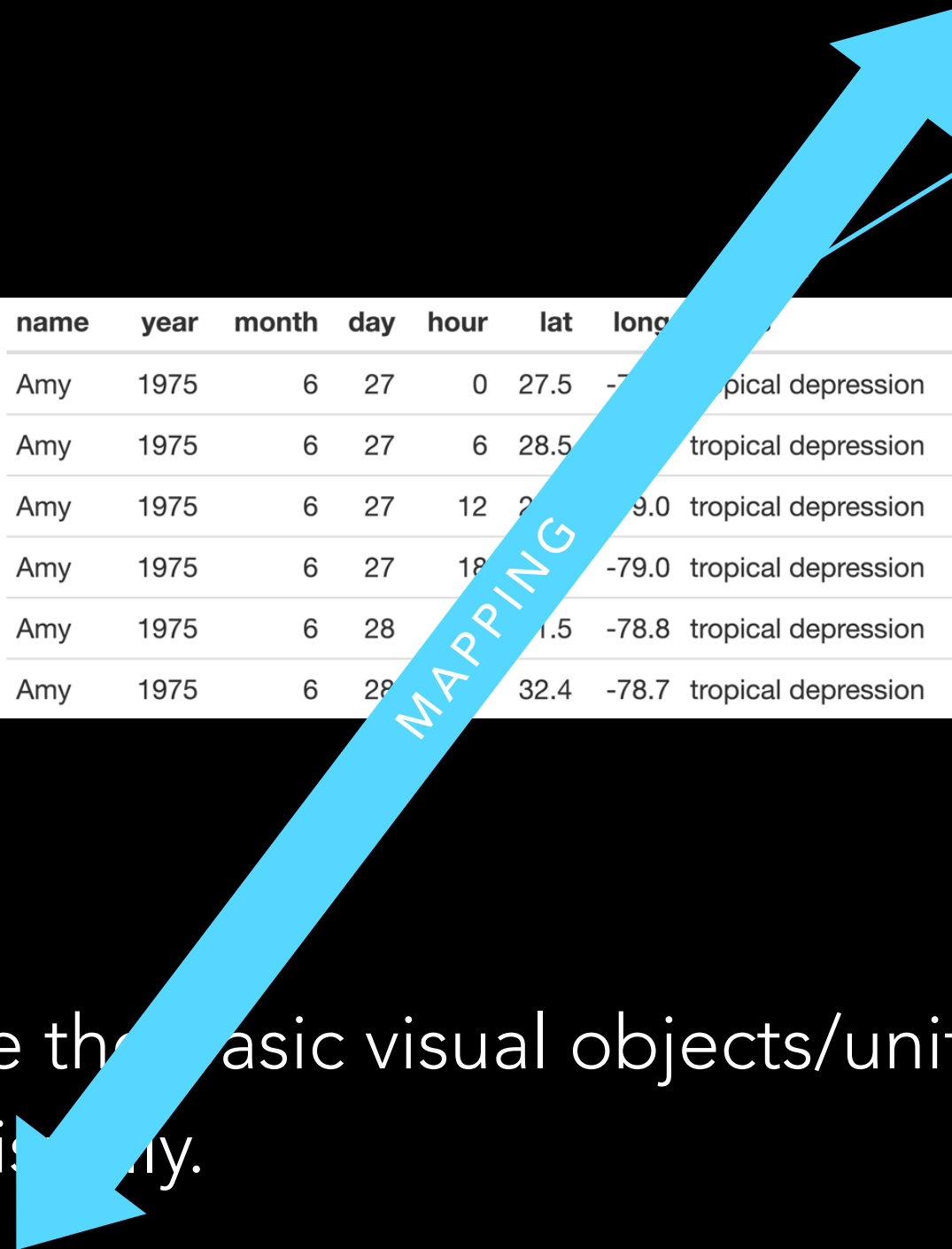
name	year	month	day	hour	lat	long	status	category	wind	pressure	ts_diameter	hu_diameter
Amy	1975	6	27	0	27.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	6	28.5	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	12	29.5	-79.0	tropical depression	-1	25	1013	NA	NA
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Amy	1975	6	28	6	32.4	-78.7	tropical depression	-1	25	1012	NA	NA

- **Marks** are the basic visual objects/units that represent data objects visually.
- **Visual channels** are visual variables we can use to represent characteristics of these objects.

Variables
(attributes)

name	year	month	day	hour	lat	long	category	wind	pressure	ts_diameter	hu_diameter	
Amy	1975	6	27	0	27.5	-78.9	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	6	28.5	-78.9	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	12	29.0	-79.0	tropical depression	-1	25	1013	NA	NA
Amy	1975	6	27	18	-79.0	-79.0	tropical depression	-1	25	1013	NA	NA
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Items
(cases)



- **Marks** are the basic visual objects/units that represent data objects visually.
- **Visual channels** are visual variables we can use to represent characteristics of these objects.

