

Learning `ggplot2` in R

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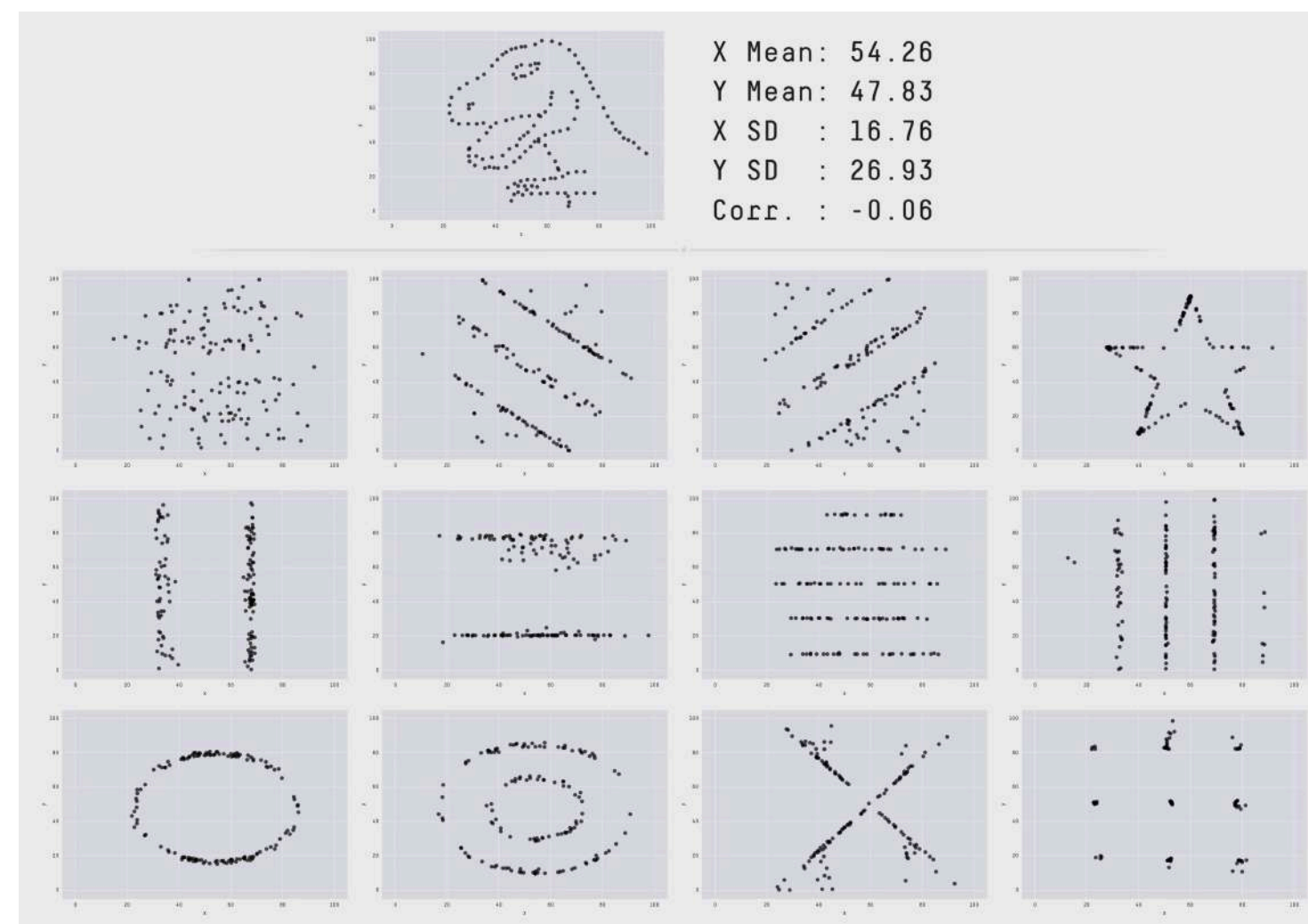
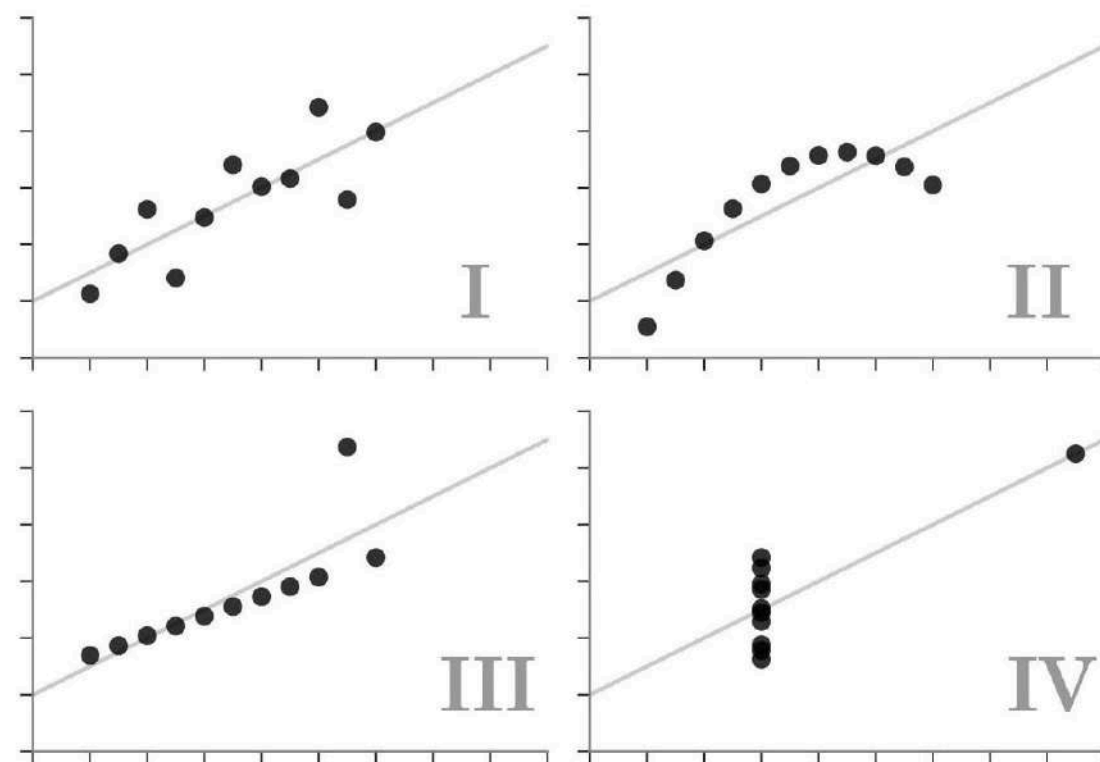
2025-04-07

Graphs

- Essential part of data analyses
- Data with same summary statistics can look very different when plotted out.

Anscombe's Quartet

Each dataset has the same summary statistics (mean, standard deviation, correlation), and the datasets are *clearly different*, and *visually distinct*.

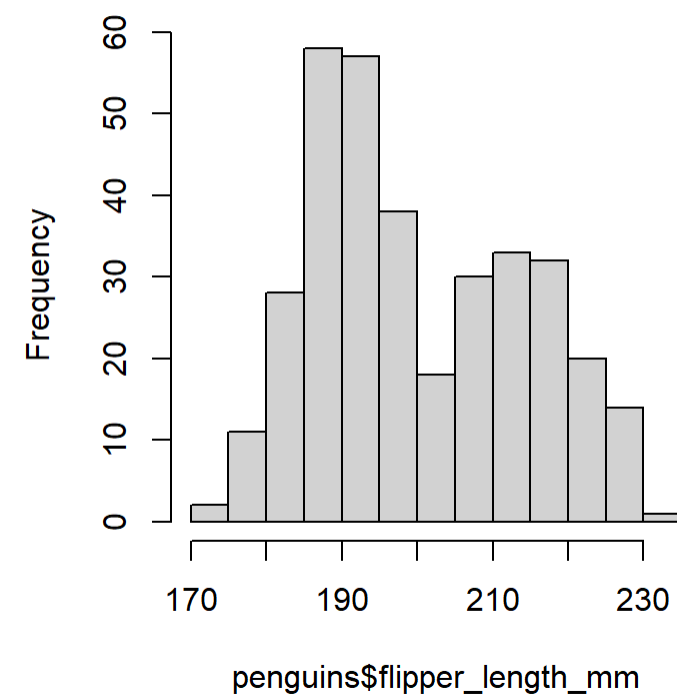


Anscombe's quartet, Datasaurus

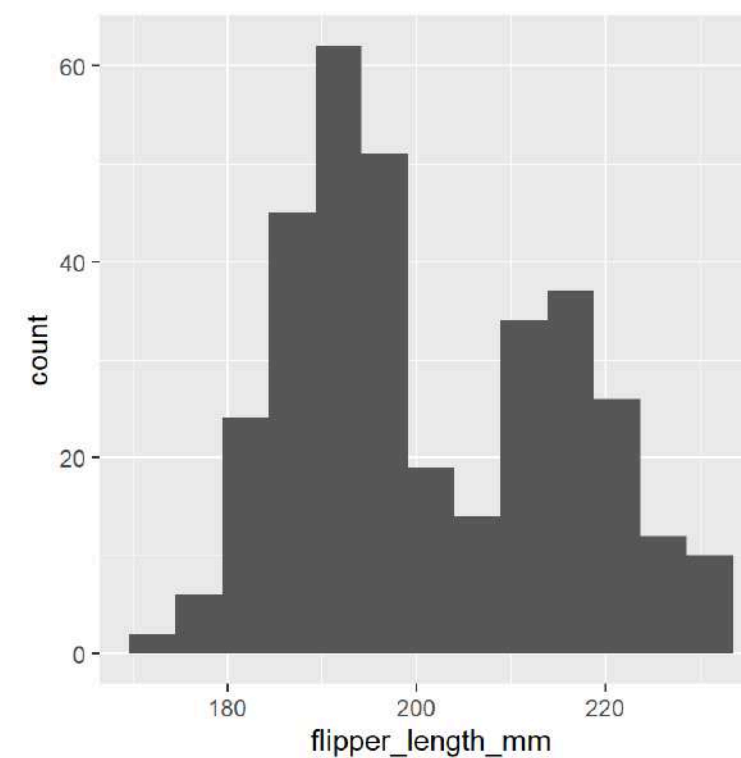
Base Graphics vs ggplot2

```
1 hist(penguins$flipper_length_mm)
```

Histogram of penguins\$flipper_length_mm



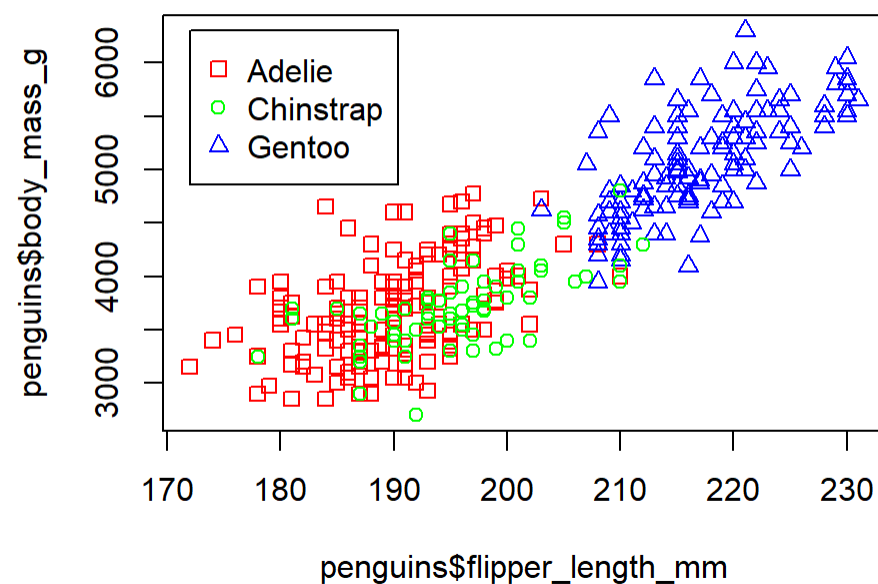
```
1 ggplot(penguins, aes(flipper_length_mm)) +
2   geom_histogram(bins = 13)
```