MARKS: BASIC GEOMETRIC ELEMENTS

➔ Points



0D

➔ Lines



1D

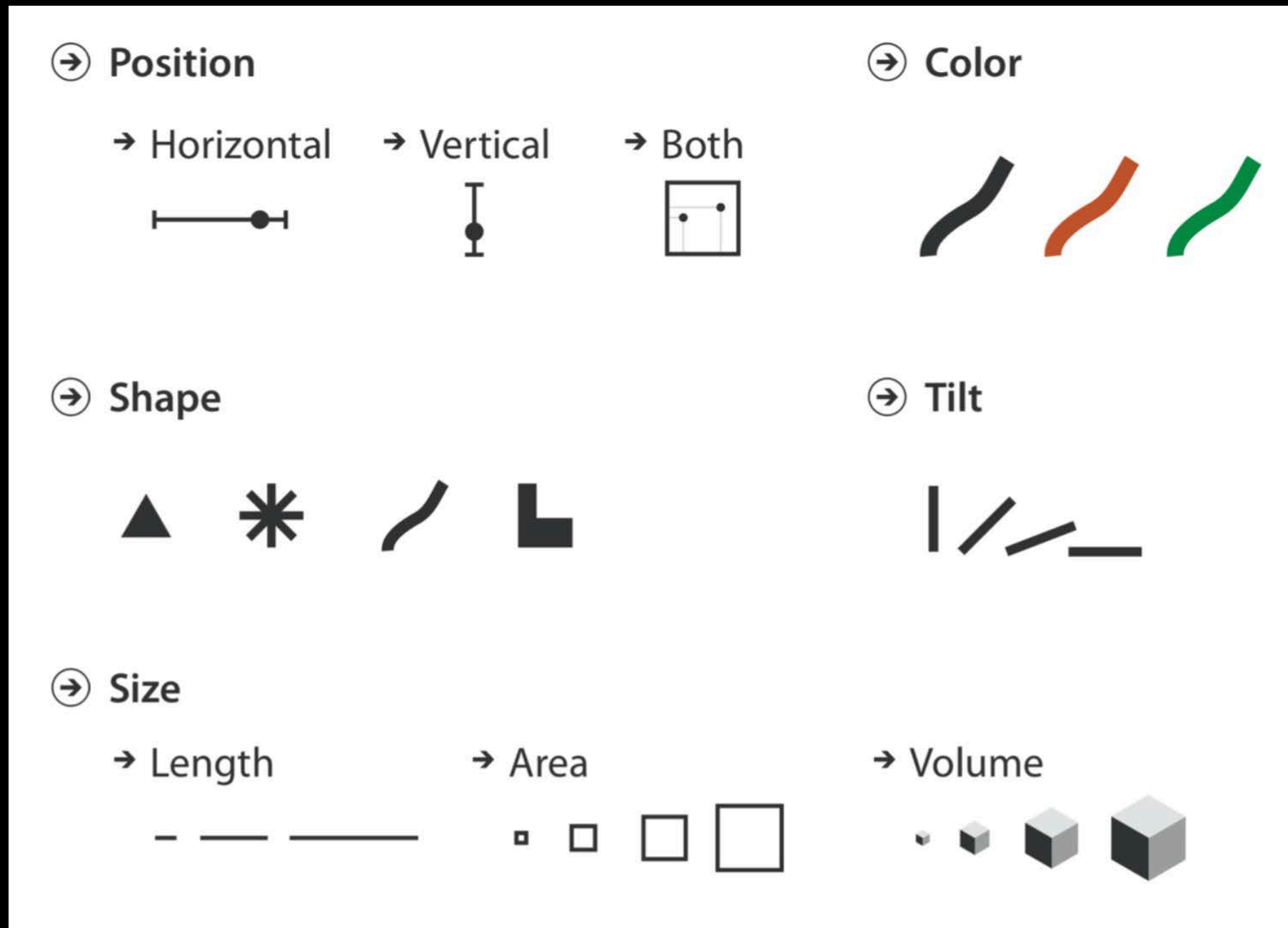
➔ Areas



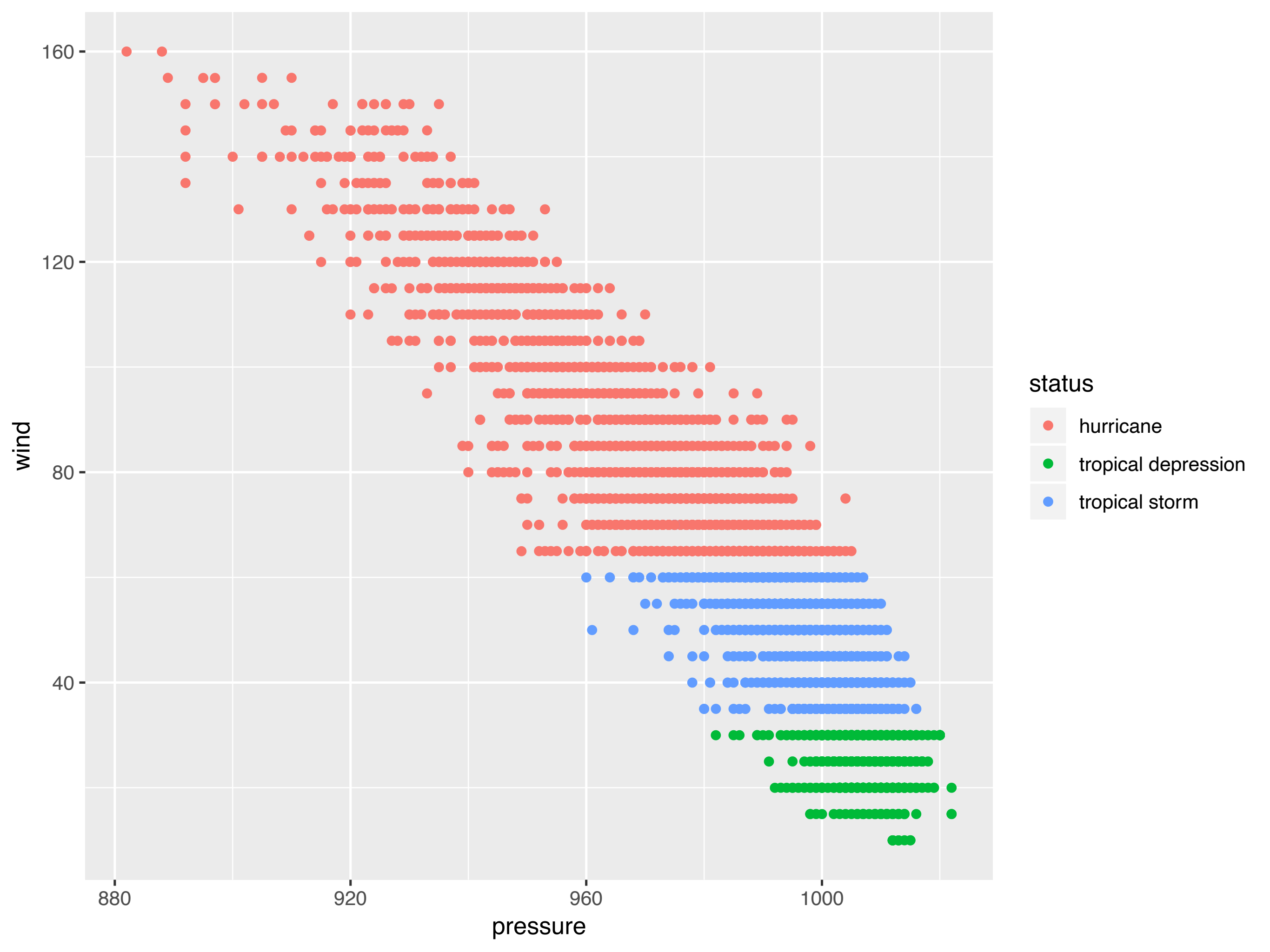
2D

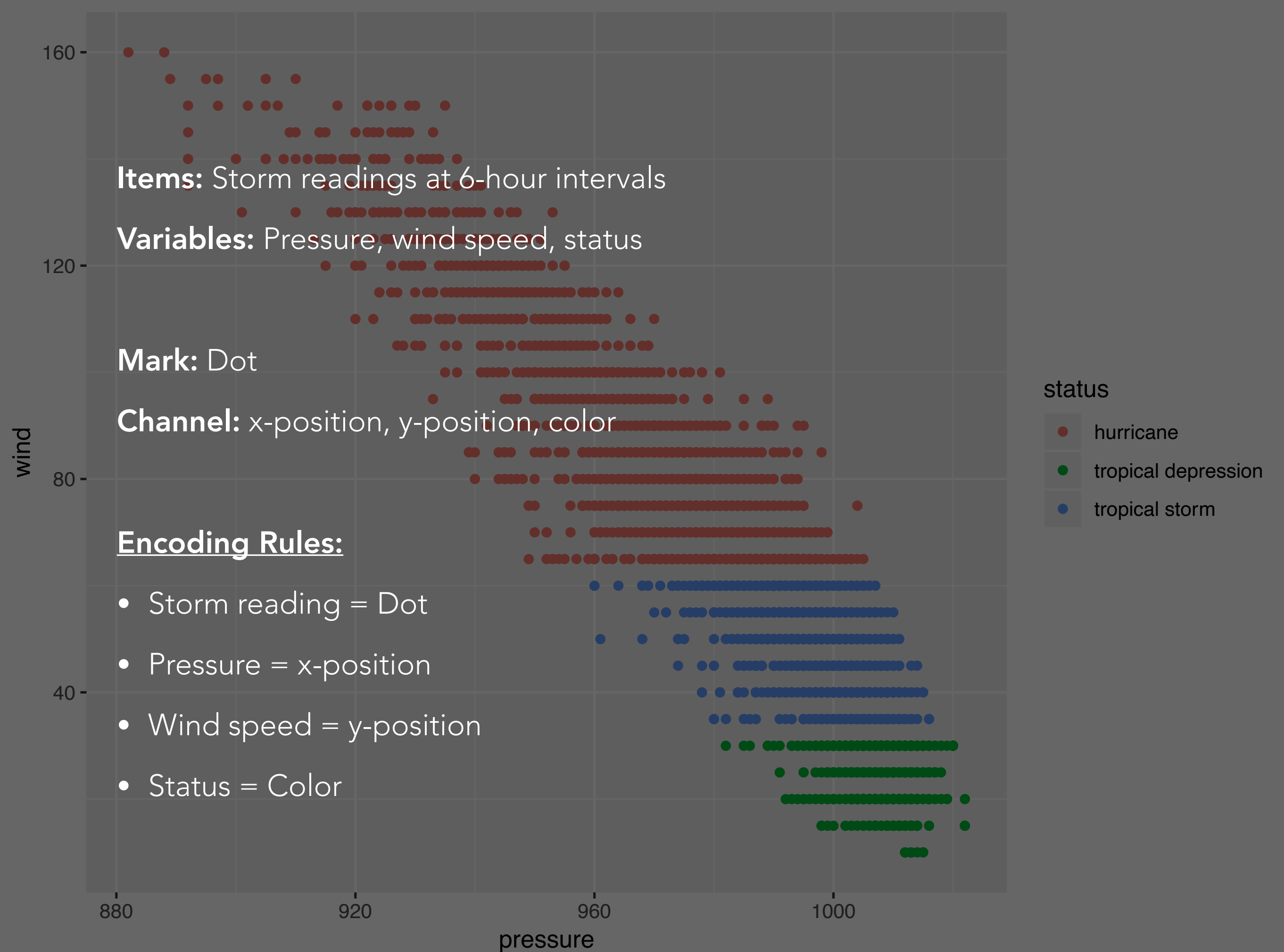
- These are “geoms” in ggplot2’s jargon

CHANNELS: VISUAL VARIABLES

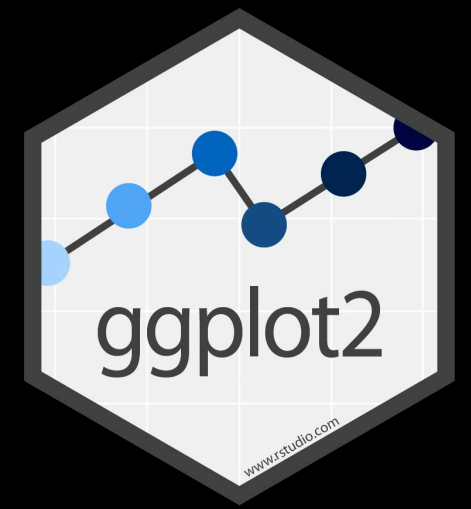


- These are “aesthetics” in ggplot2’s jargon

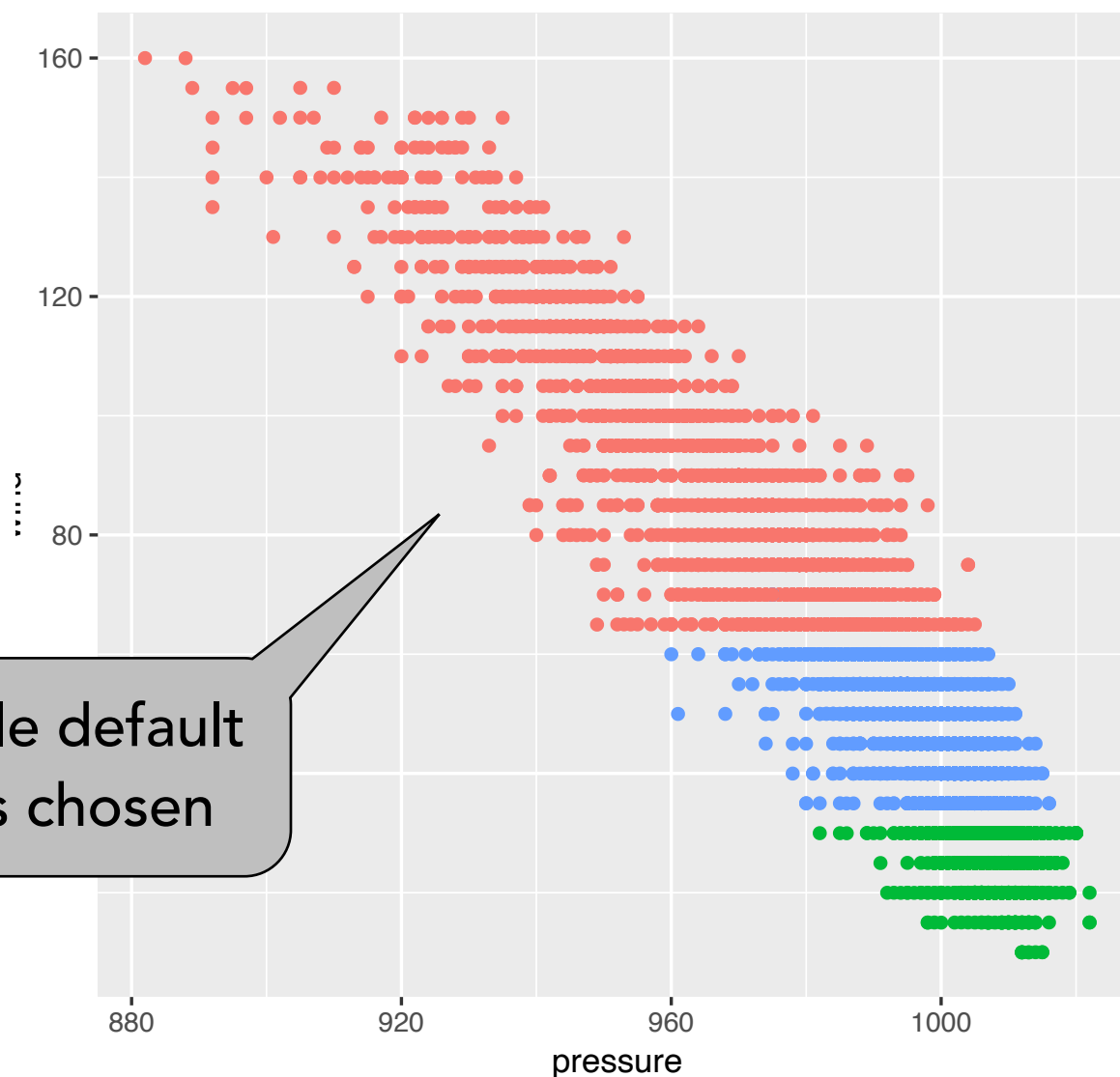




AESTHETICS



```
ggplot(data = storms) +  
  geom_point(mapping = aes(x = pressure, y = wind, color = status))
```



Sensible default
colors chosen

Channel
(aesthetic
property)

Variable to
map to the
channel

status

- hurricane
- tropical depression
- tropical storm

Legend added
automatically

To make a graph

[template]

```
ggplot(data = <DATA>) +  
  <GEOM_FUNCTION>(mapping = aes(<MAPPINGS>))
```

To make a graph

1. Pick a **data** set

mpg	cyl	disp	hp
21.0	6	160.0	2
21.0	6	160.0	2
22.8	4	108.0	1
21.4	6	258.0	2
18.7	8	360.0	3
18.1	6	225.0	2
14.3	8	360.0	5
24.4	4	146.7	1
22.8	4	140.8	1
19.2	6	167.6	2
17.8	6	167.6	2
16.4	8	275.8	3
17.3	8	275.8	3
15.2	8	275.8	3
10.4	8	472.0	4
10.4	8	460.0	4
14.7	8	440.0	4
32.4	4	78.7	1
30.4	4	75.7	1
33.9	4	71.1	1

data

```
ggplot(data = <DATA>) +  
  <GEOM_FUNCTION>(mapping = aes(<MAPPINGS>))
```



To make a graph

mpg	cyl	disp	hp
21.0	6	160.0	2
21.0	6	160.0	2
22.8	4	108.0	1
21.4	6	258.0	2
18.7	8	360.0	3
18.1	6	225.0	2
14.3	8	360.0	5
24.4	4	146.7	1
22.8	4	140.8	1
19.2	6	167.6	2
17.8	6	167.6	2
16.4	8	275.8	3
17.3	8	275.8	3
15.2	8	275.8	3
10.4	8	472.0	4
10.4	8	460.0	4
14.7	8	440.0	4
32.4	4	78.7	1
30.4	4	75.7	1
33.9	4	71.1	1

data

geom

1. Pick a **data** set

```
ggplot(data = <DATA>) +  
  <GEOM_FUNCTION>(mapping = aes(<MAPPINGS>))
```

2. Choose a **geom**
to display cases



To make a graph

mappings

mpg	cyl	disp	hp	fill	geom
21.0	6	160.0	2	blue	point
21.0	6	160.0	2	blue	point
22.8	4	108.0	1	green	point
21.4	6	258.0	2	blue	point
18.7	8	360.0	3	red	point
18.1	6	225.0	2	blue	point
14.3	8	360.0	5	purple	point
24.4	4	146.7	1	green	point
22.8	4	140.8	1	green	point
19.2	6	167.6	2	blue	point
17.8	6	167.6	2	blue	point
16.4	8	275.8	3	red	point
17.3	8	275.8	3	red	point
15.2	8	275.8	3	red	point
10.4	8	472.0	4	yellow	point
10.4	8	460.0	4	yellow	point
14.7	8	440.0	4	yellow	point
32.4	4	78.7	1	green	point
30.4	4	75.7	1	green	point
33.9	4	71.1	1	green	point

1. Pick a **data** set

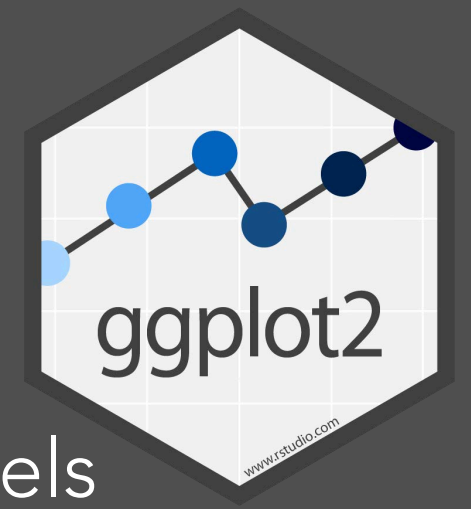
```
ggplot(data = <DATA>) +  
  <GEOM_FUNCTION>(mapping = aes(<MAPPINGS>))
```

2. Choose a **geom**
to display cases

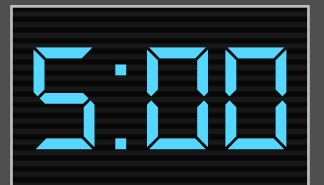
3. **Map** aesthetic
properties to
variables



YOUR TURN



- In your R script, experiment with different visual channels by mapping color, shape, size, and alpha to storm status.
- What happens when you map the same variable to multiple aesthetics?
 - Hint: `color = status, shape = status`.
- Try mapping color to the variable category rather than status. What changes and why?
- What happens if you map color to two different variables (both status and category)?



LET'S TAKE A SHORT BREAK

Resume at:

10:15

TYPES OF CHANNELS

➔ **Magnitude** Channels: **Ordered** Attributes

Position on common scale



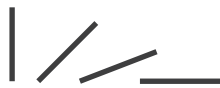
Position on unaligned scale



Length (1D size)



Tilt/angle



Area (2D size)



Depth (3D position)



Color luminance



Color saturation



Curvature



Volume (3D size)



Same

➔ **Identity** Channels: **Categorical** Attributes

Spatial region



Color hue



Motion



Shape



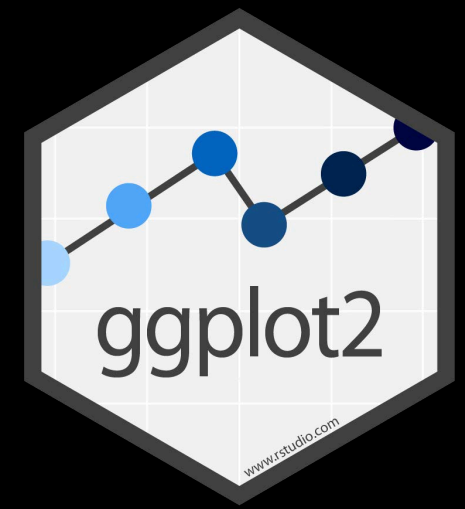
Effectiveness

Most

Least

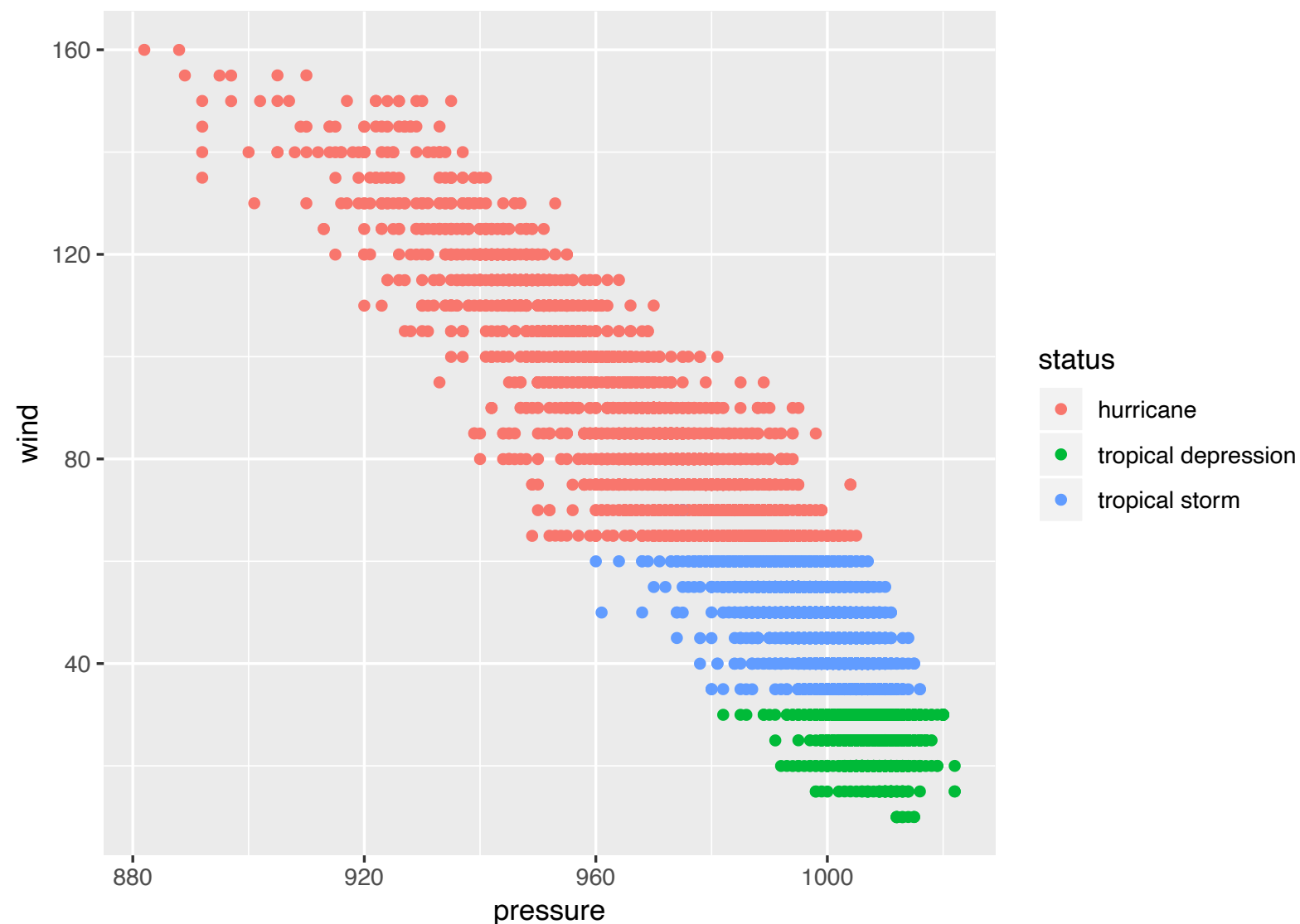
MAPPING VS SETTING
AESTHETIC PROPERTIES

LET'S GO BACK TO THIS PLOT...

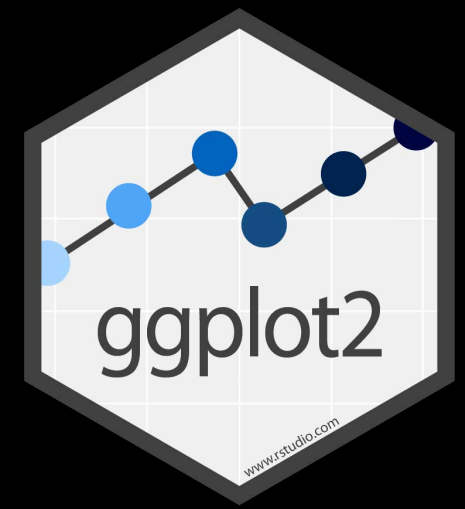


First, a time-saver...

```
ggplot(data = storms) +  
  geom_point(mapping = aes(x = pressure, y = wind, color = status))
```

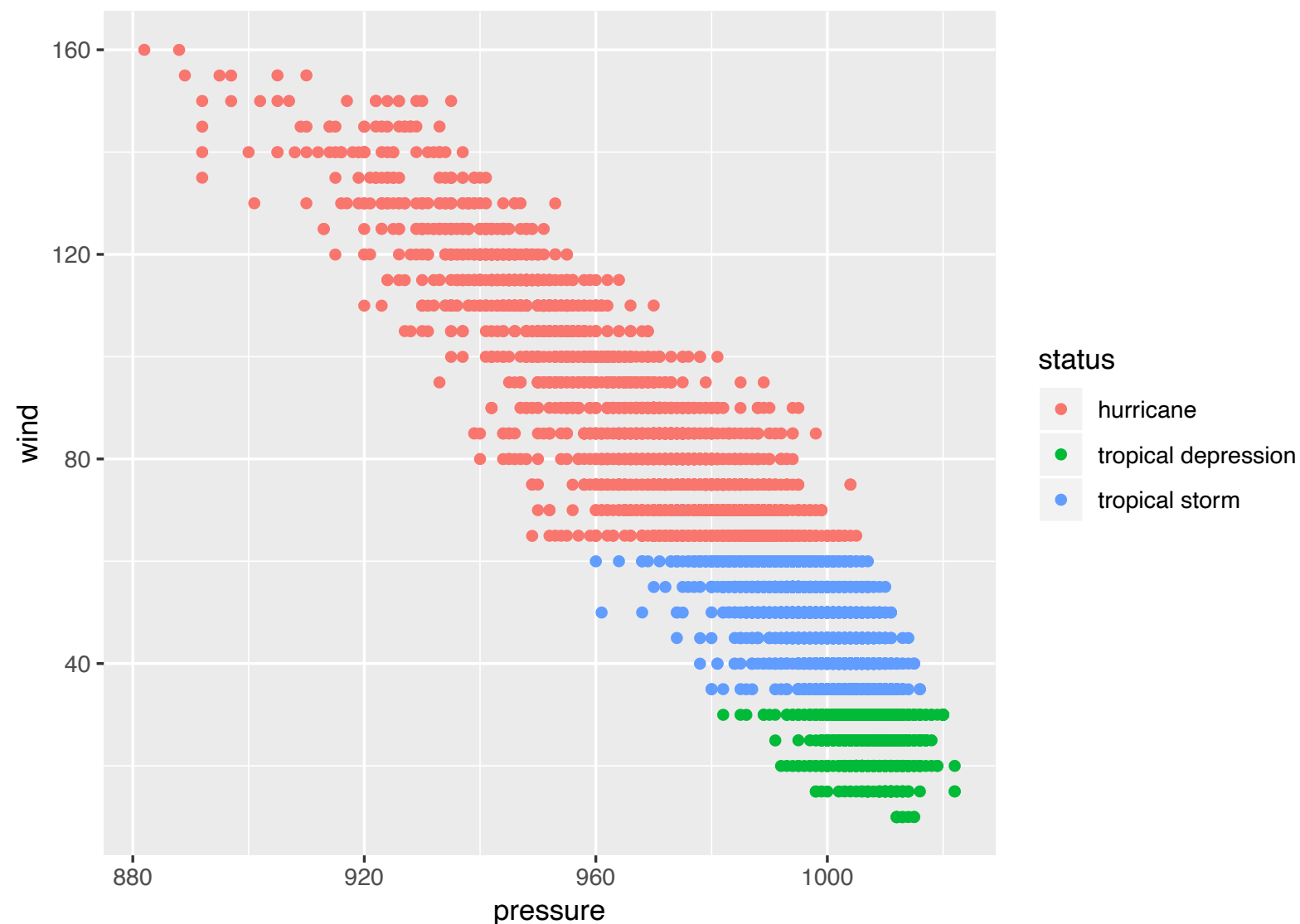


LET'S GO BACK TO THIS PLOT...