Base Graphics vs ggplot2

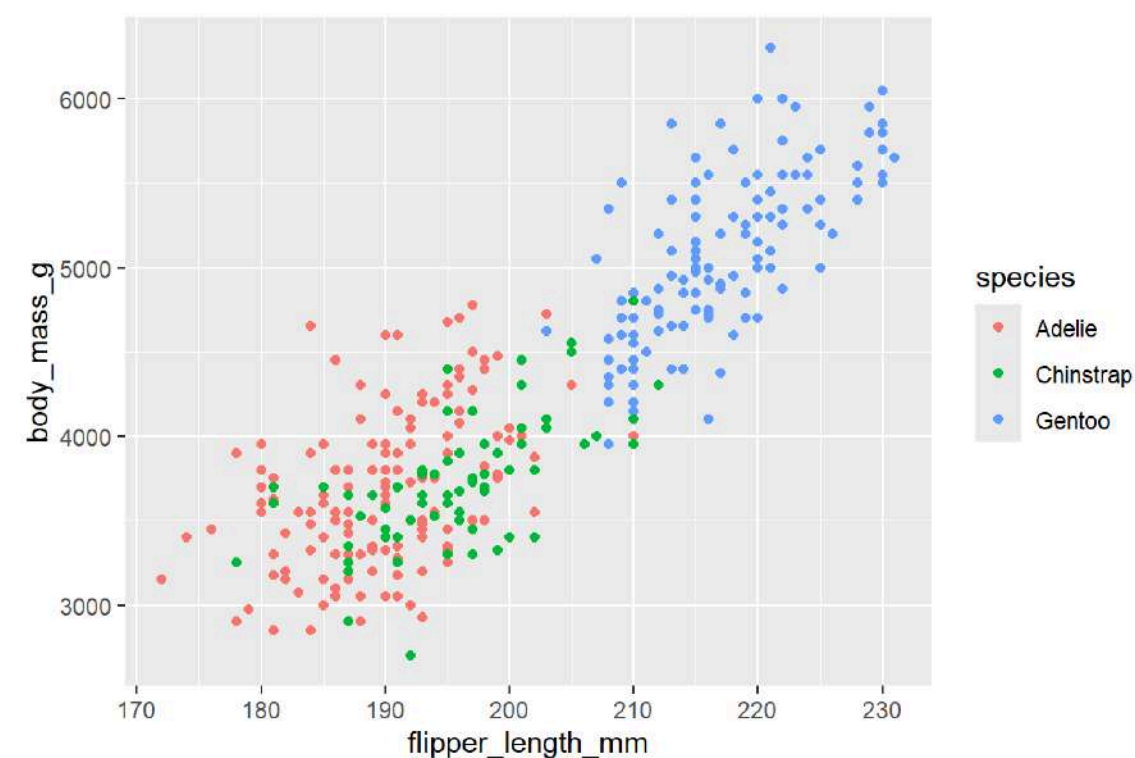
basic r plot

```
1 plot(penguins$flipper_length_mm,penguins$body_mass_g,
2       col=c("red","green","blue")[penguins$species],
3       pch=c(0,1,2)[penguins$species])
4 legend(x=172,y=6300,
5        legend=c("Adelie","Chinstrap","Gentoo"),
6        pch=c(0,1,2),col=c("red","green","blue"))
```



ggplot

```
1 ggplot(penguins, aes(flipper_length_mm,body_mass_g,
2                       color=species)) +
3   geom_point()
```

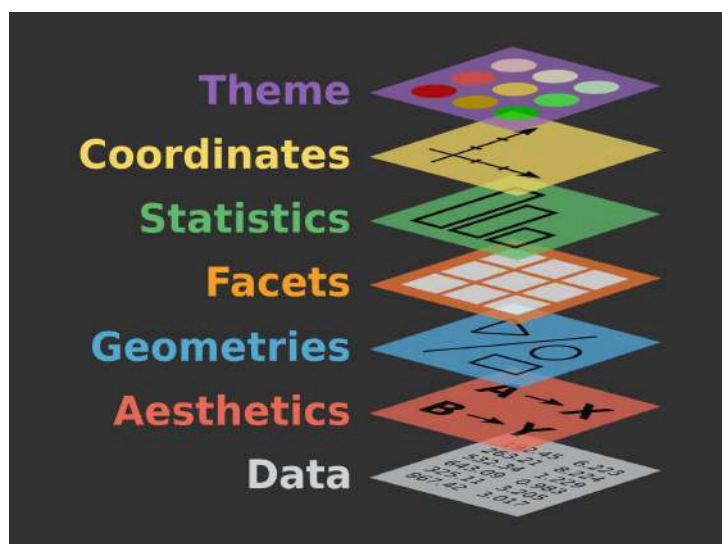


Why `ggplot2`?

- Consistent code
- Flexible (Add/remove components)
- Automatic legends, colors etc
- Save plot objects
- Themes for reusing styles
- Numerous add-ons/extensions
- Nearly complete structured graphing solution
- Adapted to other programming languages
 - `Gadfly` in Julia, `gramm` in MatLab, `GGPlot` in Perl, `Vega` in Javascript, `PlotNine`, `ggpy`, `lets-plot` in Python.

Grammar Of Graphics

Building A Graph: • Syntax



```

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

required

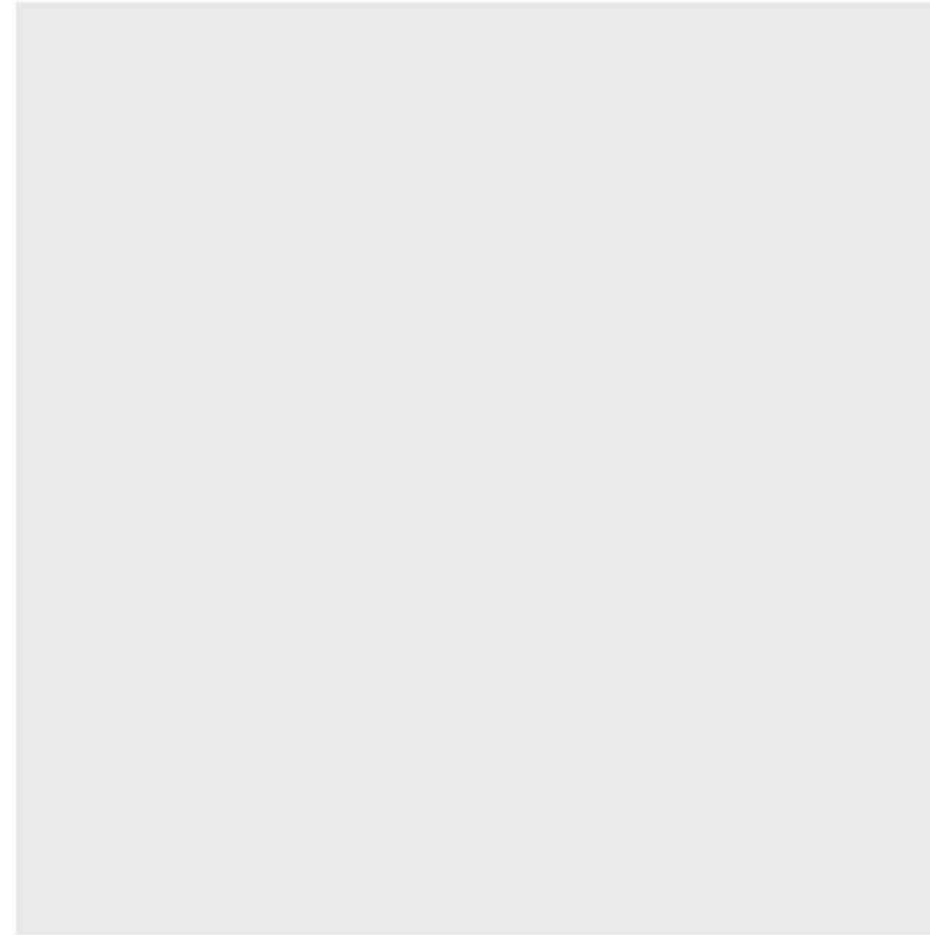
Not required, sensible defaults supplied

Building A Graph

```
1 require(ggplot2)           # load ggplot2
2 require(palmerpenguins)    # load penguins data pack
3
4 data("penguins")           # load penguins data
```

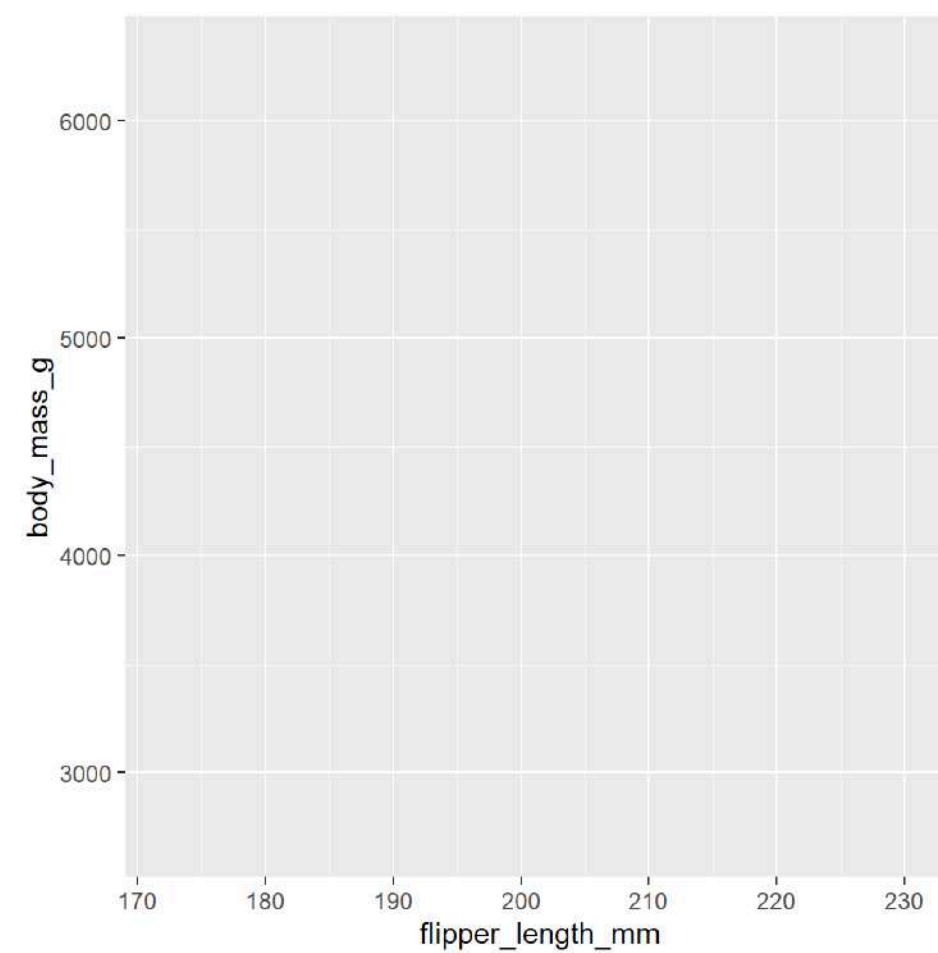

Building A Graph

```
1 ggplot(data = penguins)
```



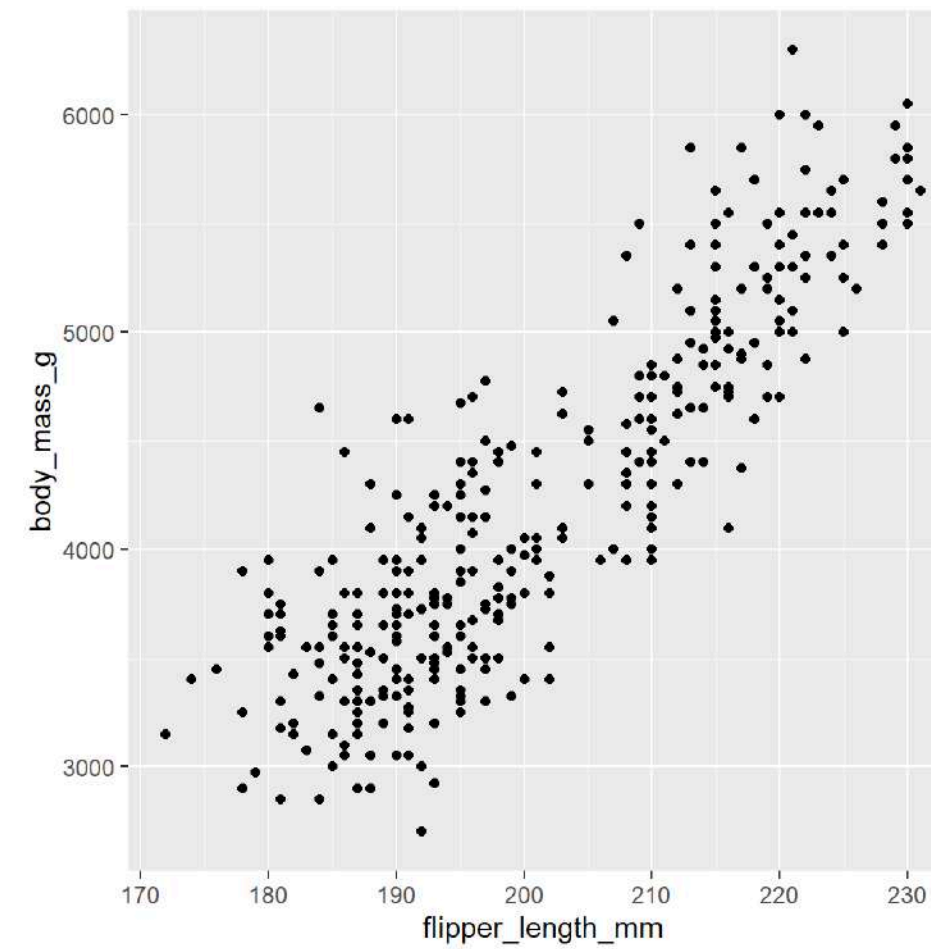
Building A Graph

```
1 ggplot(data = penguins,  
2 mapping = aes(x = flipper_length_mm,  
3               y = body_mass_g))
```



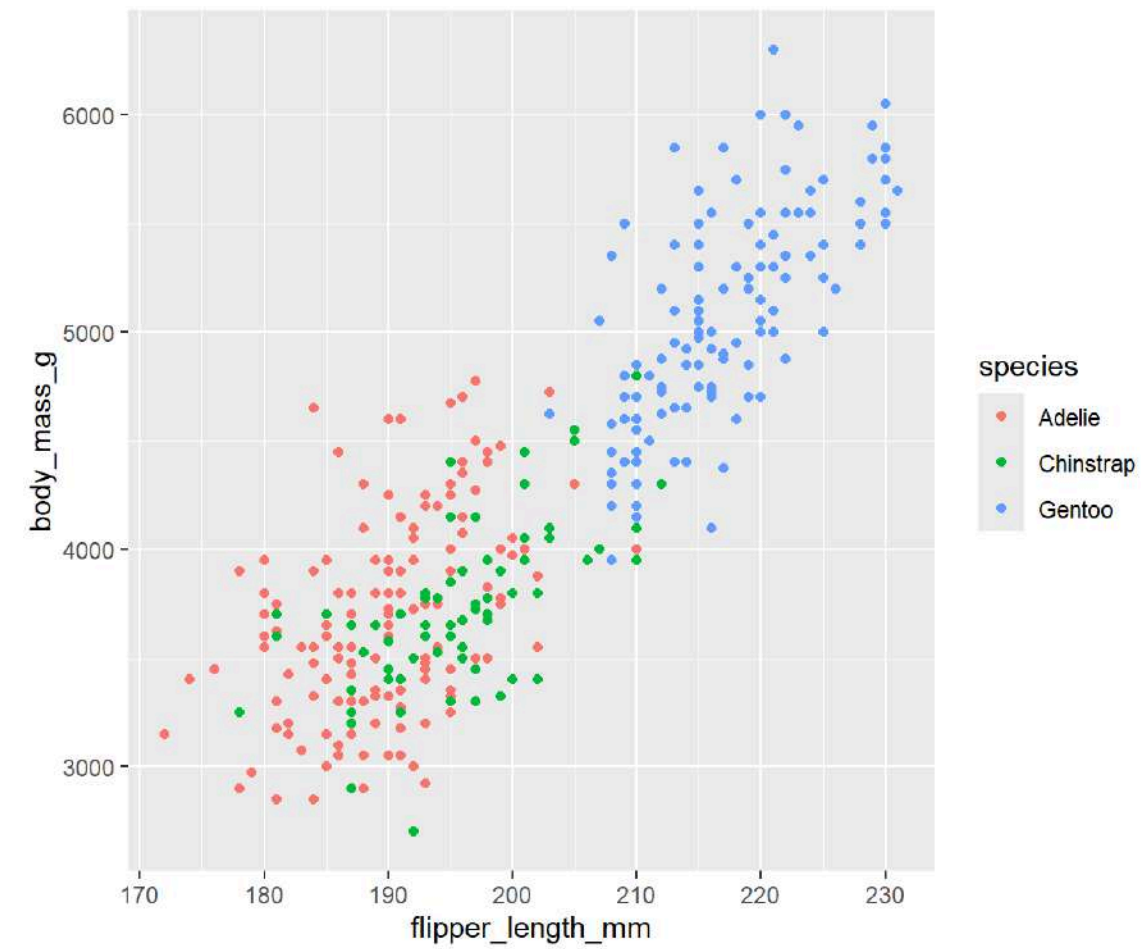
Building A Graph

```
1 ggplot(data = penguins,  
2 mapping = aes(x = flipper_length_mm,  
3                y = body_mass_g)) +  
4 geom_point()
```



Building A Graph

```
1 ggplot(data = penguins,  
2 mapping = aes(x = flipper_length_mm,  
3               y = body_mass_g,  
4               colour = species)) +  
5 geom_point()
```

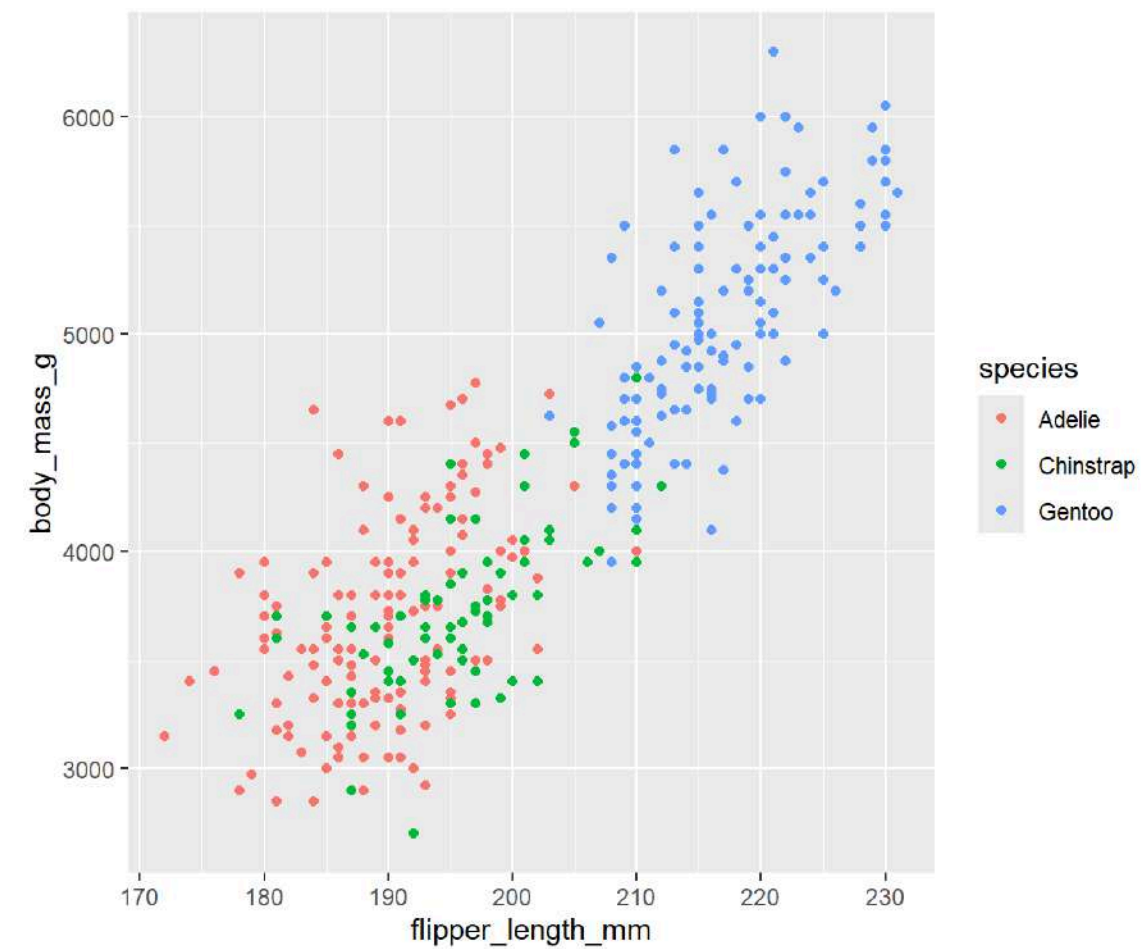


Building A Graph

```
1 ggplot(data = penguins,  
2 mapping = aes(x = flipper_length_mm,  
3               y = body_mass_g,  
4               colour = species)) +  
5 geom_point()
```

Or

```
1 ggplot(data = penguins) +  
2 geom_point(mapping = aes(x = flipper_length_mm,  
3                           y = body_mass_g,  
4                           colour = species))
```

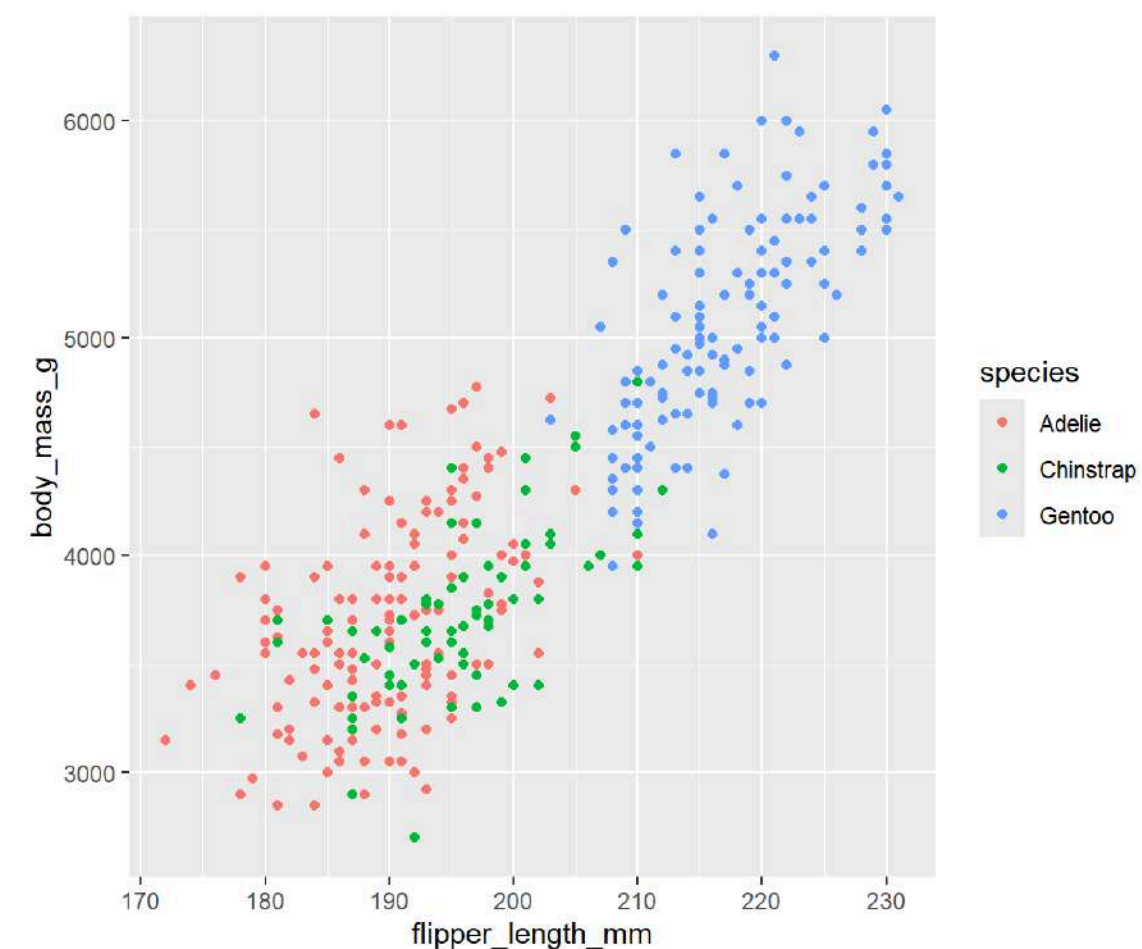


Building A Graph