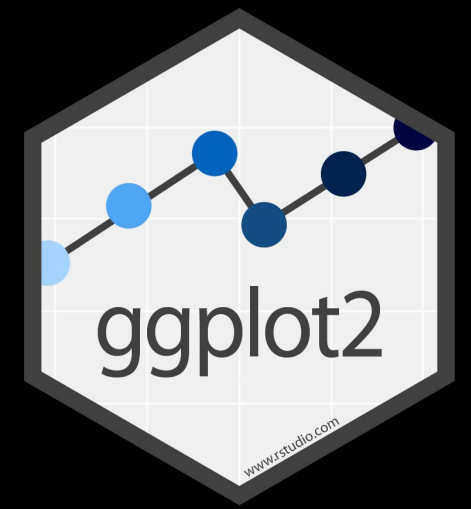


...these can be omitted

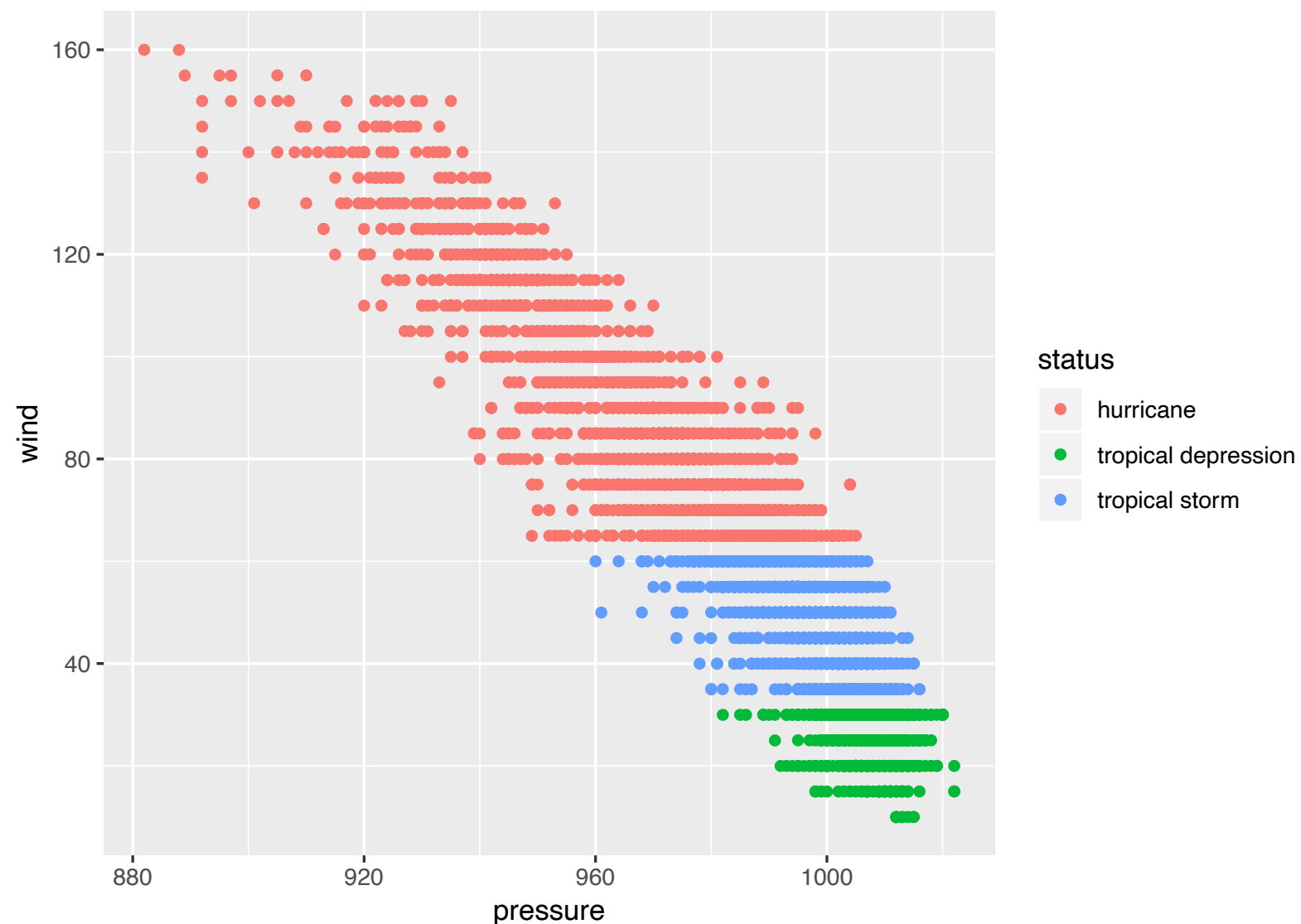
```
ggplot(data = storms) +  
  geom_point(mapping = aes(x = pressure, y = wind, color = status))
```



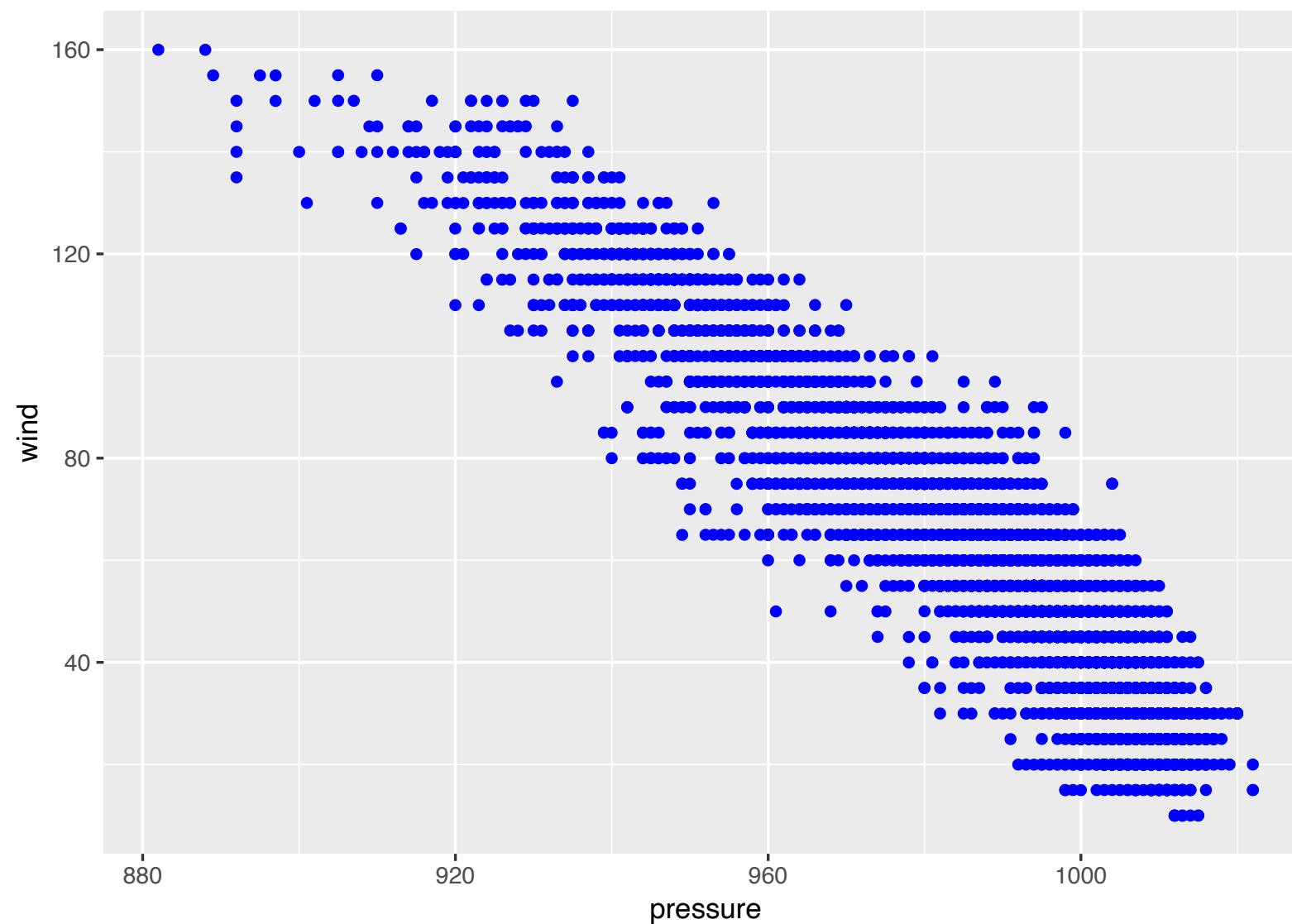
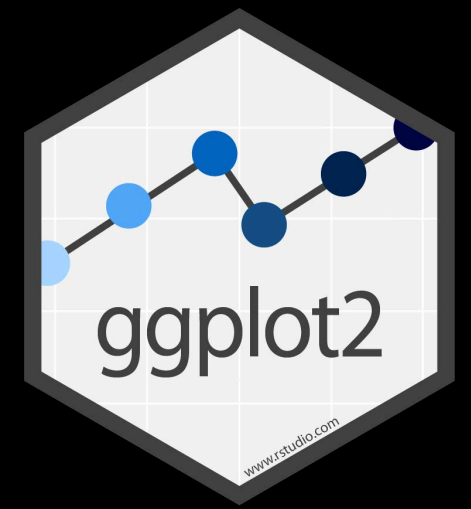
LET'S GO BACK TO THIS PLOT...



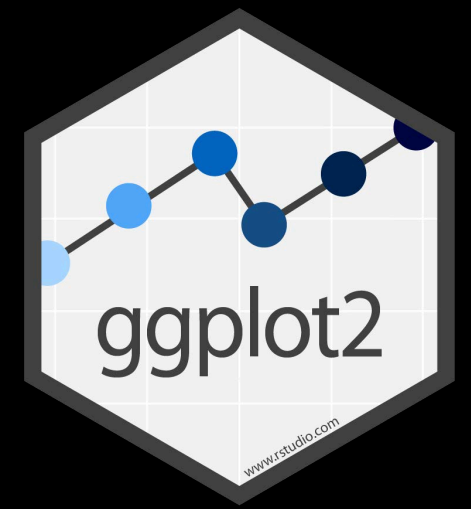
```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind, color = status))
```



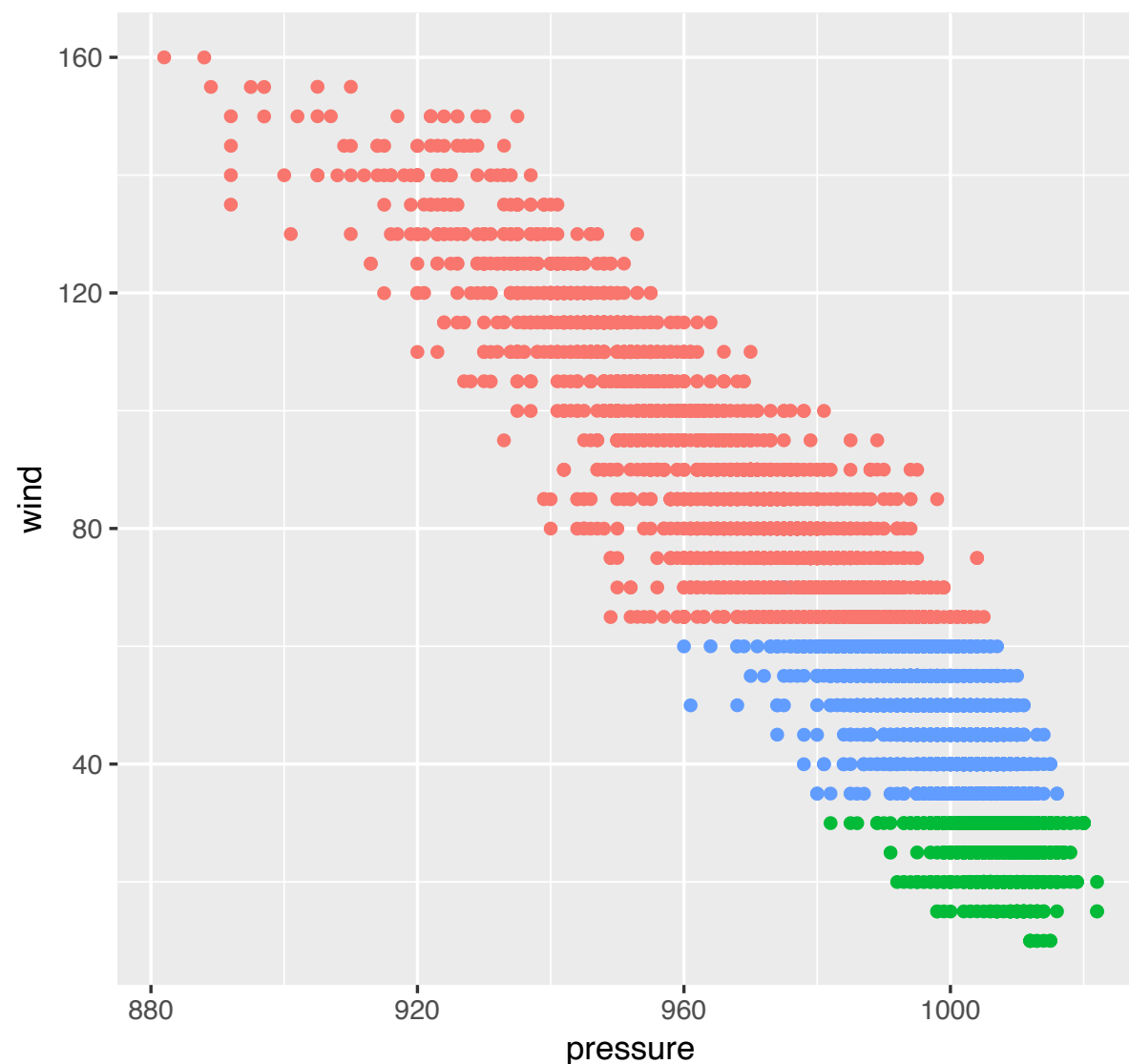
HOW WOULD YOU MAKE THIS PLOT?



AESTHETICS



```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind, color = status))
```

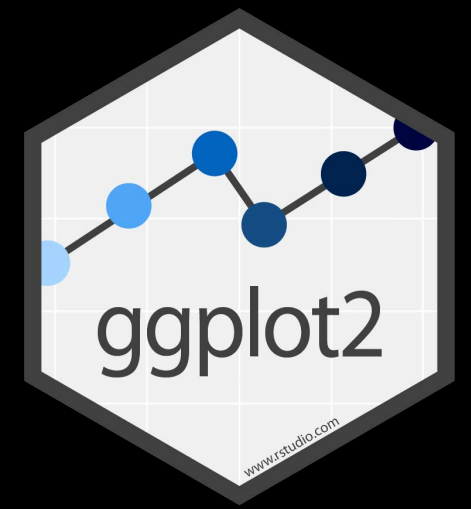


Inside of `aes()`: maps an aesthetic to a variable

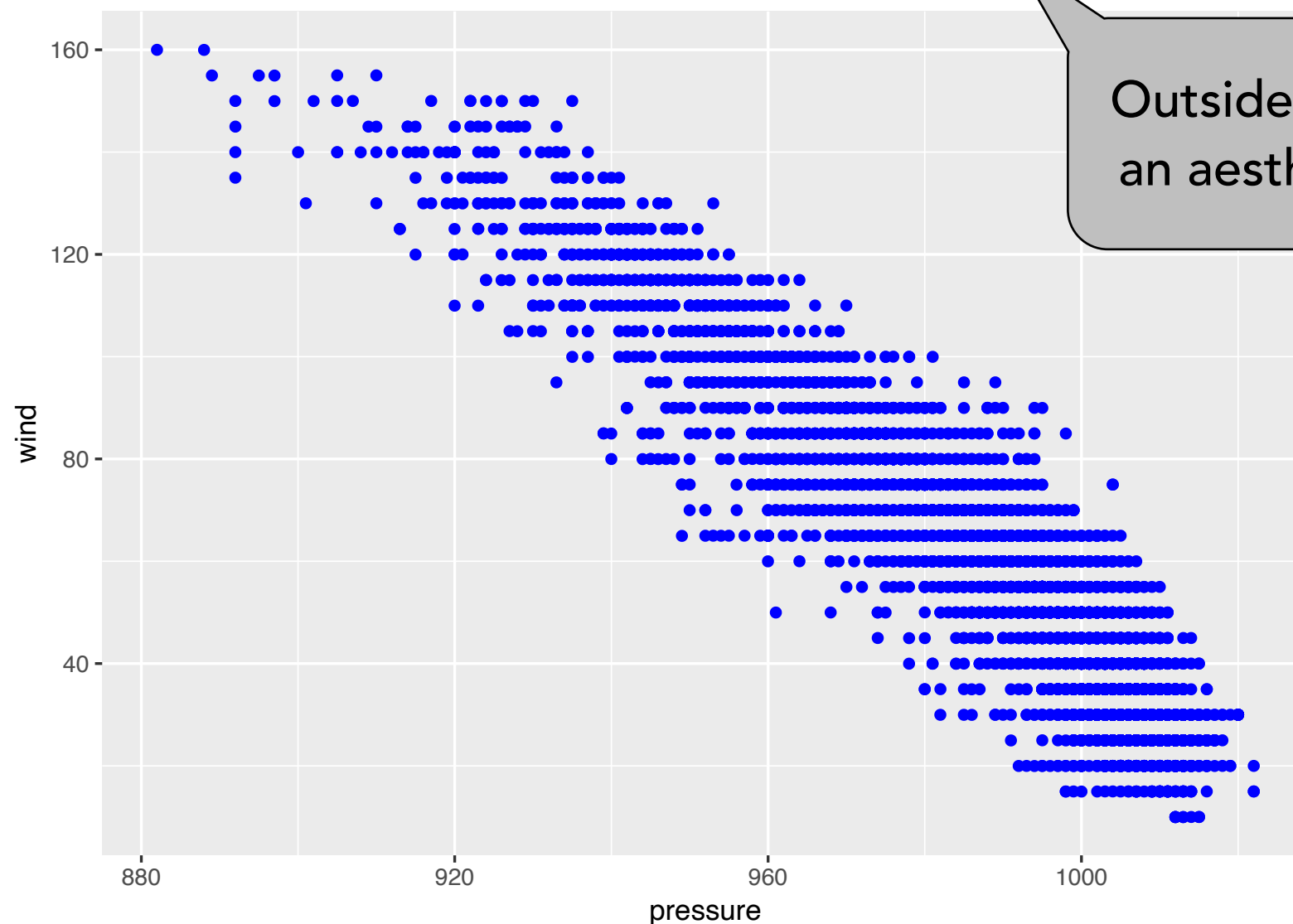
status

- hurricane
- tropical depression
- tropical storm

AESTHETICS

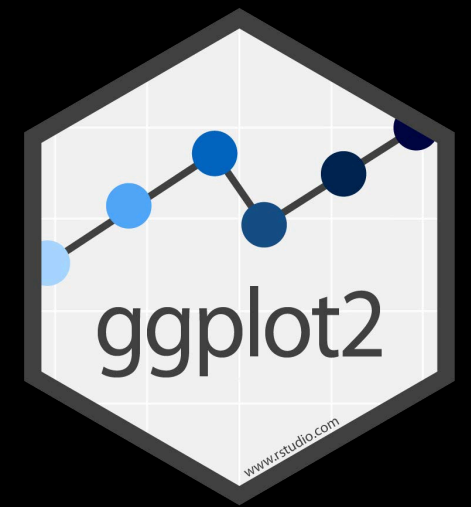


```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind), color = "blue")
```