```
1 ggplot(data = penguins,
2 mapping = aes(x = flipper_length_mm,
3               y = body_mass_g,
4               colour = species)) +
5 geom_point()
```

Or

```
1 ggplot(data = penguins) +
2 geom_point(mapping = aes(x = flipper_length_mm,
3                           y = body_mass_g,
4                           colour = species))
```



ggplot (data = <DATA>) + } required

<GEOM_FUNCTION> (mapping = aes(<MAPPINGS>),

stat = <STAT>, position = <POSITION>) +

<COORDINATE_FUNCTION> +

<FACET_FUNCTION> +

<SCALE_FUNCTION> +

<THEME_FUNCTION> } Not required, sensible defaults supplied

Data • penguins

- Input data is always an R `data.frame` object

species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
Adelie	Torgersen	39.1	18.7	181	3750
Adelie	Torgersen	39.5	17.4	186	3800
Adelie	Torgersen	40.3	18.0	195	3250

```
1 str(penguins)
```

```
tibble [344 × 8] (S3: tbl_df/tbl/data.frame)
 $ species      : Factor w/ 3 levels "Adelie","Chinstrap",...: 1 1 1 1 1 1 1 1 1 1 ...
 $ island       : Factor w/ 3 levels "Biscoe","Dream",...: 3 3 3 3 3 3 3 3 3 3 ...
 $ bill_length_mm : num [1:344] 39.1 39.5 40.3 NA 36.7 39.3 38.9 39.2 34.1 42 ...
 $ bill_depth_mm  : num [1:344] 18.7 17.4 18 NA 19.3 20.6 17.8 19.6 18.1 20.2 ...
 $ flipper_length_mm: int [1:344] 181 186 195 NA 193 190 181 195 193 190 ...
 $ body_mass_g    : int [1:344] 3750 3800 3250 NA 3450 3650 3625 4675 3475 4250 ...
 $ sex           : Factor w/ 2 levels "female","male": 2 1 1 NA 1 2 1 2 NA NA ...
 $ year          : int [1:344] 2007 2007 2007 2007 2007 2007 2007 2007 2007 2007 ...
```

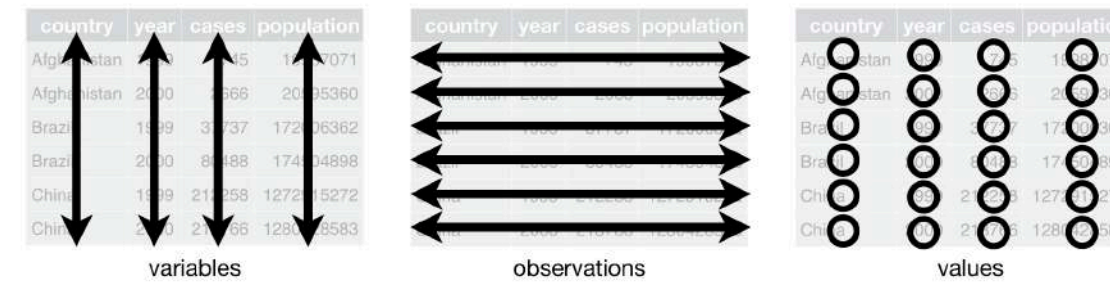
Data • diamonds

carat	cut	color	clarity	depth	table	price	x	y	z
0.23	Ideal	E	SI2	61.5	55	326	3.95	3.98	2.43
0.21	Premium	E	SI1	59.8	61	326	3.89	3.84	2.31
0.23	Good	E	VS1	56.9	65	327	4.05	4.07	2.31
0.29	Premium	I	VS2	62.4	58	334	4.20	4.23	2.63
0.31	Good	J	SI2	63.3	58	335	4.34	4.35	2.75
0.24	Very Good	J	VVS2	62.8	57	336	3.94	3.96	2.48

```
tibble [53,940 × 10] (S3: tbl_df/tbl/data.frame)
 $ carat   : num [1:53940] 0.23 0.21 0.23 0.29 0.31 0.24 0.24 0.26 0.22 0.23 ...
 $ cut     : Ord.factor w/ 5 levels "Fair"<"Good"<...: 5 4 2 4 2 3 3 3 1 3 ...
 $ color   : Ord.factor w/ 7 levels "D"<"E"<"F"<"G"<...: 2 2 2 6 7 7 6 5 2 5 ...
 $ clarity: Ord.factor w/ 8 levels "I1"<"SI2"<"SI1"<...: 2 3 5 4 2 6 7 3 4 5 ...
 $ depth   : num [1:53940] 61.5 59.8 56.9 62.4 63.3 62.8 62.3 61.9 65.1 59.4 ...
 $ table   : num [1:53940] 55 61 65 58 58 57 57 55 61 61 ...
 $ price   : int  [1:53940] 326 326 327 334 335 336 336 337 337 338 ...
 $ x       : num [1:53940] 3.95 3.89 4.05 4.2 4.34 3.94 3.95 4.07 3.87 4 ...
 $ y       : num [1:53940] 3.98 3.84 4.07 4.23 4.35 3.96 3.98 4.11 3.78 4.05 ...
 $ z       : num [1:53940] 2.43 2.31 2.31 2.63 2.75 2.48 2.47 2.53 2.49 2.39 ...
```


Data • format

- Transforming data into 'long' or 'wide' formats



Wide

```
1 head(penguins, n=4)
```

```
# A tibble: 4 × 8
```

```
  species island bill_length_mm bill_depth_mm flipper_length_mm body_mass_g
  <fct>   <fct>         <dbl>         <dbl>         <int>         <int>
1 Adelie Torgersen      39.1           18.7           181           3750
2 Adelie Torgersen      39.5           17.4           186           3800
3 Adelie Torgersen      40.3            18           195           3250
4 Adelie Torgersen      NA             NA             NA             NA
```

```
# i 2 more variables: sex <fct>, year <int>
```

Long

```
  species island sex year variables value
1 Adelie Torgersen male 2007 bill_length_mm 39.1
2 Adelie Torgersen male 2007 bill_depth_mm 18.7
3 Adelie Torgersen male 2007 flipper_length_mm 181.0
4 Adelie Torgersen male 2007 body_mass_g 3750.0
```

Geoms • types

Basic



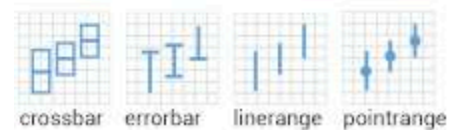
One variable



Two variables



Error



Three variables



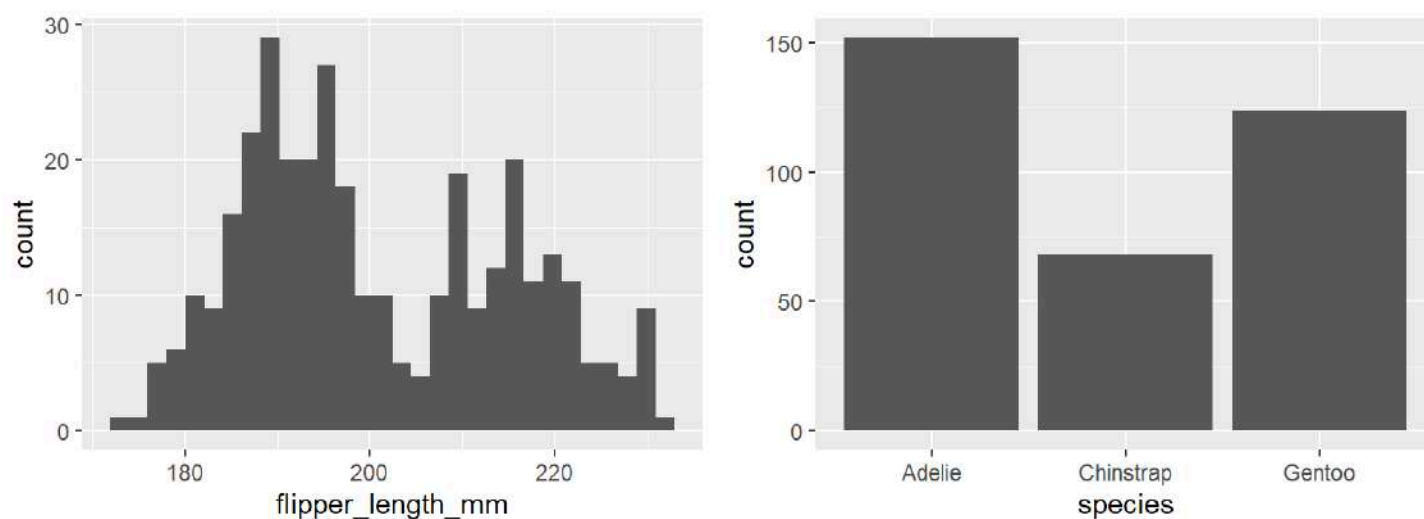
Map



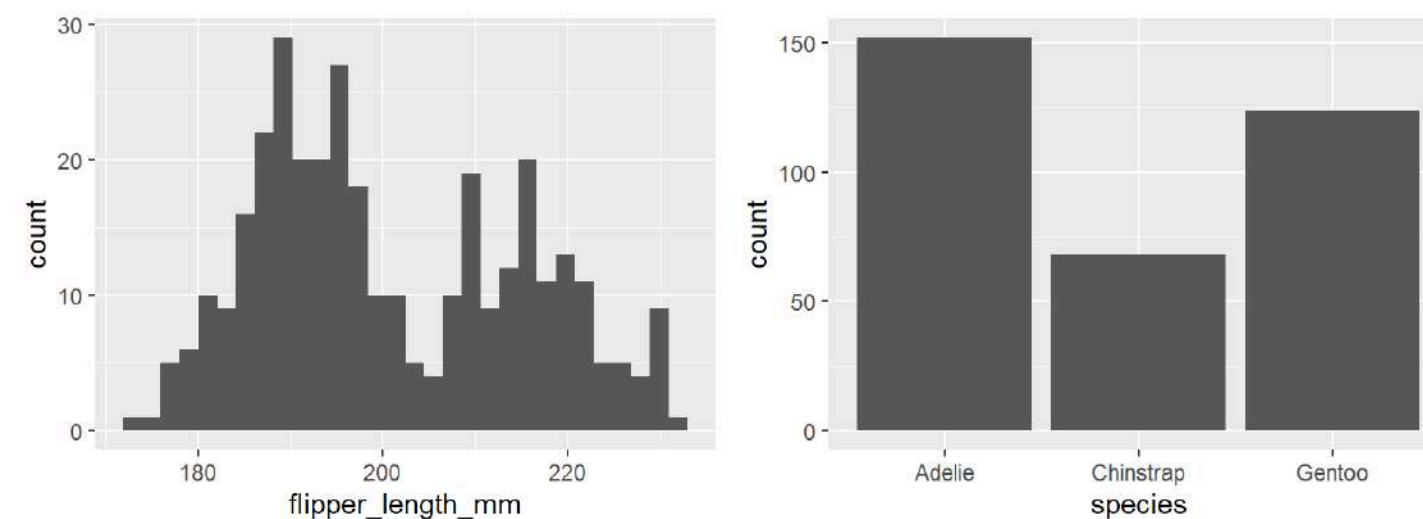
Stats

- **Stats** compute new variables from input data.
- **Geoms** have default stats.
- Plots can be built with stats.

```
1 x <- ggplot(data = penguins) +
2   geom_bar(aes(x=flipper_length_mm), stat="bin")
3
4 y <- ggplot(data = penguins) +
5   geom_bar(aes(x=species), stat="count")
6
7 wrap_plots(x, y, nrow=1)
```



```
1 x <- ggplot(data = penguins) +
2   stat_bin(aes(x=flipper_length_mm), geom="bar")
3
4 y <- ggplot(data = penguins) +
5   stat_count(aes(x=species), geom="bar")
6
7 wrap_plots(x, y, nrow=1)
```



Stats

- Stats have default geoms.

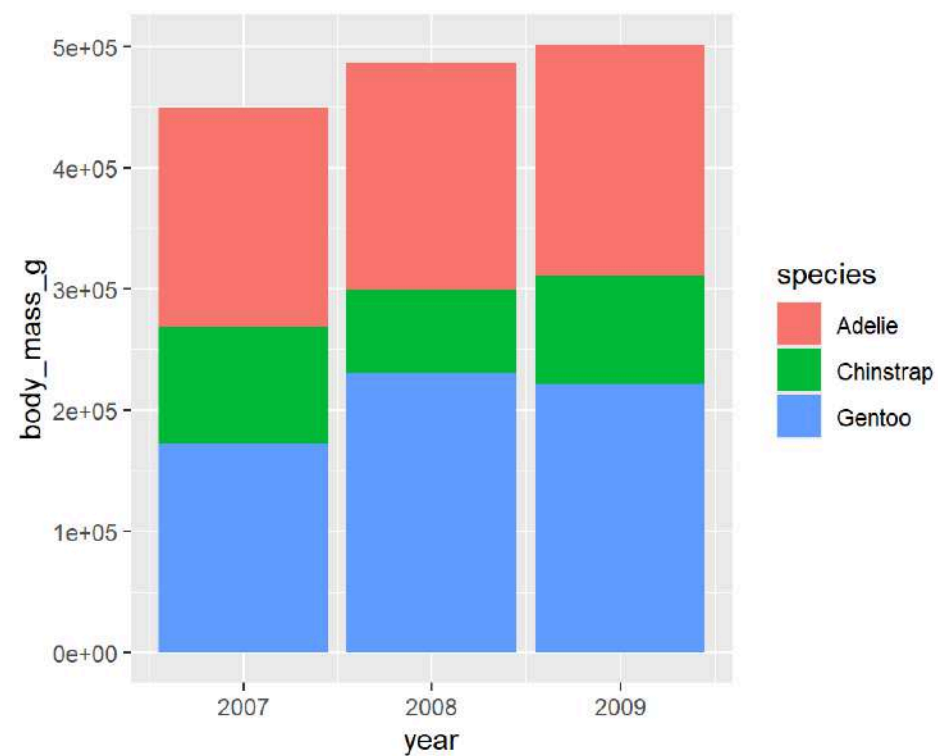
plot	stat	geom
histogram	bin	bar
smooth	smooth	line
boxplot	boxplot	boxplot
density	density	line
freqpoly	freqpoly	line

Use `args(geom_bar)` to check arguments.

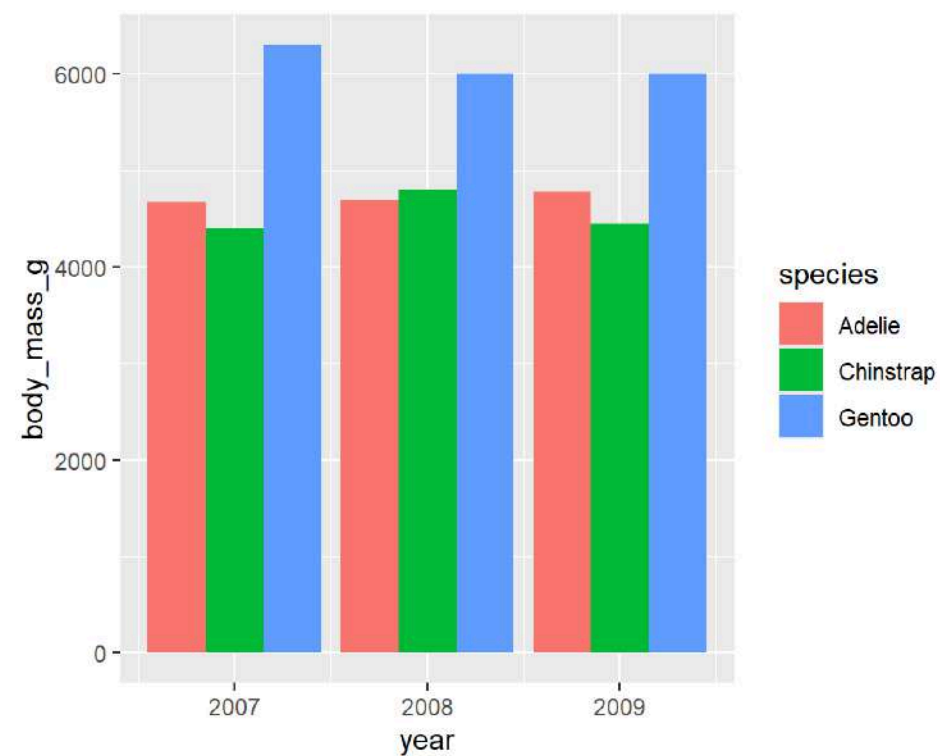
Position

```
1 p <- ggplot(penguins, aes(x=year, y=body_mass_g, fill=species))
```

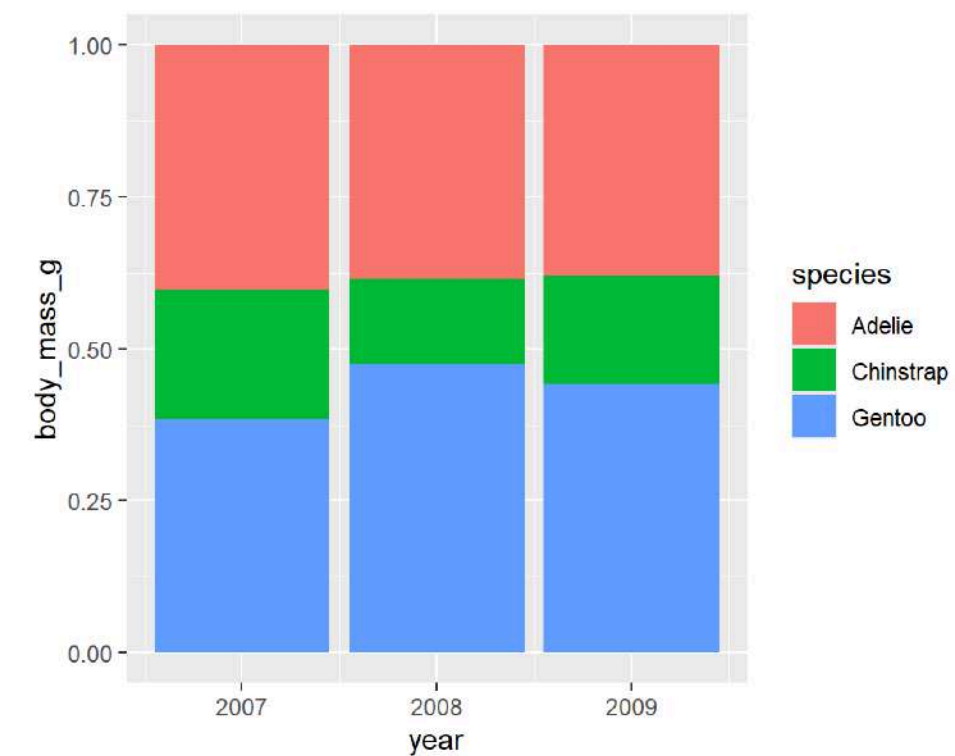
```
1 p + geom_bar(stat="identity",  
2             position="stack")
```



```
1 p + geom_bar(stat="identity",  
2             position="dodge")
```



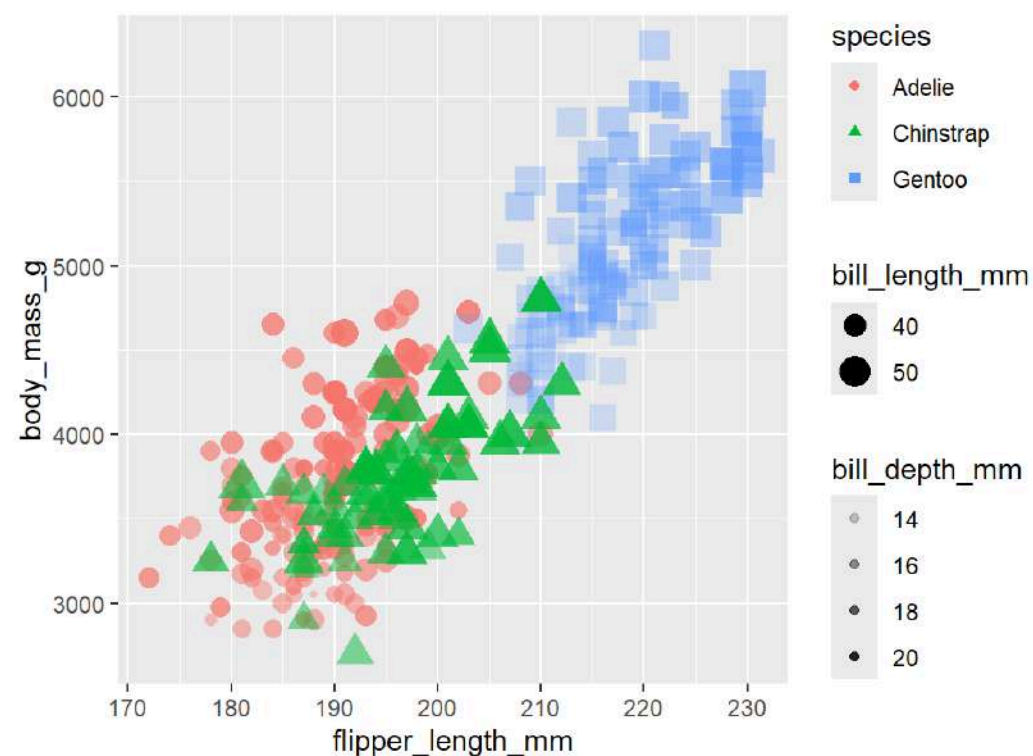
```
1 p + geom_bar(stat="identity",  
2             position="fill")
```



Aesthetics

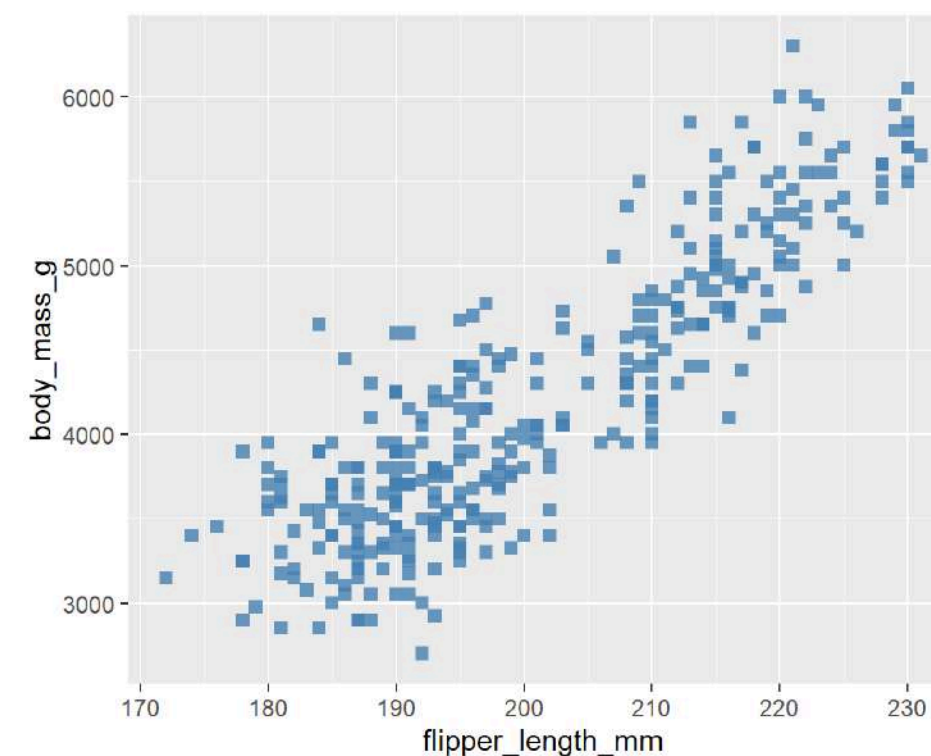
- Aesthetic **mapping**