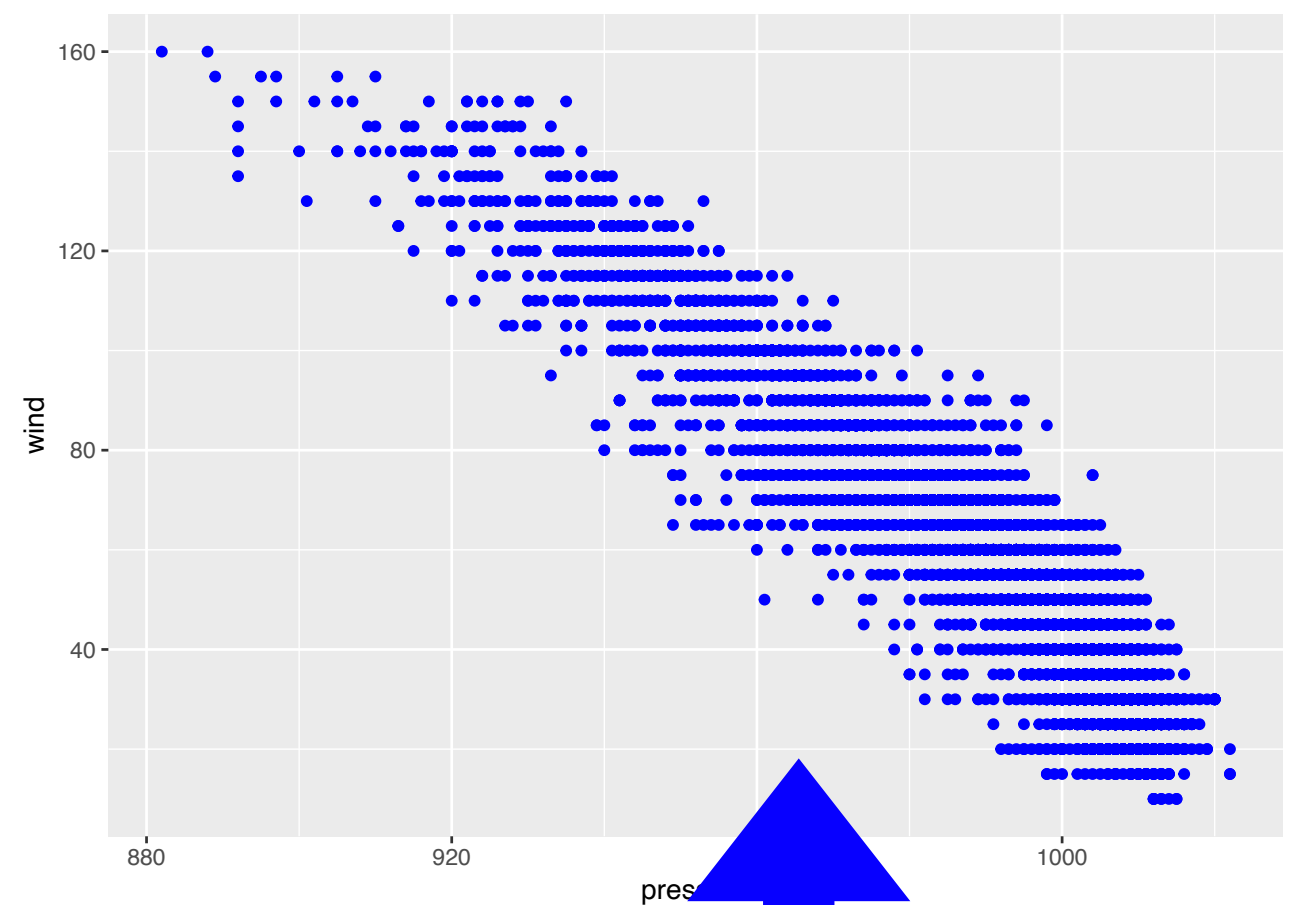
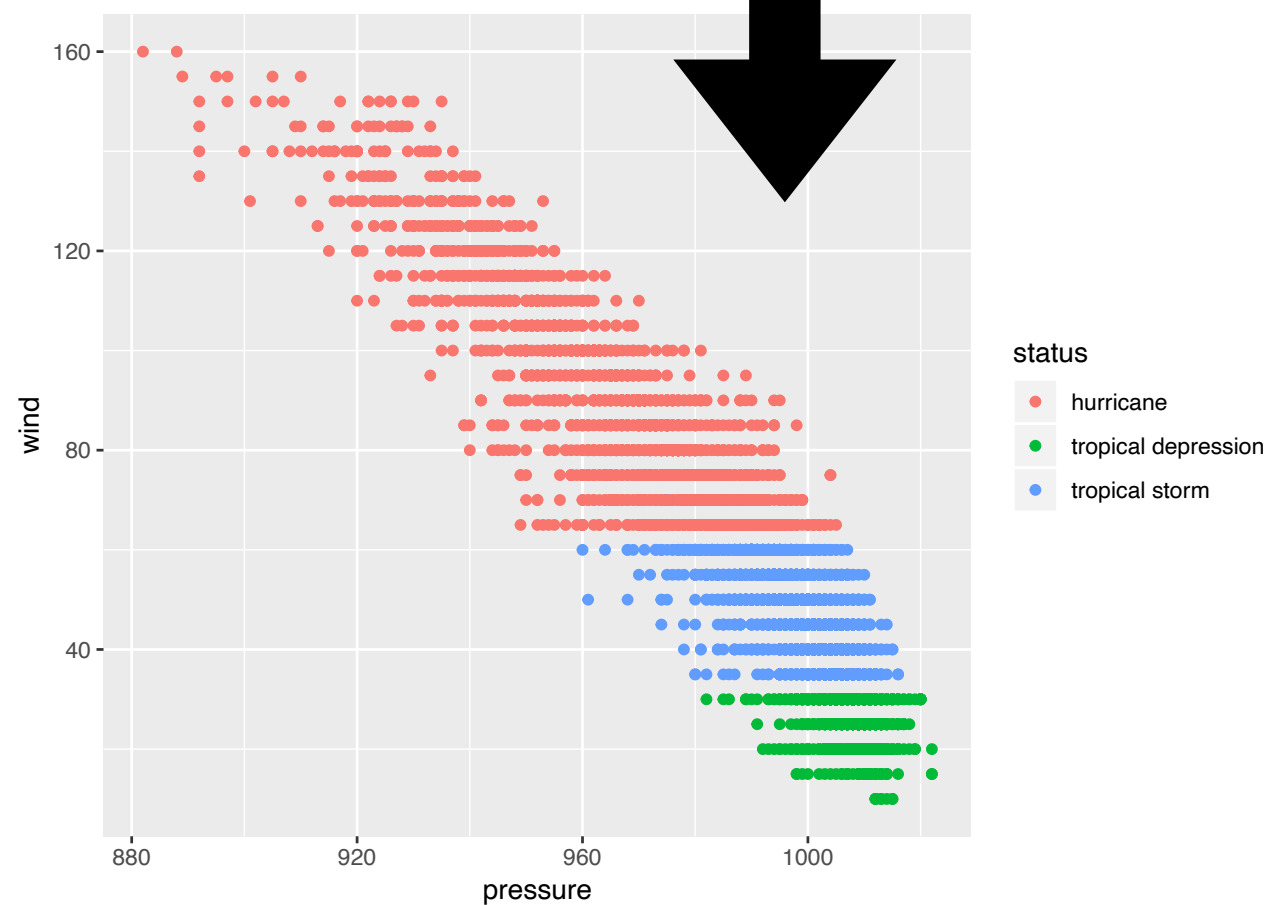


Outside of `aes()`: sets
an aesthetic to a value

AESTHETICS

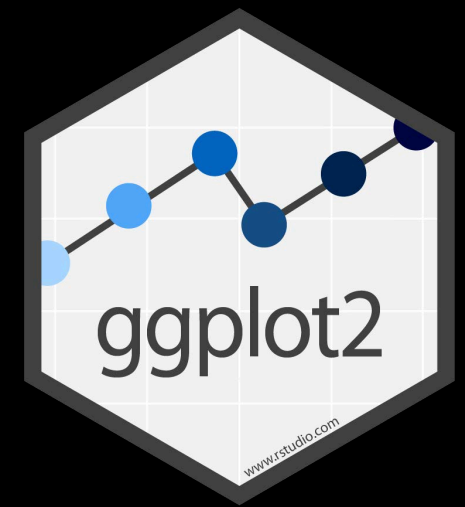


```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind, color = status))
```



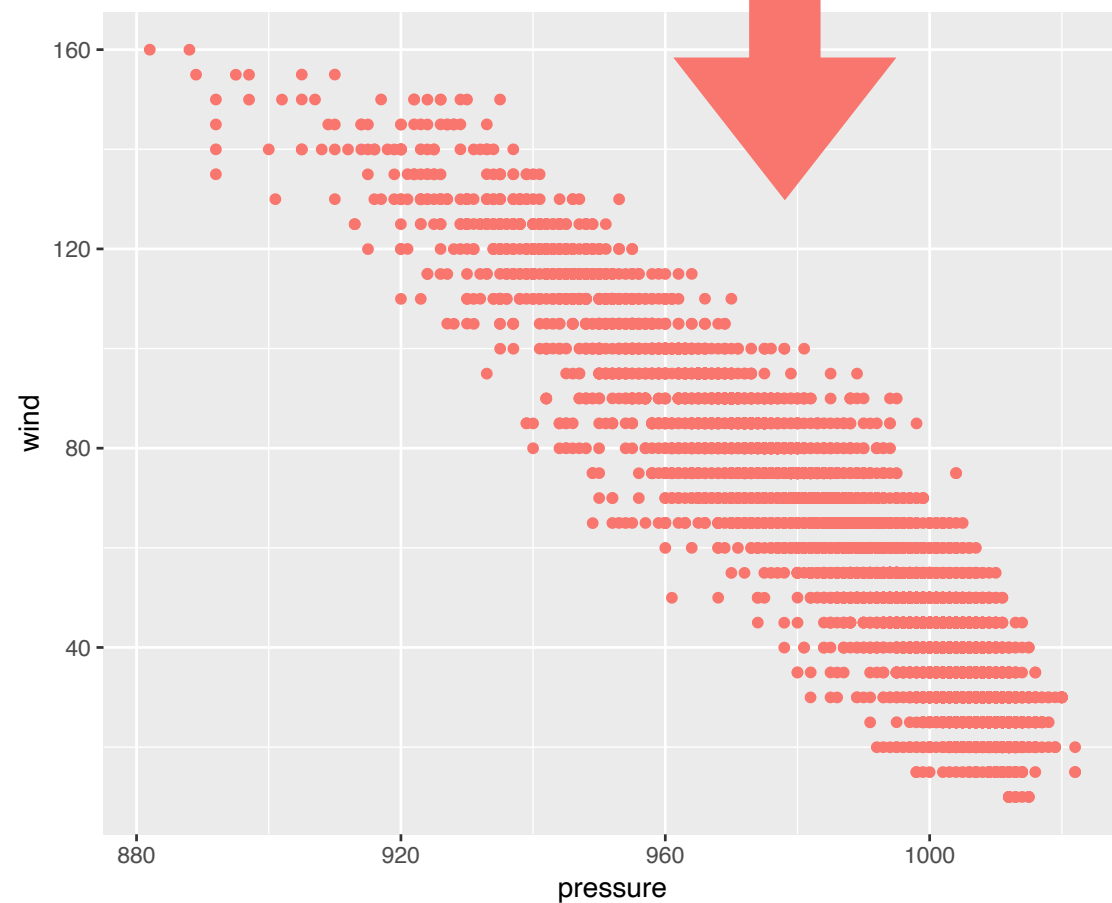
```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind), color = "blue")
```

AESTHETICS

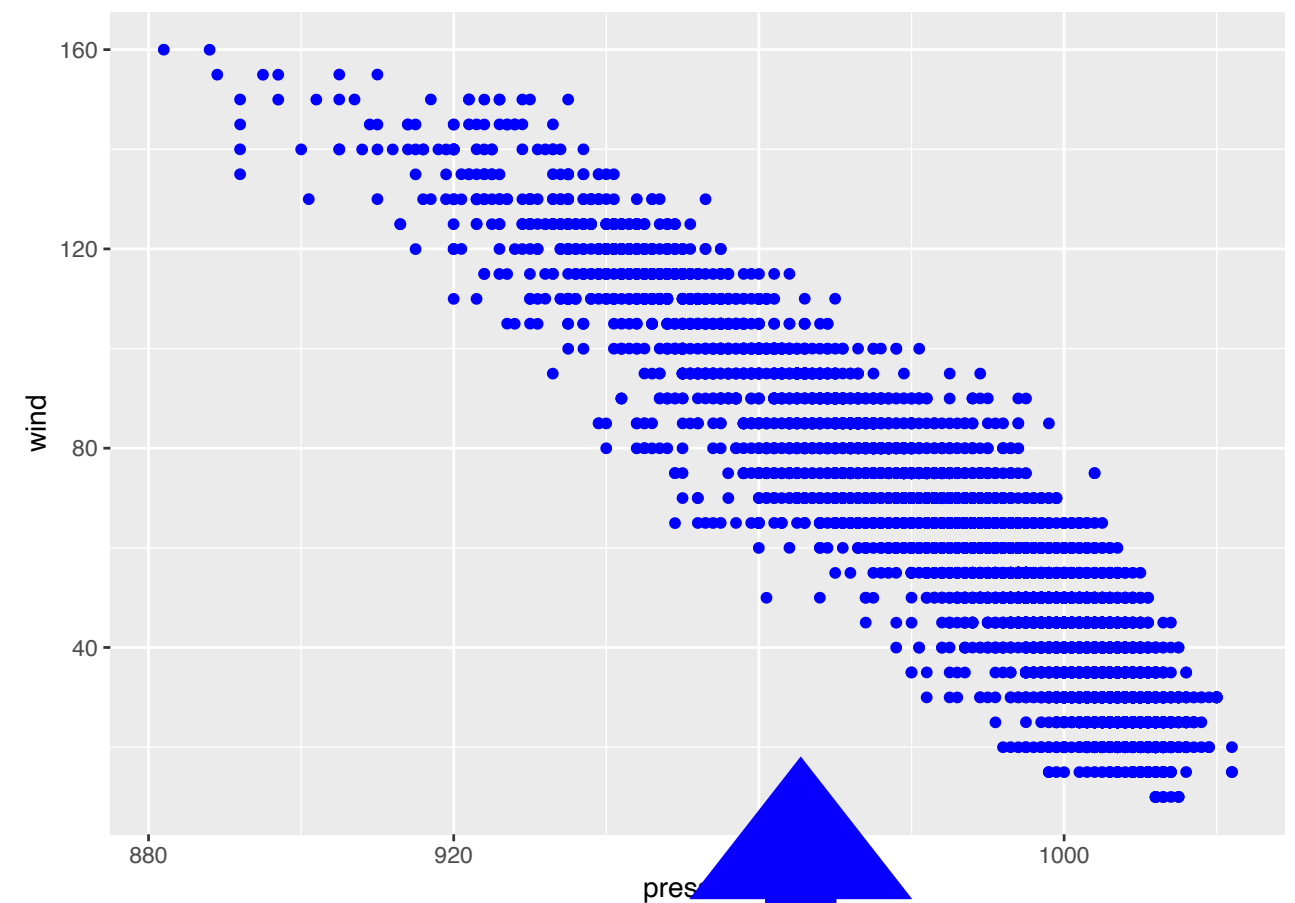


Watch out for this!

```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind, color = "blue"))
```