```
1 ggplot(data = penguins)+
2   geom_point(aes(x=flipper_length_mm,
3                 y=body_mass_g,
4                 size=bill_length_mm,
5                 alpha=bill_depth_mm,
6                 shape=species,
7                 color=species))
```



- Aesthetic **parameter**

```
1 ggplot(data = penguins)+
2   geom_point(aes(x=flipper_length_mm,
3                 y=body_mass_g),
4             size=2,
5             alpha=0.8,
6             shape=15,
7             color="steelblue")
```

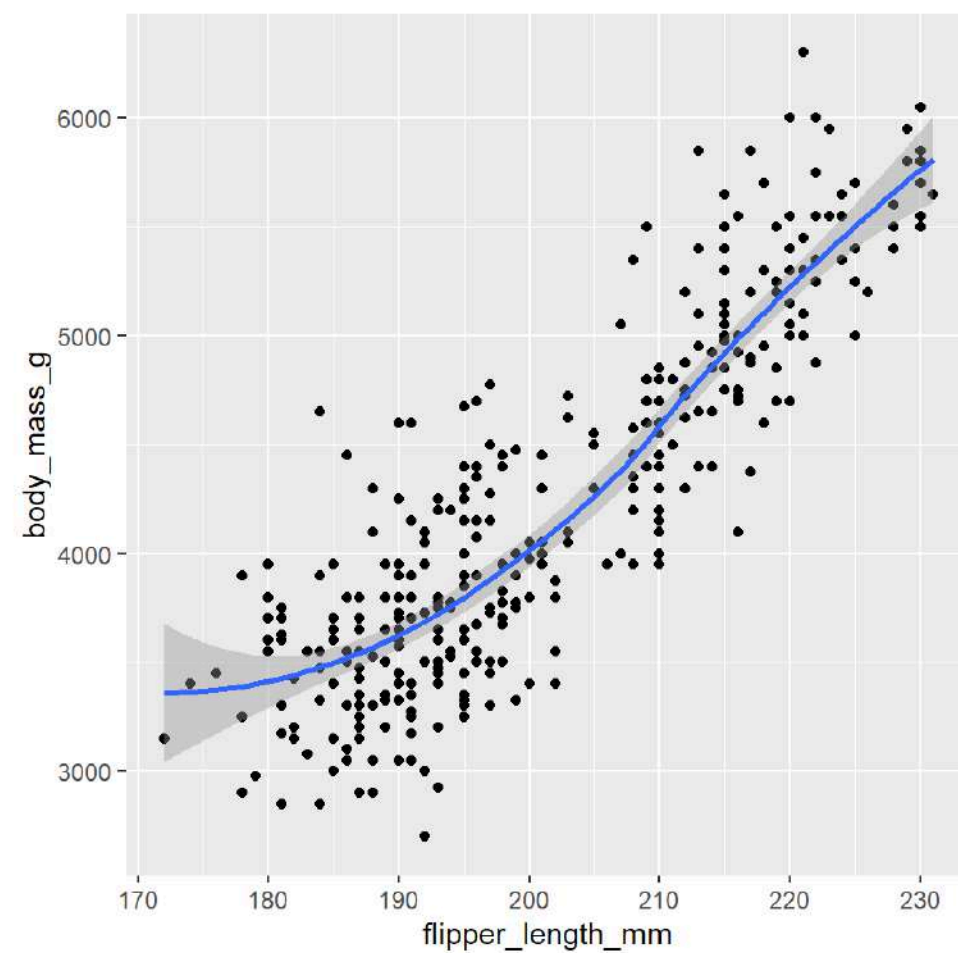
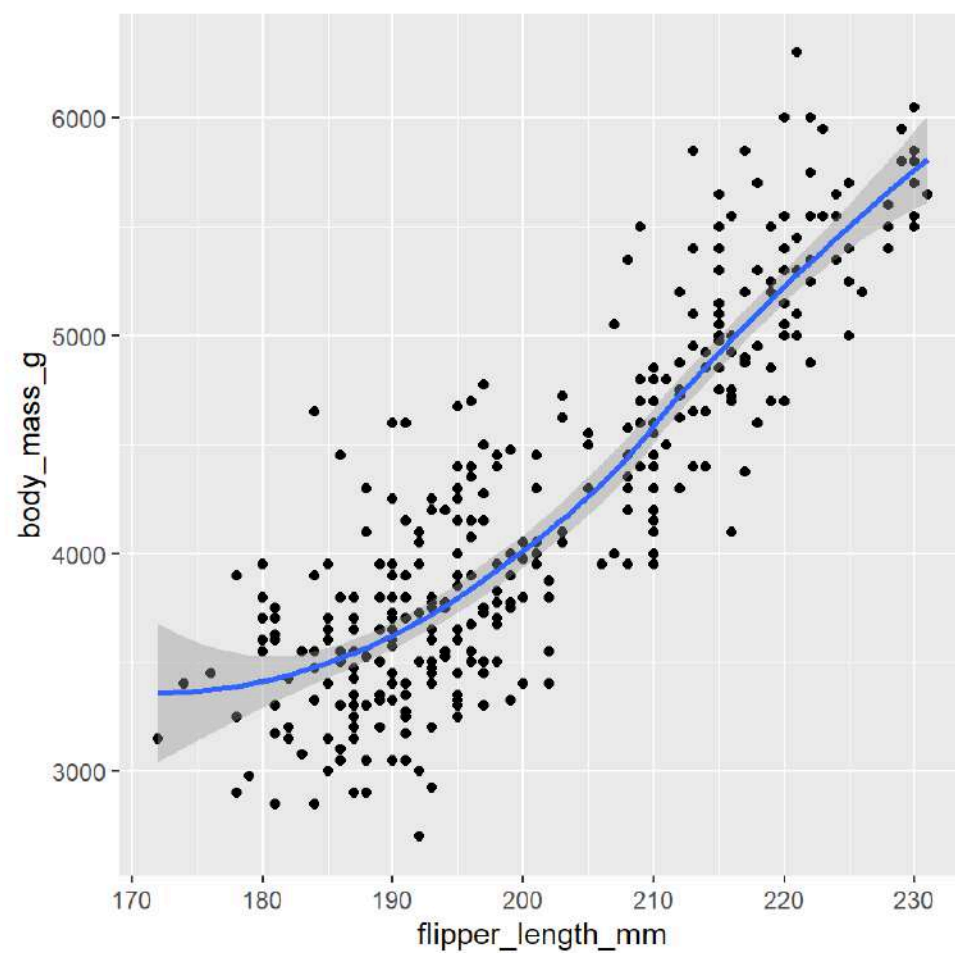


Aesthetics

```

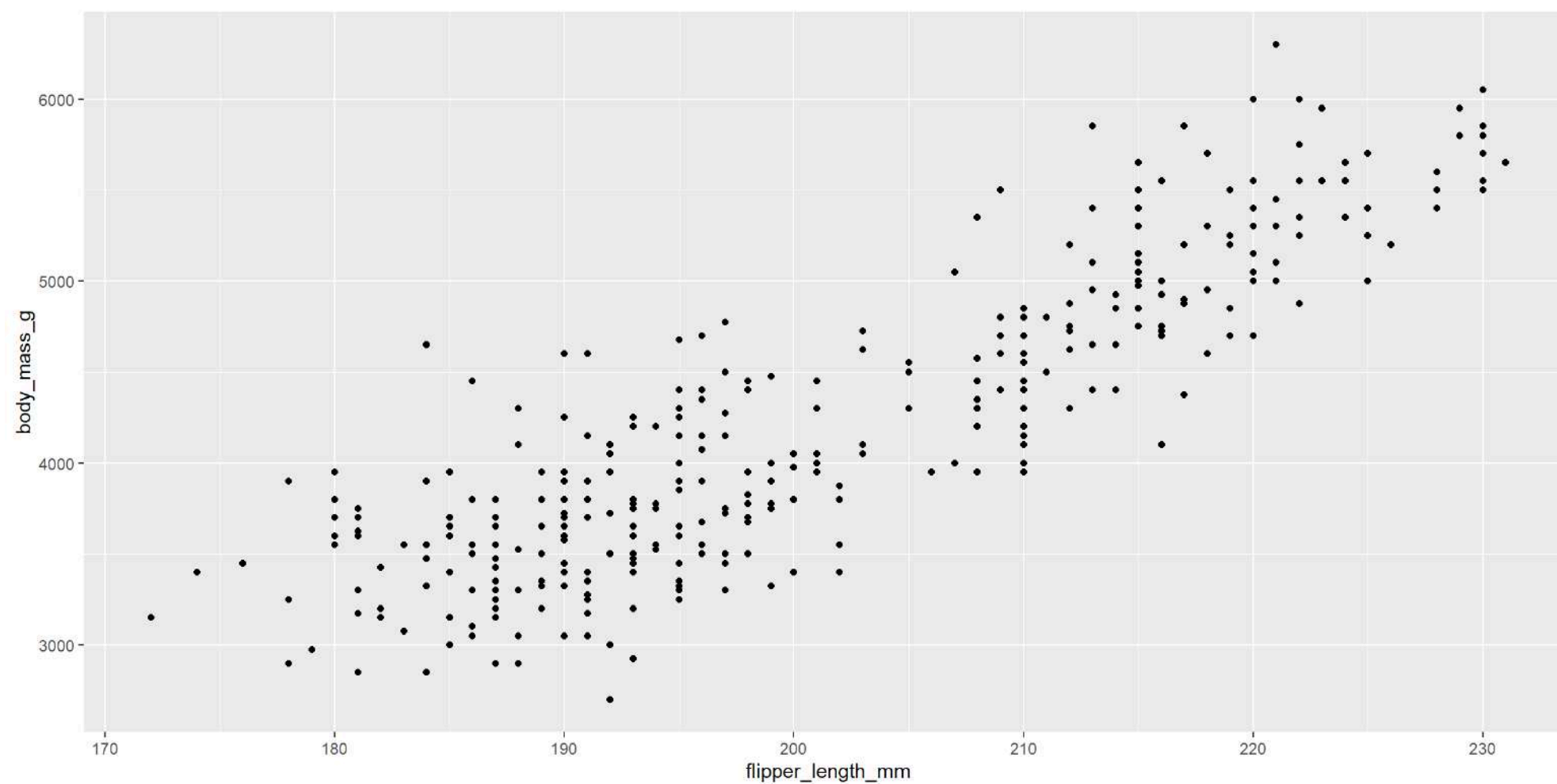
1 x1 <- ggplot(penguins) +
2   geom_point(aes(x=flipper_length_mm, y=body_mass_g)) +
3   stat_smooth(aes(x=flipper_length_mm, y=body_mass_g))
4
5 x2 <- ggplot(penguins, aes(x=flipper_length_mm, y=body_mass_g)) +
6   geom_point() +
7   geom_smooth()
8
9 wrap_plots(x1, x2, nrow=1, ncol=2)

```



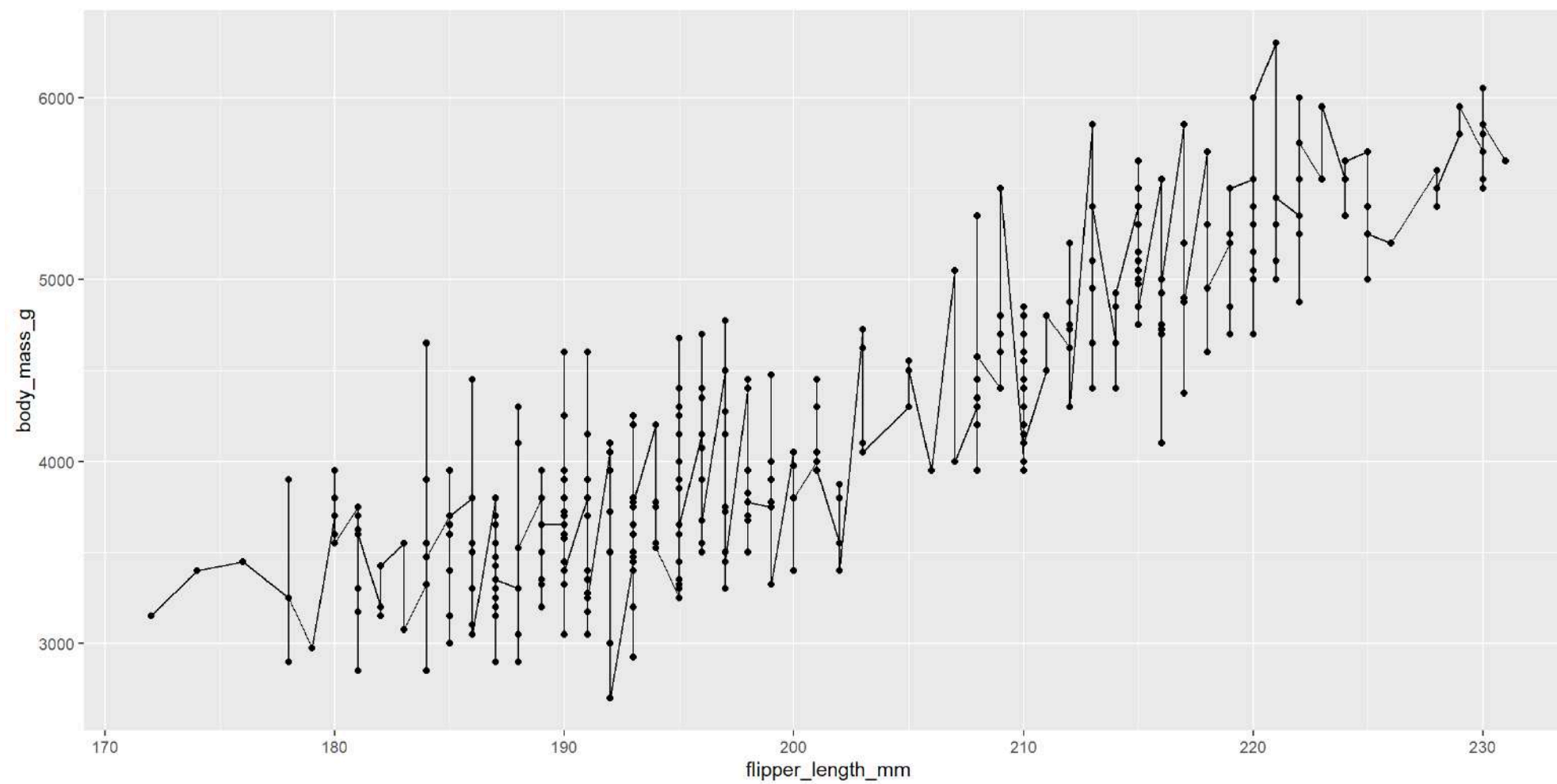
Multiple Geoms