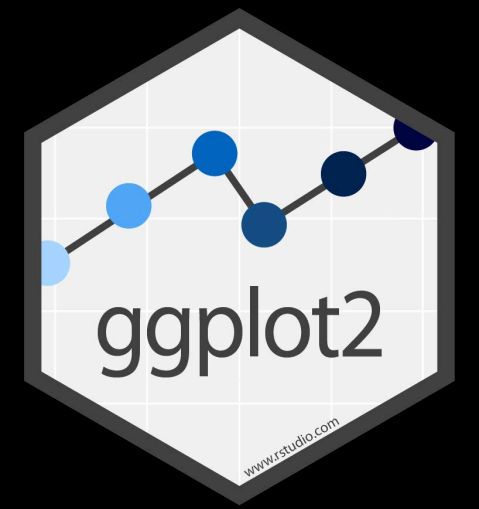


colour
• blue

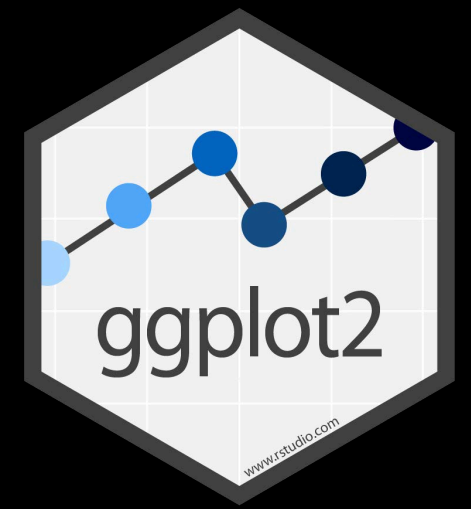


```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind), color = "blue")
```

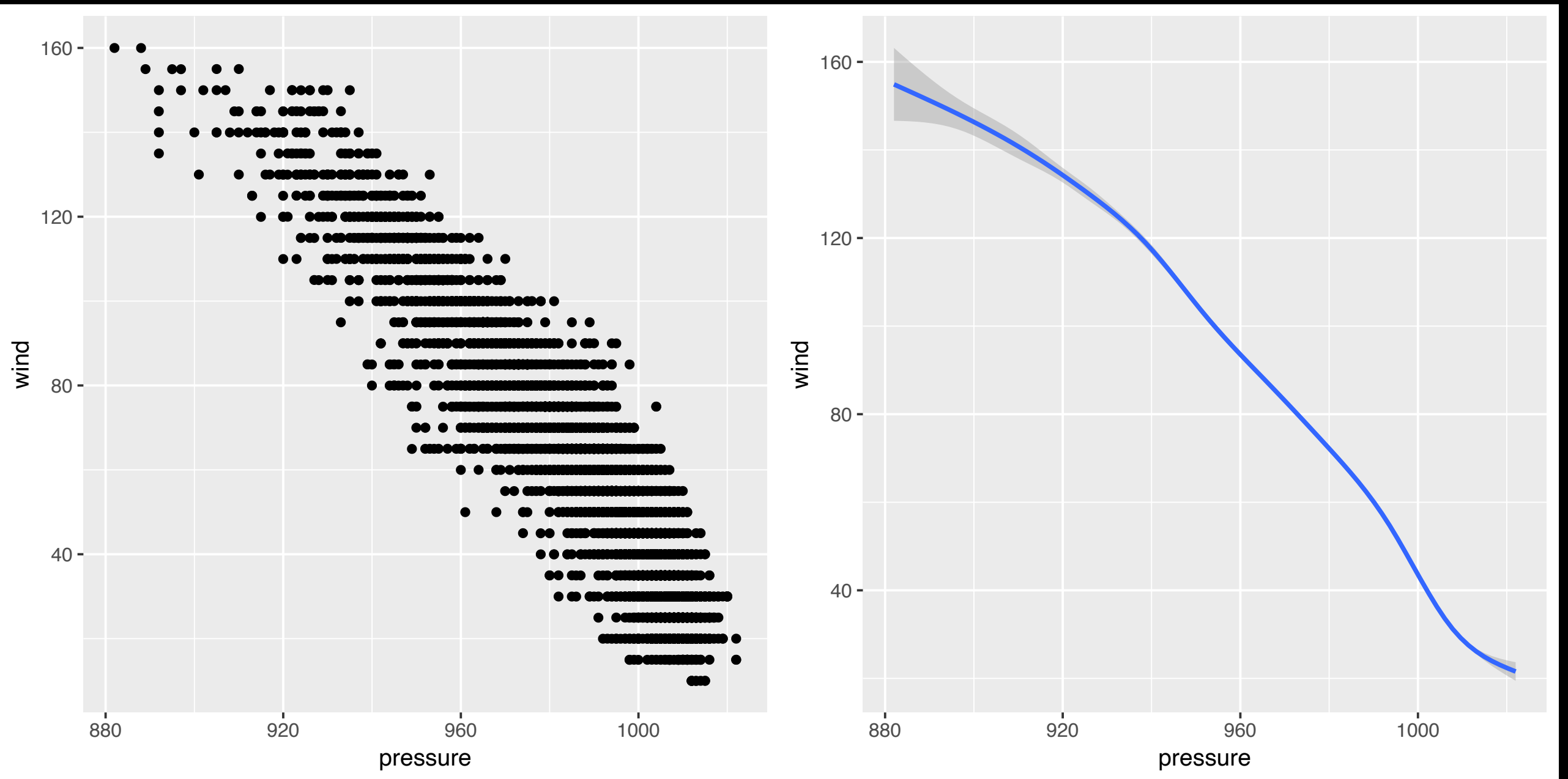
GEOMS



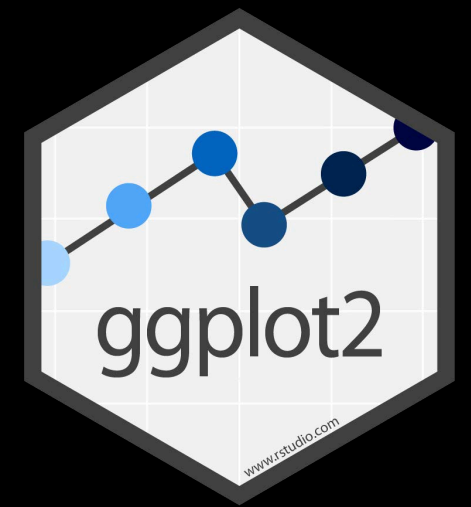
GEOMS



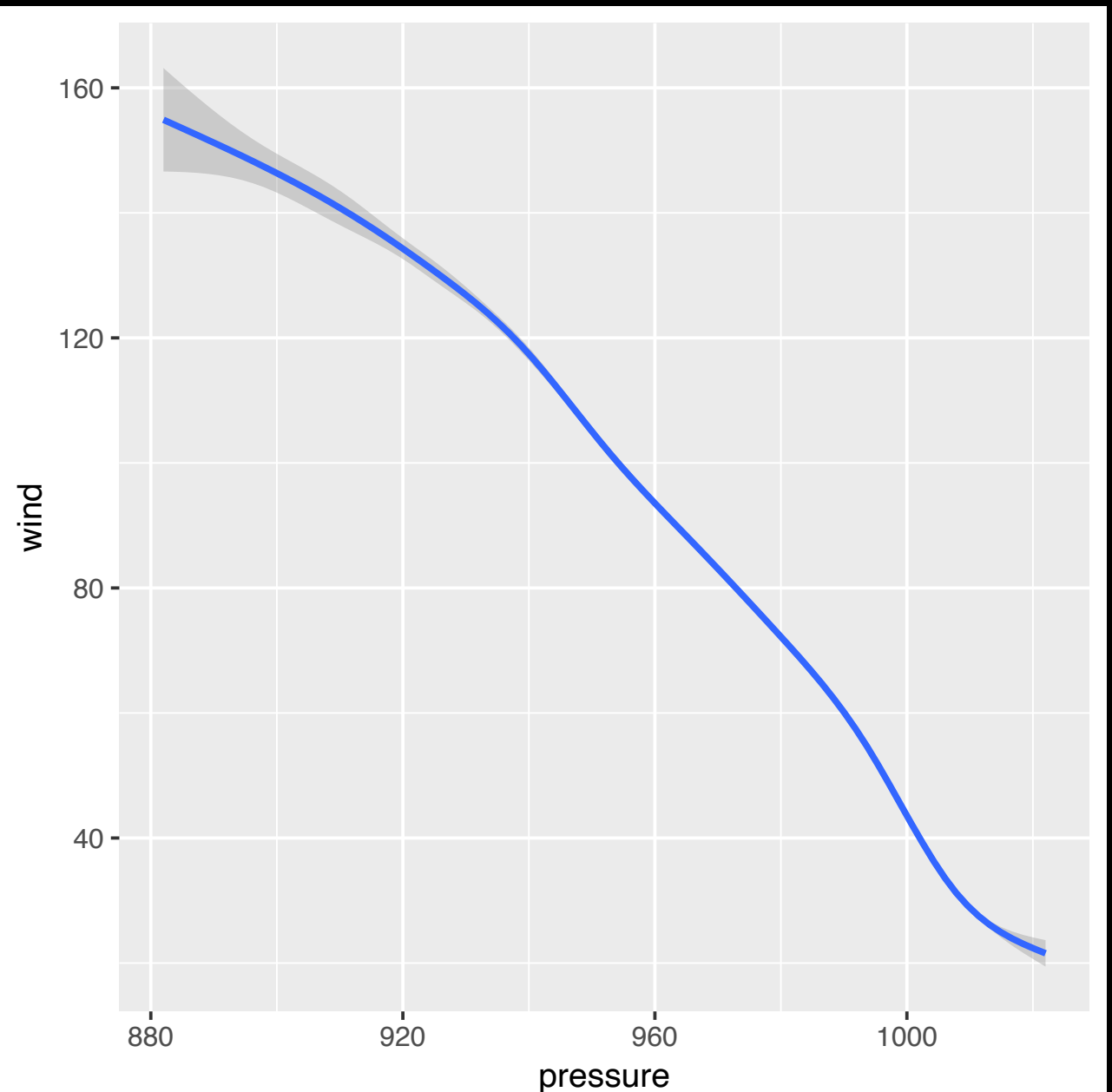
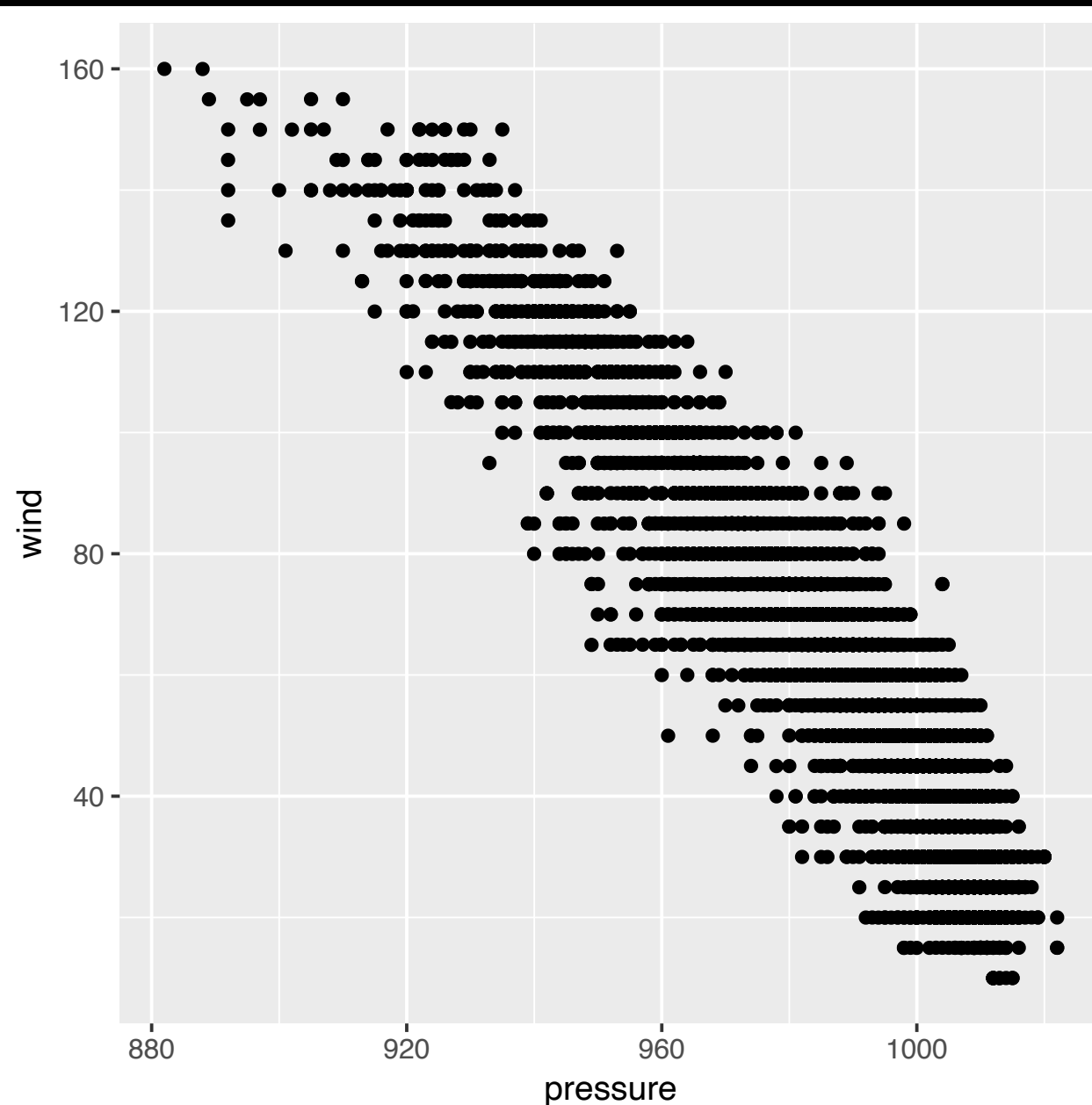
- How are these plots similar?



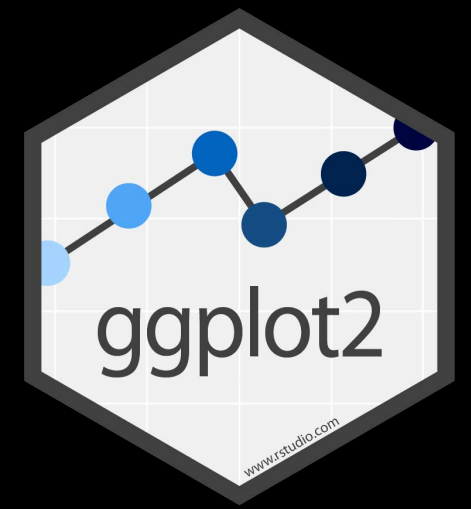
GEOMS



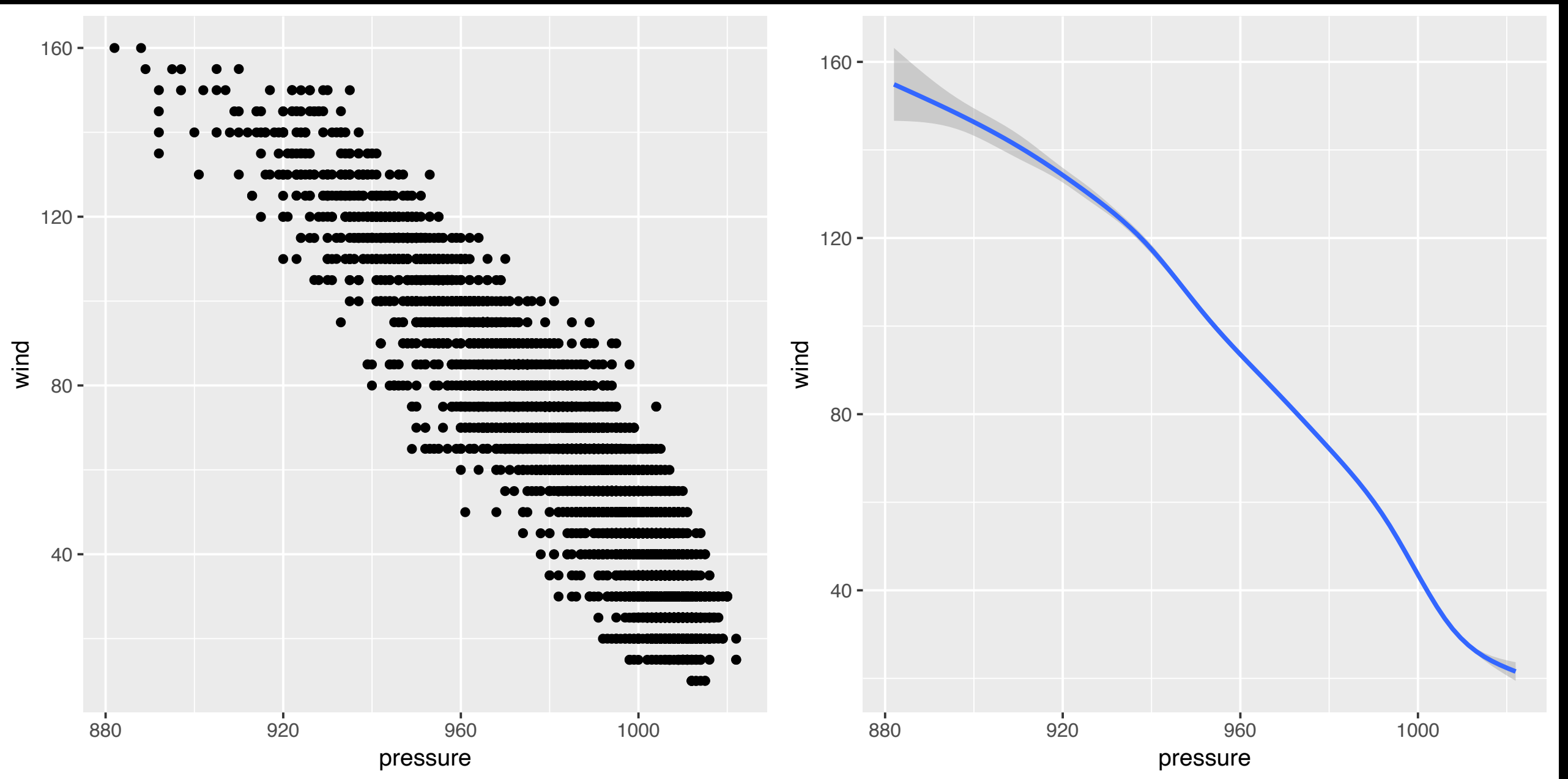
- **Same:** x variable, y variable → built with same data



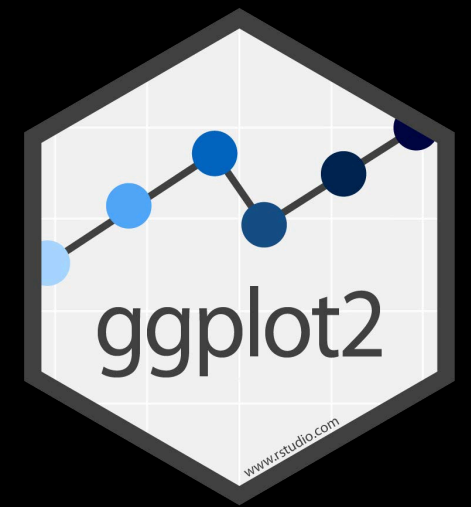
GEOMS



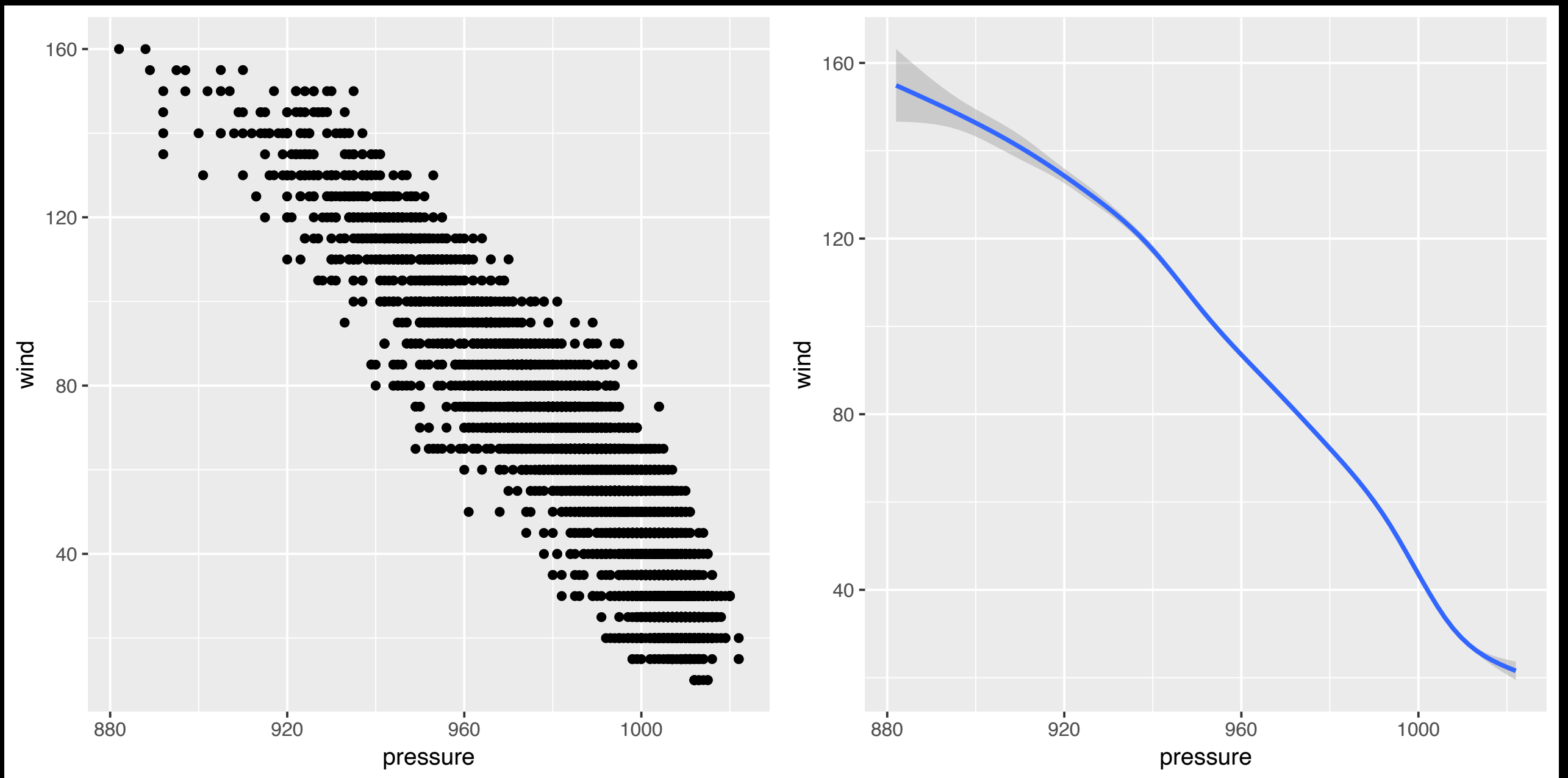
- How are these plots different?



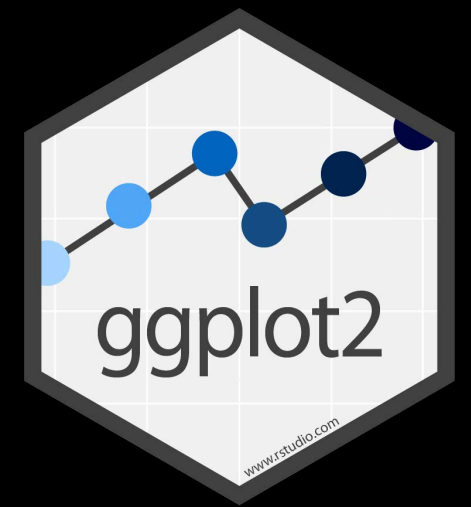
GEOMS



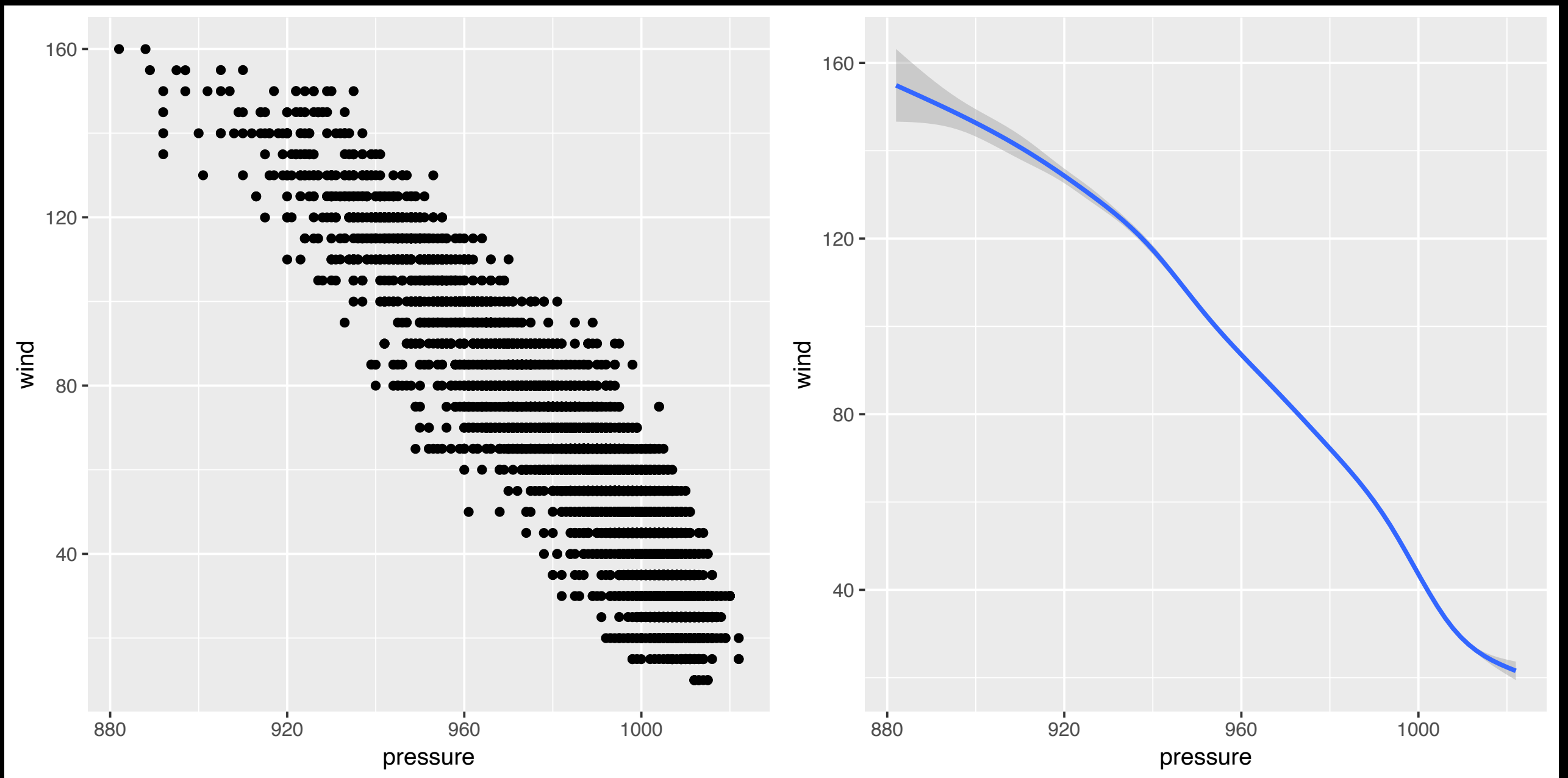
- **Different:** geometric object (point vs. line & ribbon)



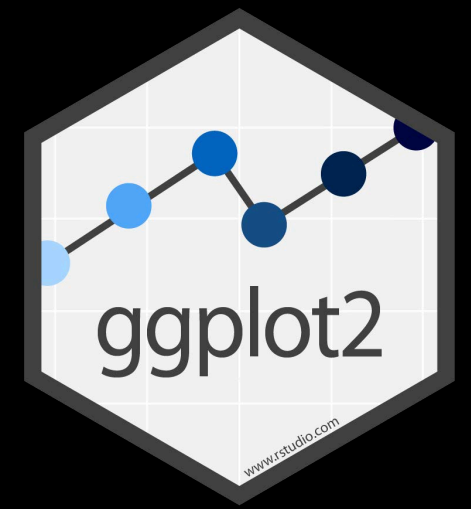
GEOMS



- **Different:** statistical summarization (identity vs. smooth)



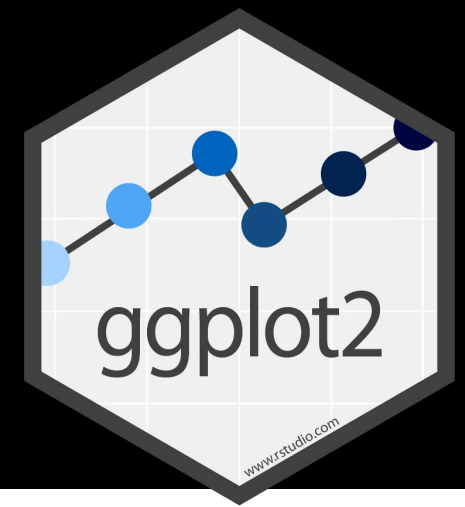
GEOMS



- Remember our template:

```
ggplot(data = <DATA>) +  
  <GEOM_FUNCTION>(mapping = aes(<MAPPINGS>))
```

GEOMS



Each `geom_` function requires a mapping








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

One Variable


Continuous

`a <- ggplot(mpg, aes(hwy))`

-  **a + geom_area(stat = "bin")**
x, y, alpha, color, fill, linetype, size
b + geom_area(aes(y = ..density..), stat = "bin")
-  **a + geom_density(kernel = "gaussian")**
x, y, alpha, color, fill, linetype, size, weight
b + geom_density(aes(y = ..county..))
-  **a + geom_dotplot()**
x, y, alpha, color, fill
-  **a + geom_freqpoly()**
x, y, alpha, color, linetype, size
b + geom_freqpoly(aes(y = ..density..))
-  **a + geom_histogram(binwidth = 5)**
x, y, alpha, color, fill, linetype, size, weight
b + geom_histogram(aes(y = ..density..))




Discrete

`b <- ggplot(mpg, aes(fl))`

-  **b + geom_bar()**
x, alpha, color, fill, linetype, size, weight

Graphical Primitives

`map <- map_data("state")`
`c <- ggplot(map, aes(long, lat))`

-  **c + geom_polygon(aes(group = group))**
x, y, alpha, color, fill, linetype, size
- `d <- ggplot(economics, aes(date, unemploy))`
-  **d + geom_path(lineend = "butt", linejoin = "round", linemitre = 1)**
x, y, alpha, color, linetype, size
-  **d + geom_ribbon(aes(ymin = unemploy - 900, ymax = unemploy + 900))**
x, ymax, ymin, alpha, color, fill, linetype, size








`e <- ggplot(seals, aes(x = long, y = lat))`

-  **e + geom_segment(aes(xend = long + delta_long, yend = lat + delta_lat))**
x, xend, y, yend, alpha, color, linetype, size
-  **e + geom_rect(aes(xmin = long, ymin = lat, xmax = long + delta_long, ymax = lat + delta_lat))**
xmax, xmin, ymax, ymin, alpha, color, fill, linetype, size

Two Variables

Continuous X, Continuous Y

`f <- ggplot(mpg, aes(cty, hwy))`

-  **f + geom_blank()**
(Useful for expanding limits)
-  **f + geom_jitter()**
x, y, alpha, color, fill, shape, size
-  **f + geom_point()**
x, y, alpha, color, fill, shape, size
-  **f + geom_quantile()**
x, y, alpha, color, linetype, size, weight
-  **f + geom_rug(sides = "bl")**
alpha, color, linetype, size
-  **f + geom_smooth(method = lm)**
x, y, alpha, color, fill, linetype, size, weight
-  **f + geom_text(aes(label = cty))**
x, y, label, alpha, angle, color, family, fontface, hjust, lineheight, size, vjust

Discrete X, Continuous Y

`g <- ggplot(mpg, aes(class, hwy))`

-  **g + geom_bar(stat = "identity")**
x, y, alpha, color, fill, linetype, size, weight
-  **g + geom_boxplot()**
lower, middle, upper, x, ymax, ymin, alpha, color, fill, linetype, shape, size, weight
-  **g + geom_dotplot(binaxis = "y", stackdir = "center")**
x, y, alpha, color, fill
-  **g + geom_violin(scale = "area")**
x, y, alpha, color, fill, linetype, size, weight




Discrete X, Discrete Y

`h <- ggplot(diamonds, aes(cut, color))`

-  **h + geom_jitter()**
x, y, alpha, color, fill, shape, size




Continuous Bivariate Distribution

`i <- ggplot(movies, aes(year, rating))`

-  **i + geom_bin2d(binwidth = c(5, 0.5))**
xmax, xmin, ymax, ymin, alpha, color, fill, linetype, size, weight
-  **i + geom_density2d()**
x, y, alpha, colour, linetype, size
-  **i + geom_hex()**
x, y, alpha, colour, fill size





Continuous Function

`j <- ggplot(economics, aes(date, unemploy))`

-  **j + geom_area()**
x, y, alpha, color, fill, linetype, size
-  **j + geom_line()**
x, y, alpha, color, linetype, size
-  **j + geom_step(direction = "hv")**
x, y, alpha, color, linetype, size

Visualizing error

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



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


Maps

`data <- data.frame(murder = USArrests$Murder, state = tolower(rownames(USArrests)))`
`map <- map_data("state")`
`l <- ggplot(data, aes(fill = murder))`
l + 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