```
1 p <- ggplot(penguins, aes(x=flipper_length_mm, y=body_mass_g)) +  
2   geom_point()  
3  
4  
5  
6  
7  
8  
9 p
```



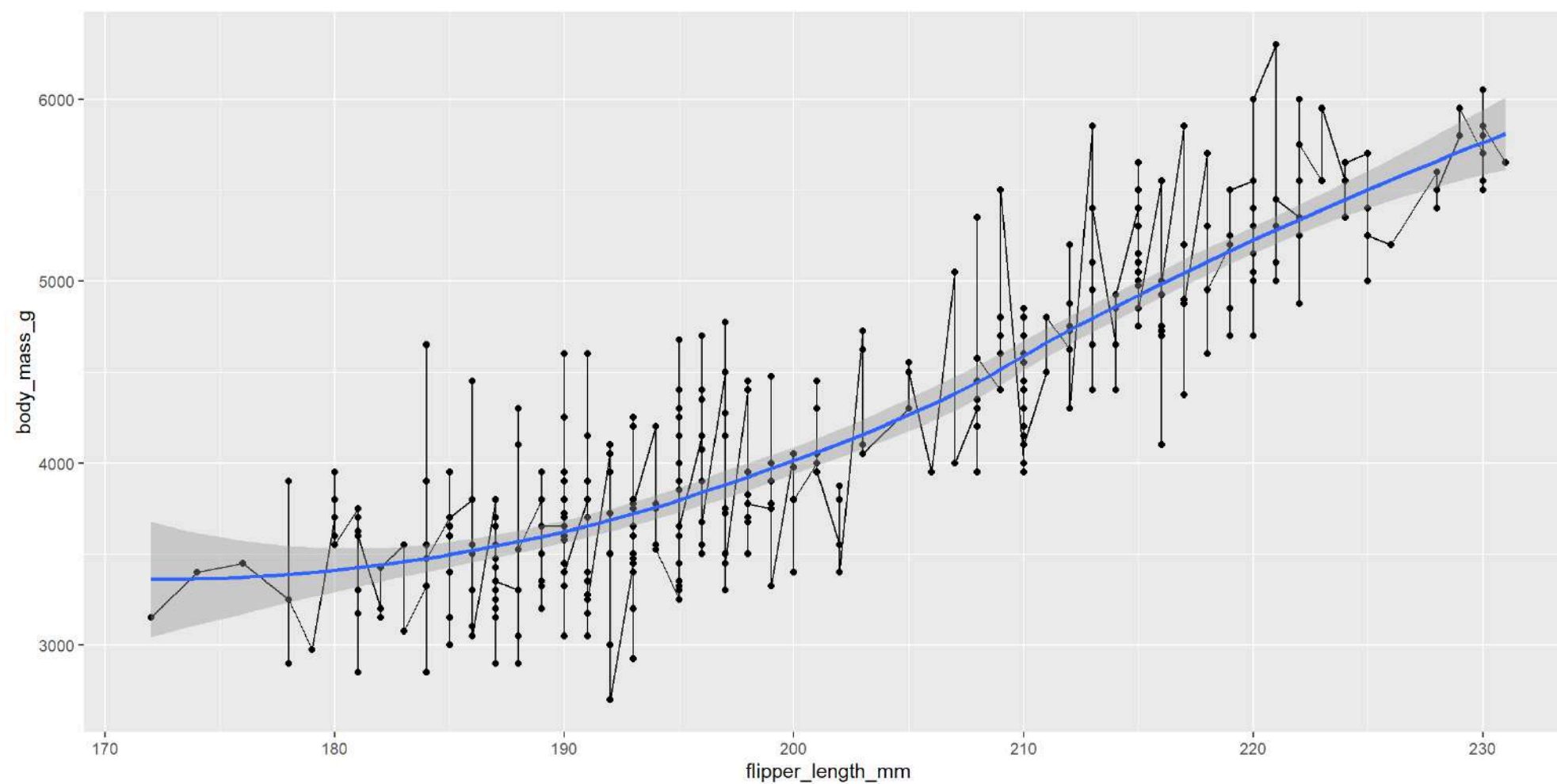
Multiple Geoms

```
1 p <- ggplot(penguins, aes(x=flipper_length_mm, y=body_mass_g)) +  
2   geom_point() +  
3   geom_line()  
4  
5  
6  
7  
8  
9 p
```



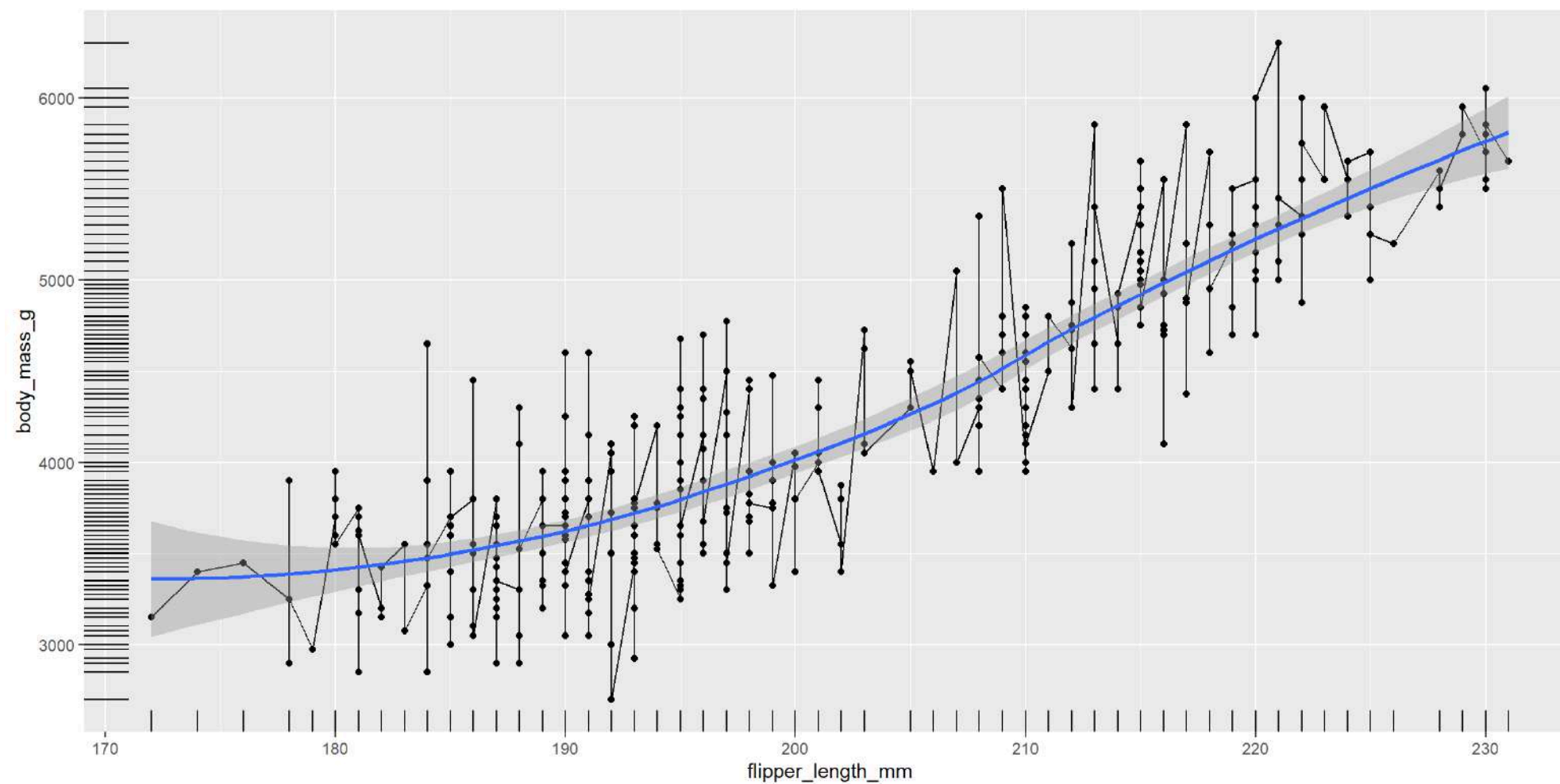
Multiple Geoms