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

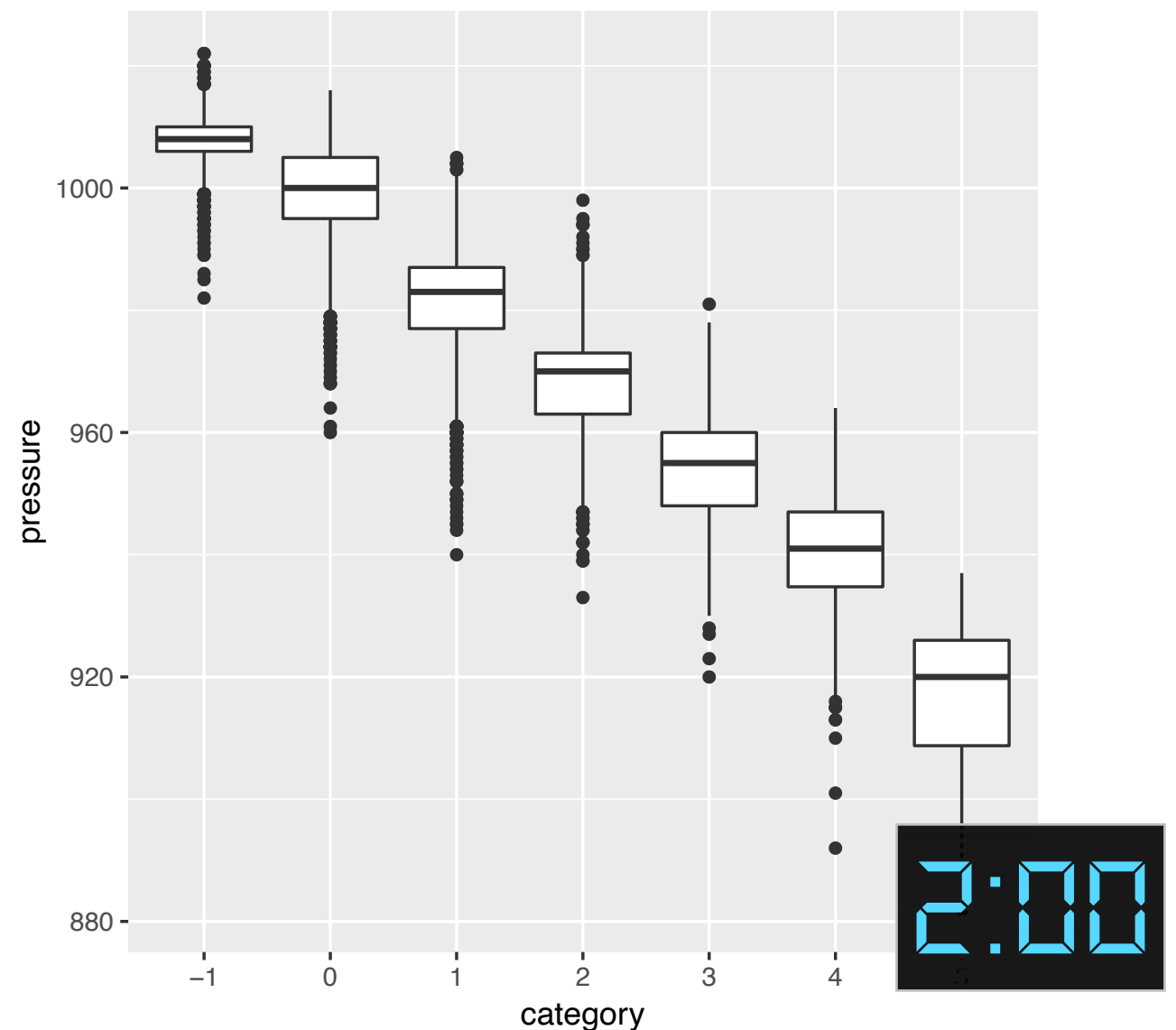
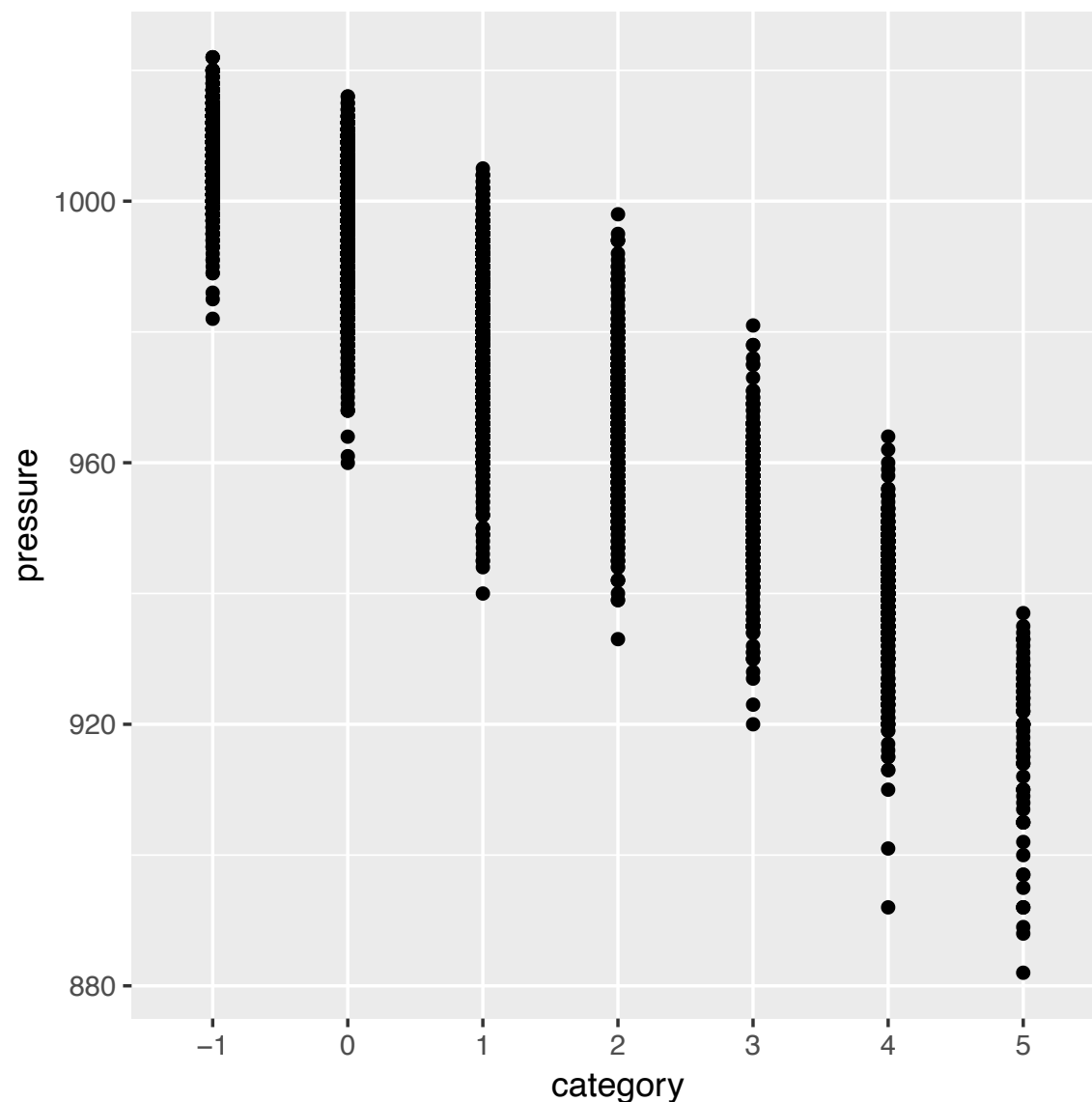
-  **m + geom_raster(aes(fill = z), hjust = 0.5, vjust = 0.5, interpolate = FALSE)**
x, y, alpha, fill (fast)
-  **m + geom_contour(aes(z = z))**
x, y, z, alpha, colour, linetype, size, weight

-  **m + geom_tile(aes(fill = z))**
x, y, alpha, color, fill, linetype, size (slow)

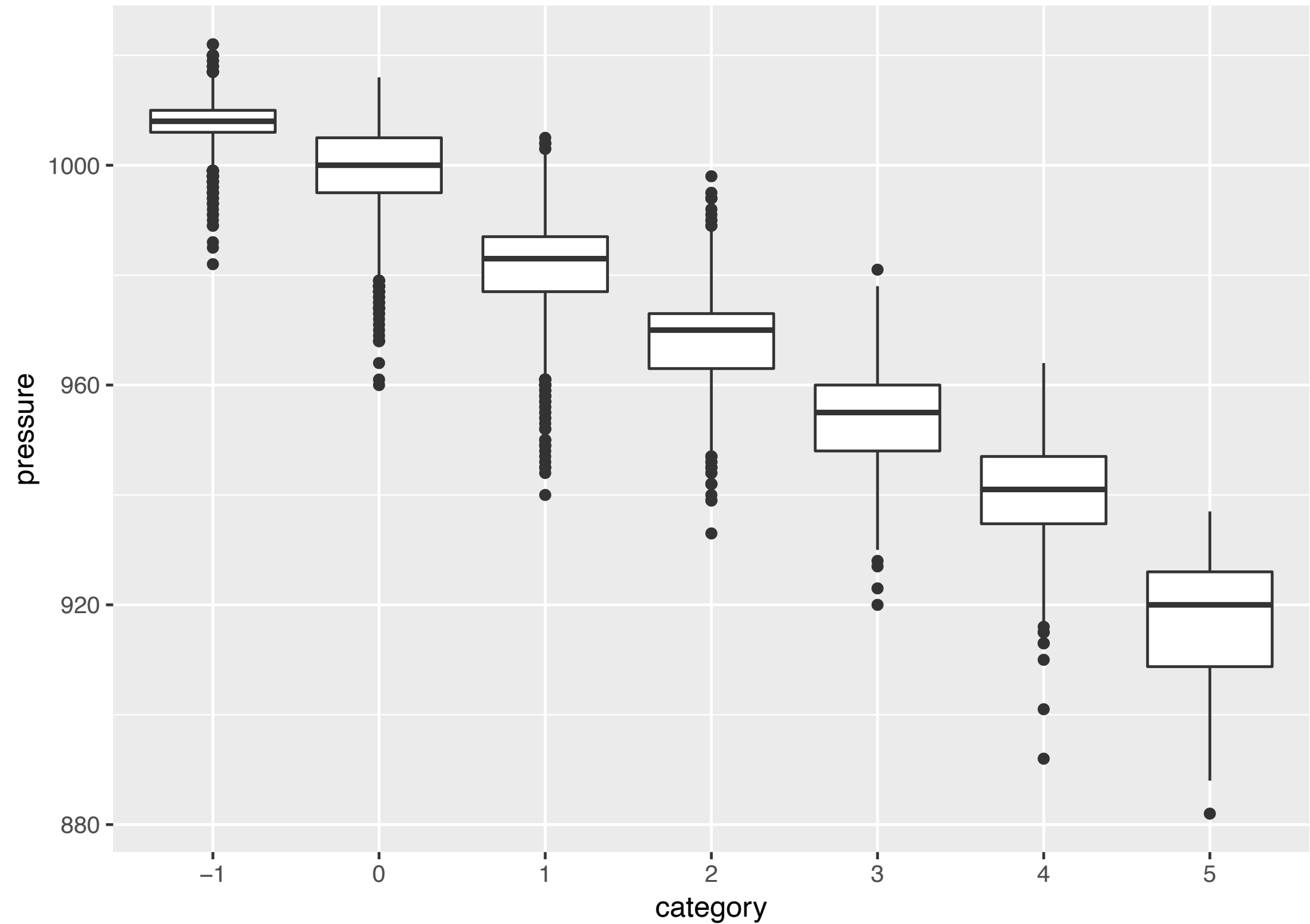
YOUR TURN

- Decide how to replace the scatterplot with boxplots. Use the cheatsheet and try your best guess.

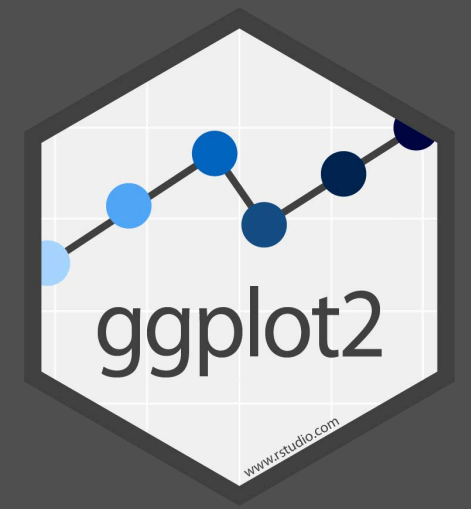
```
ggplot(storms) + geom_point(aes(x = category, y = pressure))
```




```
ggplot(storms) + geom_boxplot(aes(x = category, y = pressure))
```



YOUR TURN

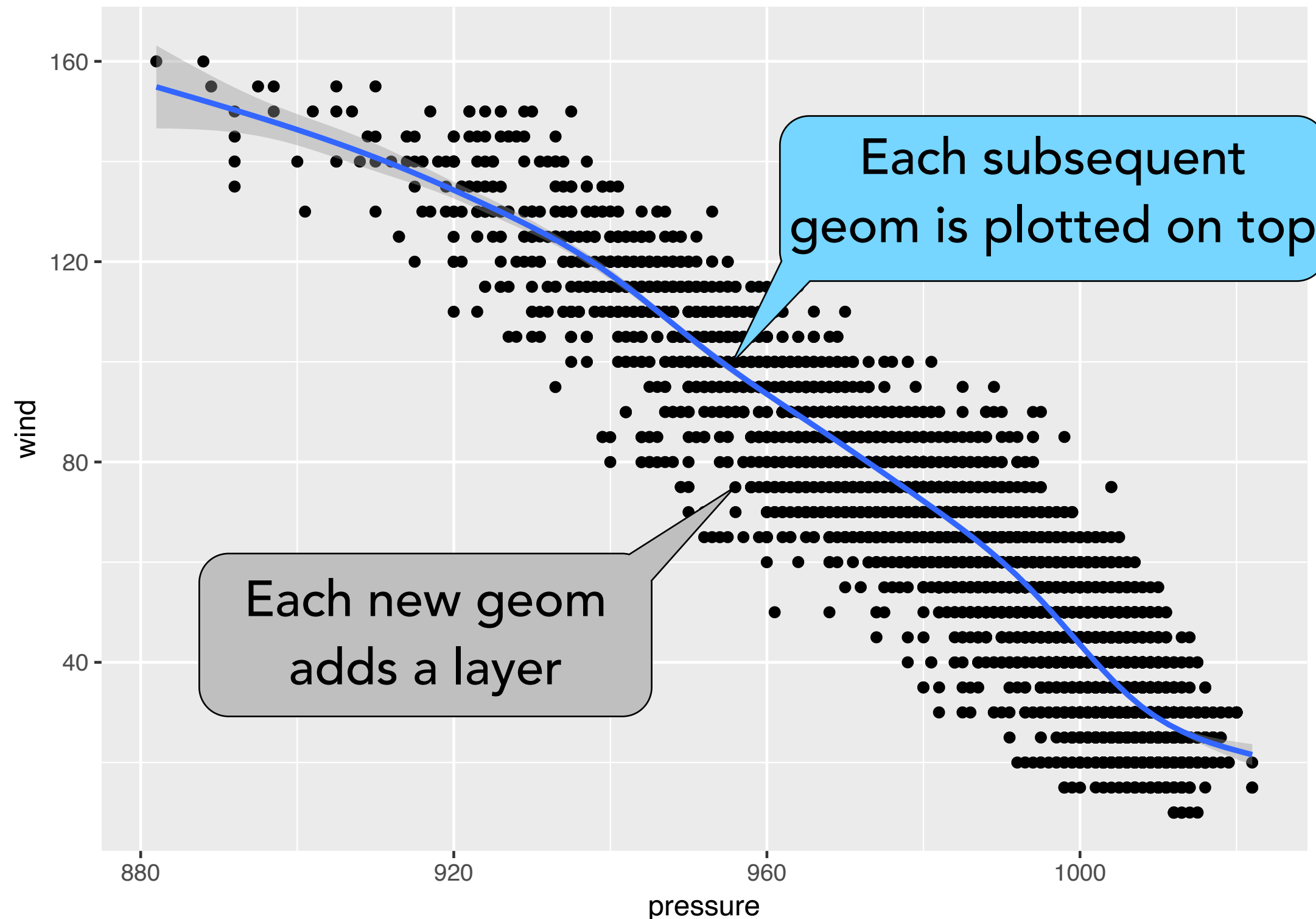


- Predict what this code will do, then run it.

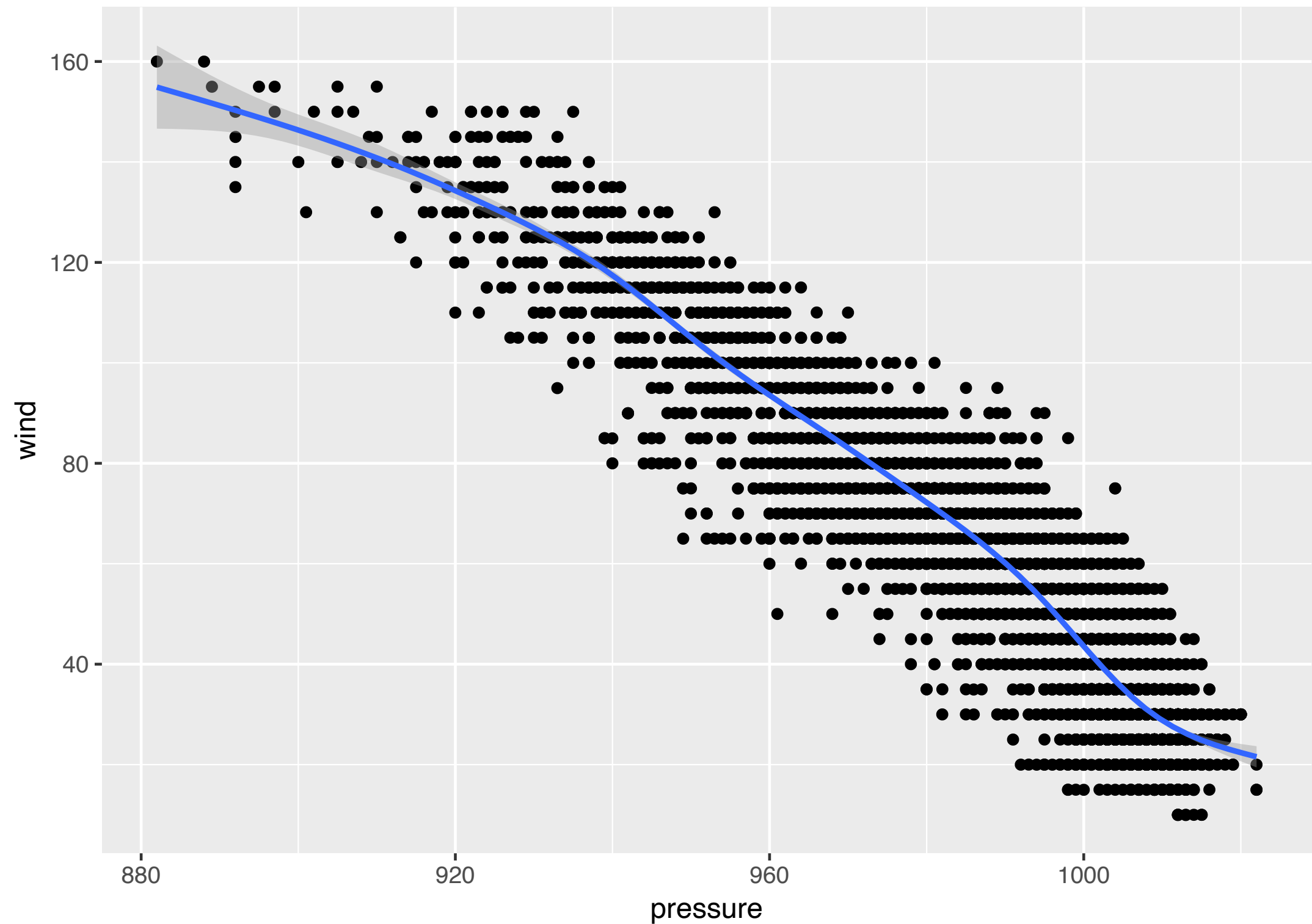
```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind)) +  
  geom_smooth(aes(x = pressure, y = wind))
```

2:00

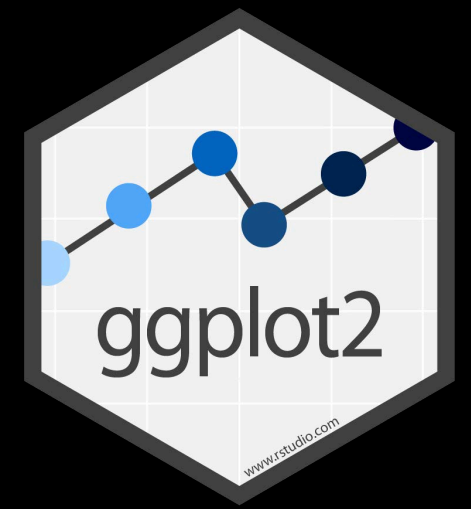
```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind)) +  
  geom_smooth(aes(x = pressure, y = wind))
```



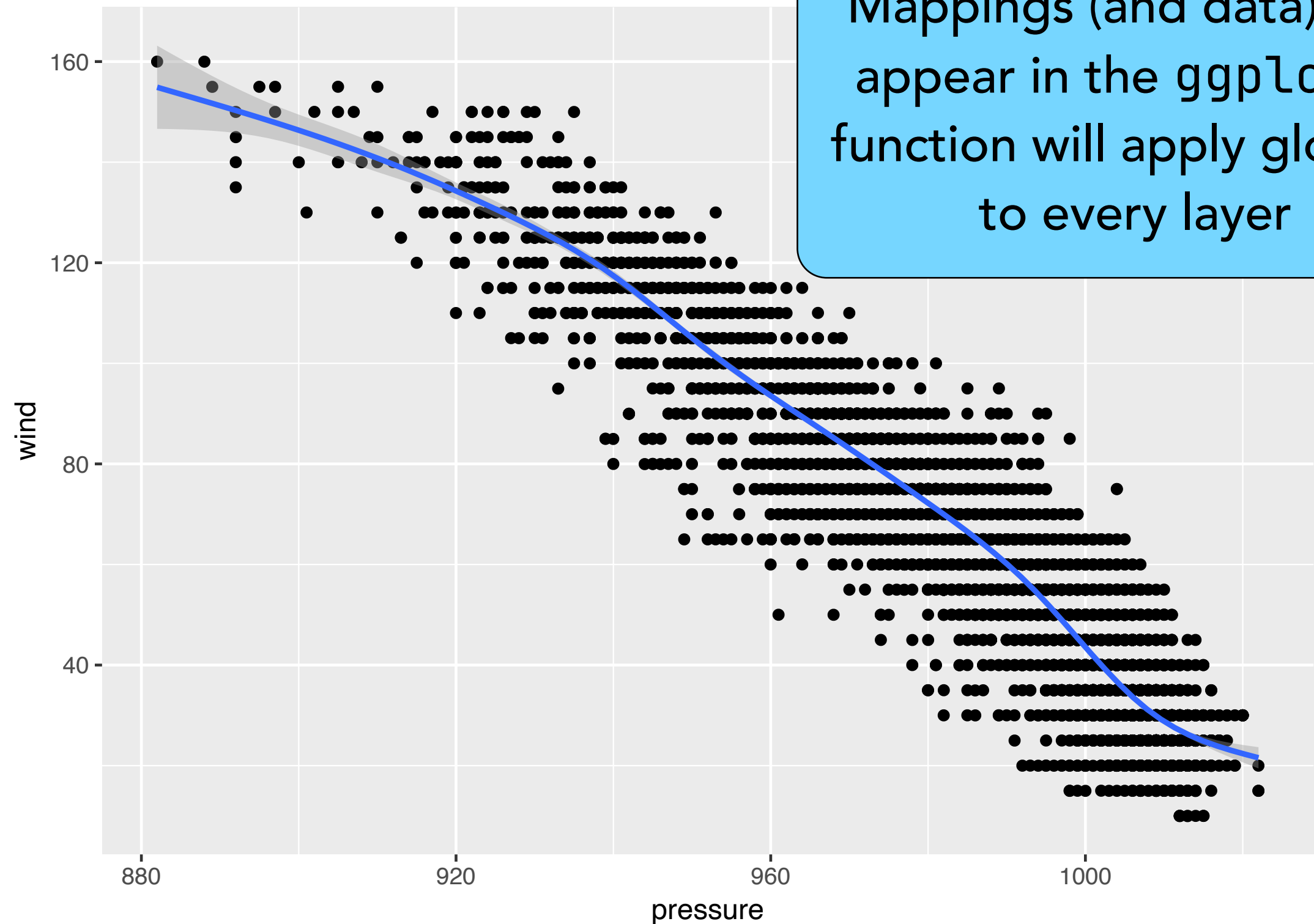
```
ggplot(storms) +  
  geom_point(aes(x = pressure, y = wind)) +  
  geom_smooth(aes(x = pressure, y = wind))
```



GLOBAL VS. LOCAL MAPPING

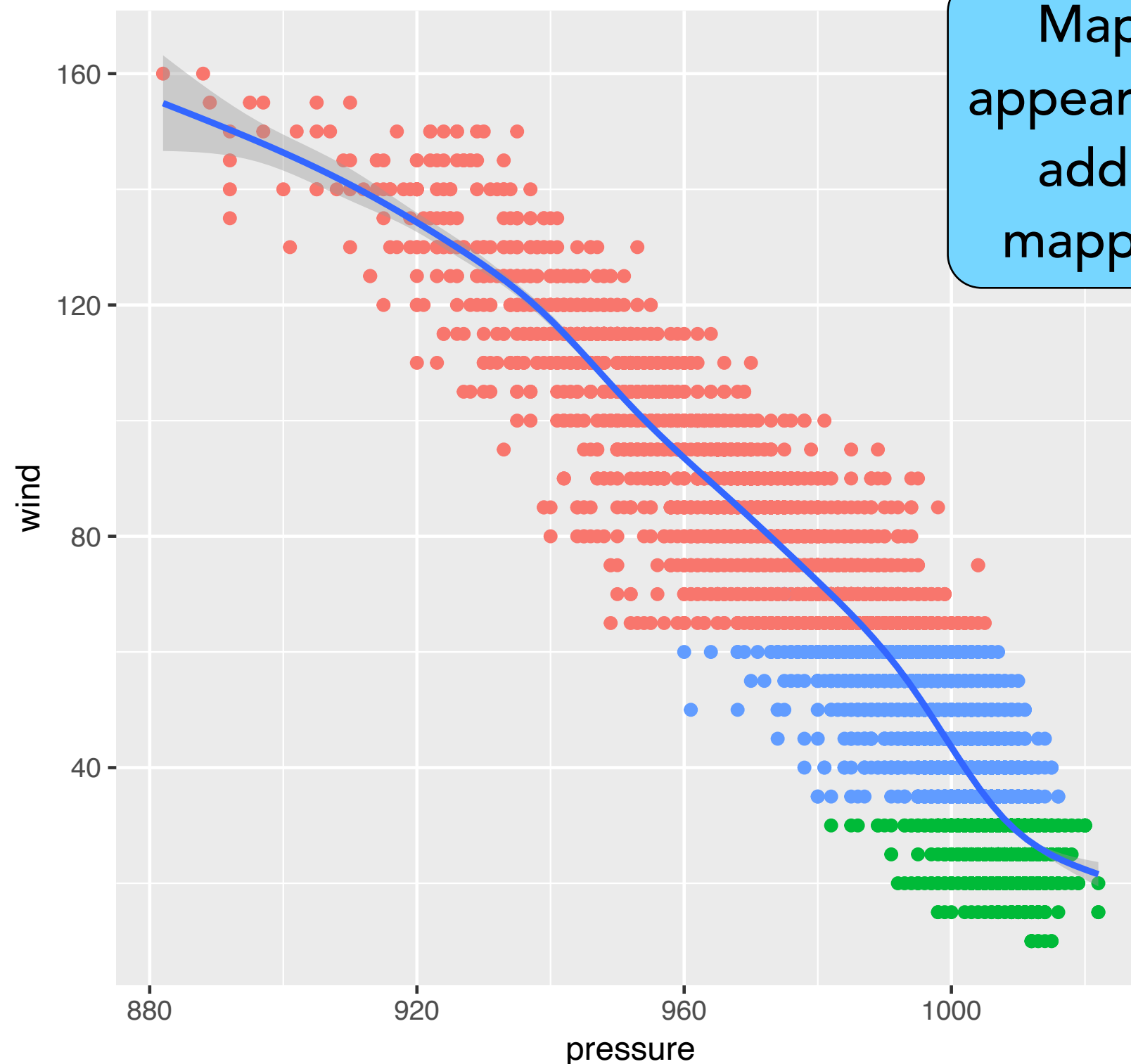


```
ggplot(storms, aes(x = pressure, y = wind)) +  
  geom_point() +  
  geom_smooth()
```



Mappings (and data) that appear in the `ggplot()` function will apply globally to every layer

```
ggplot(storms, aes(x = pressure, y = wind)) +  
  geom_point(aes(color = status)) +  
  geom_smooth()
```

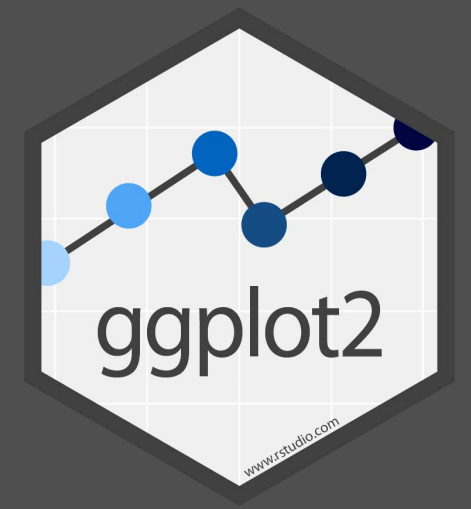


Mappings (and data) that appear in a `geom_` function will add to or override global mappings *for that layer only*

status

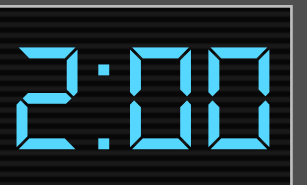
- hurricane
- tropical depression
- tropical storm

YOUR TURN

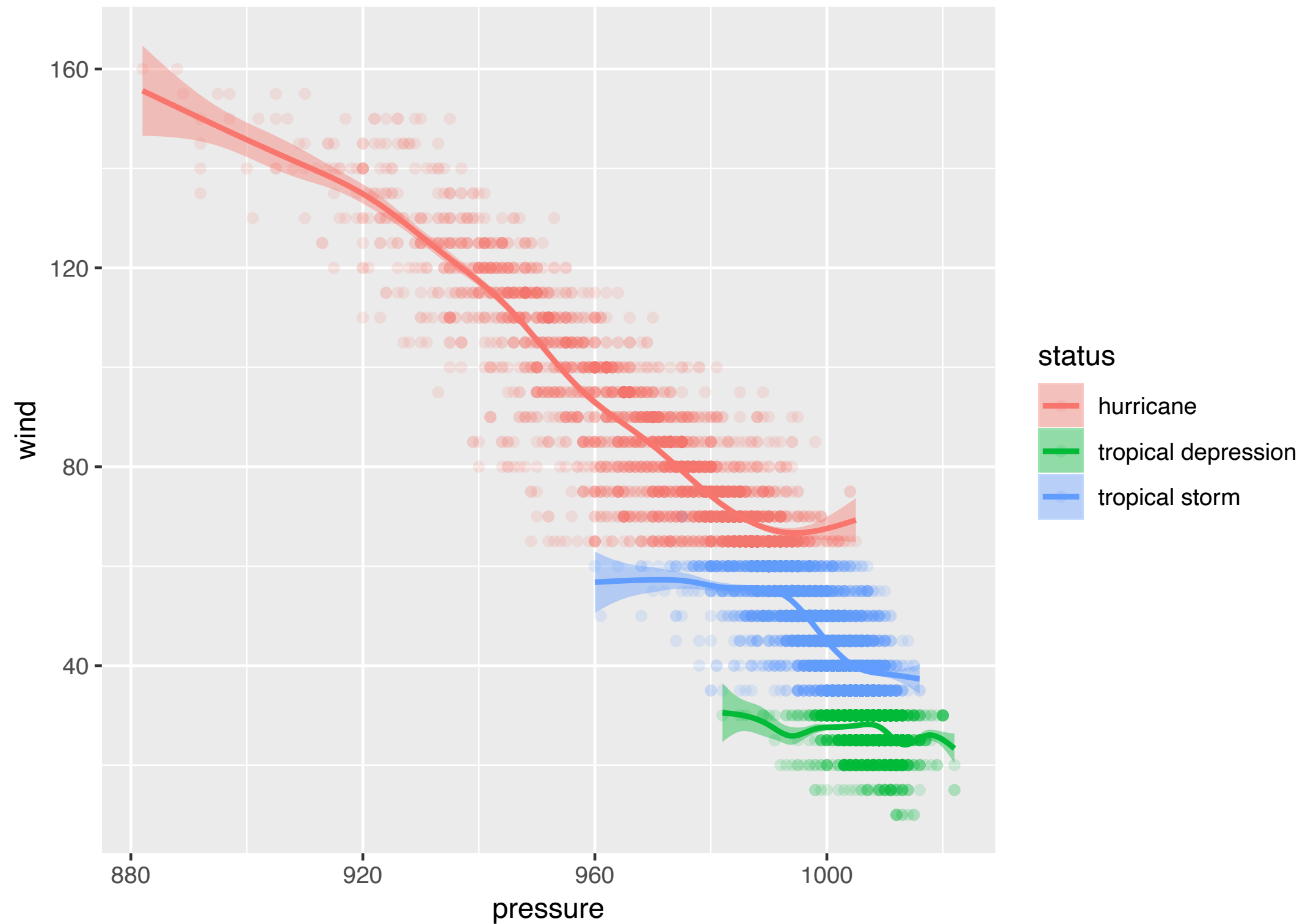


- Predict what this code will do, then run it.

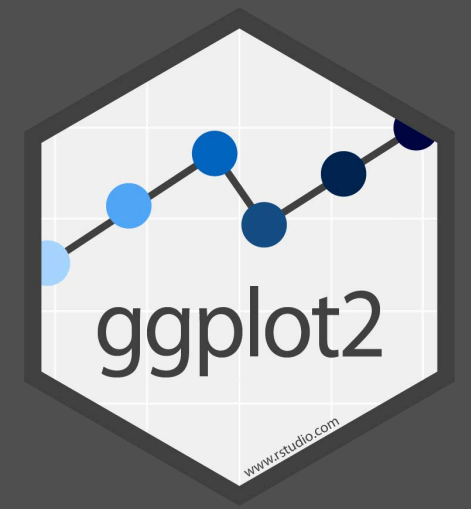
```
ggplot(storms, aes(x = pressure, y = wind,  
                  color = status, fill = status)) +  
  geom_point(alpha = 0.15) +
```



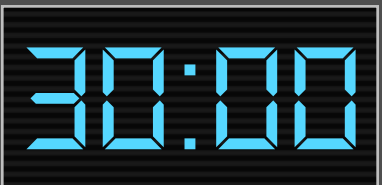

```
ggplot(storms, aes(x = pressure, y = wind,
                  color = status, fill = status)) +
  geom_point(alpha = 0.15) +
```



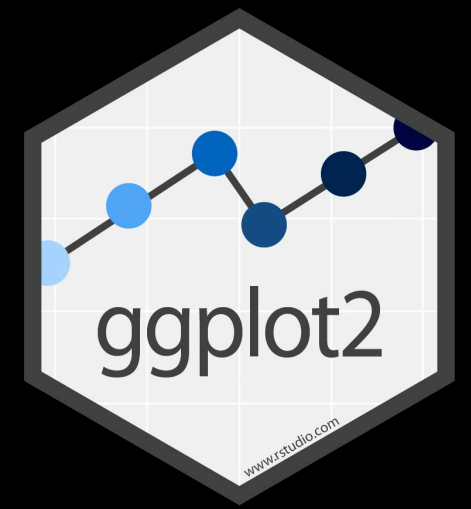
YOUR TURN



- Go to this week's assignments on the course website.
- Download the R Markdown file (save it in this week's folder in your class activities R project).
- Open the R Markdown file in R Studio.
- Follow the instructions to visualize a dataset about mammalian sleep.



WHAT ELSE?



- Stats
- Position adjustments
- Coordinates
- Facets
- Scales
- Themes

```
ggplot(data = <DATA>) +  
  <GEOM_FUNCTION> (  
    mapping = aes(<MAPPINGS>),  
    stat = <STAT> ,  
    position = <POSITION>  
  ) +  
  <COORDINATE_FUNCTION> +  
  <FACET_FUNCTION> +  
  <SCALE_FUNCTION> +  
  <THEME_FUNCTION>
```

Required

Not
required,
sensible
defaults
supplied

Storms data

Status is determined by maximum wind speed