```
1 p <- ggplot(penguins, aes(x=flipper_length_mm, y=body_mass_g)) +  
2   geom_point() +  
3   geom_line() +  
4   geom_smooth()  
5  
6  
7  
8  
9 p
```



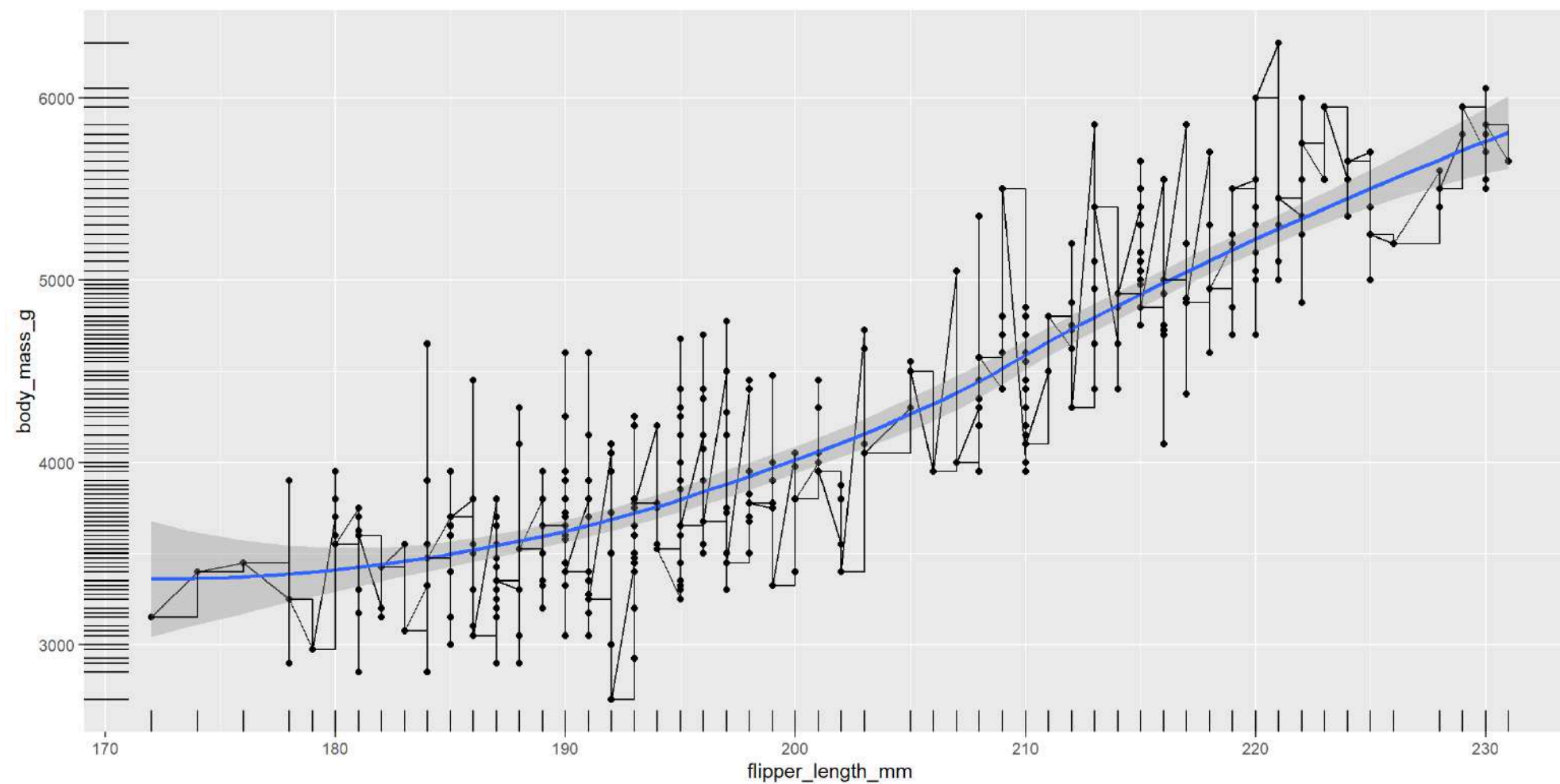
Multiple Geoms

```
1 p <- ggplot(penguins, aes(x=flipper_length_mm, y=body_mass_g)) +  
2   geom_point() +  
3   geom_line() +  
4   geom_smooth() +  
5   geom_rug()  
6  
7  
8  
9 p
```



Multiple Geoms

```
1 p <- ggplot(penguins, aes(x=flipper_length_mm, y=body_mass_g)) +  
2   geom_point() +  
3   geom_line() +  
4   geom_smooth() +  
5   geom_rug() +  
6   geom_step()  
7  
8  
9 p
```

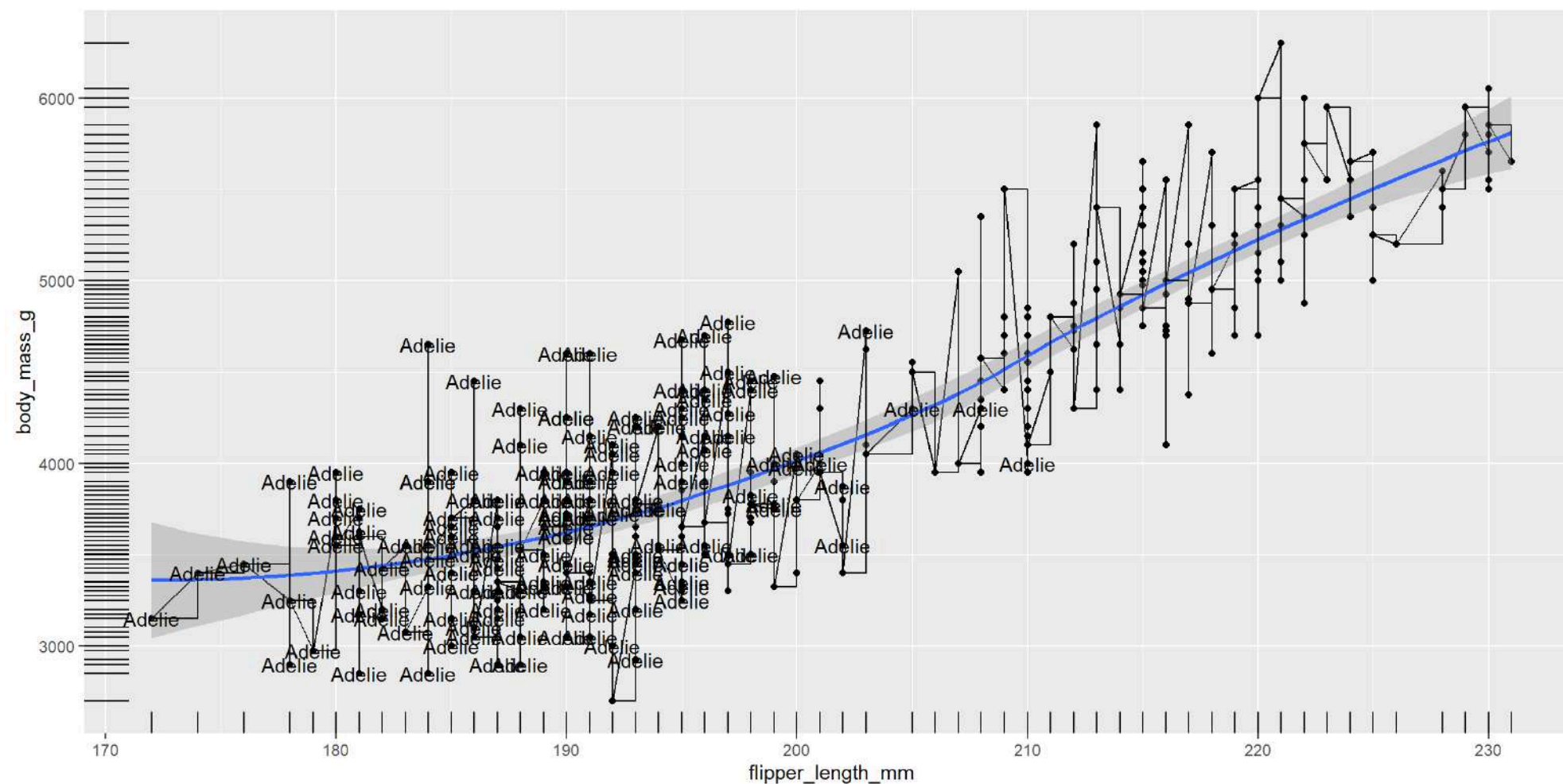


Multiple Geoms

```

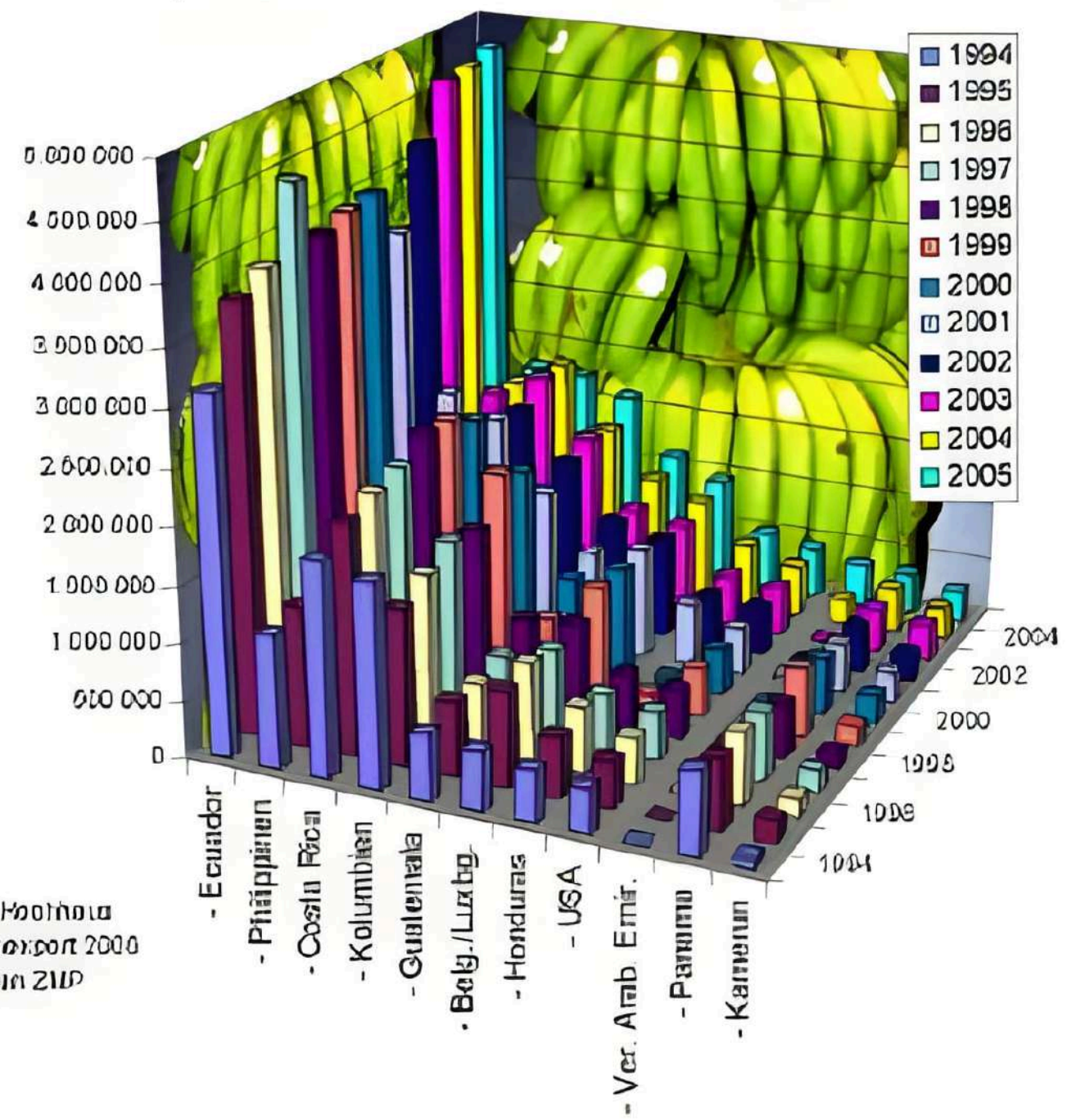
1 p <- ggplot(penguins, aes(x=flipper_length_mm, y=body_mass_g)) +
2   geom_point() +
3   geom_line() +
4   geom_smooth() +
5   geom_rug() +
6   geom_step() +
7   geom_text(data=subset(penguins, penguins$species=="Adelie"),
8             aes(label=species))
9 p

```



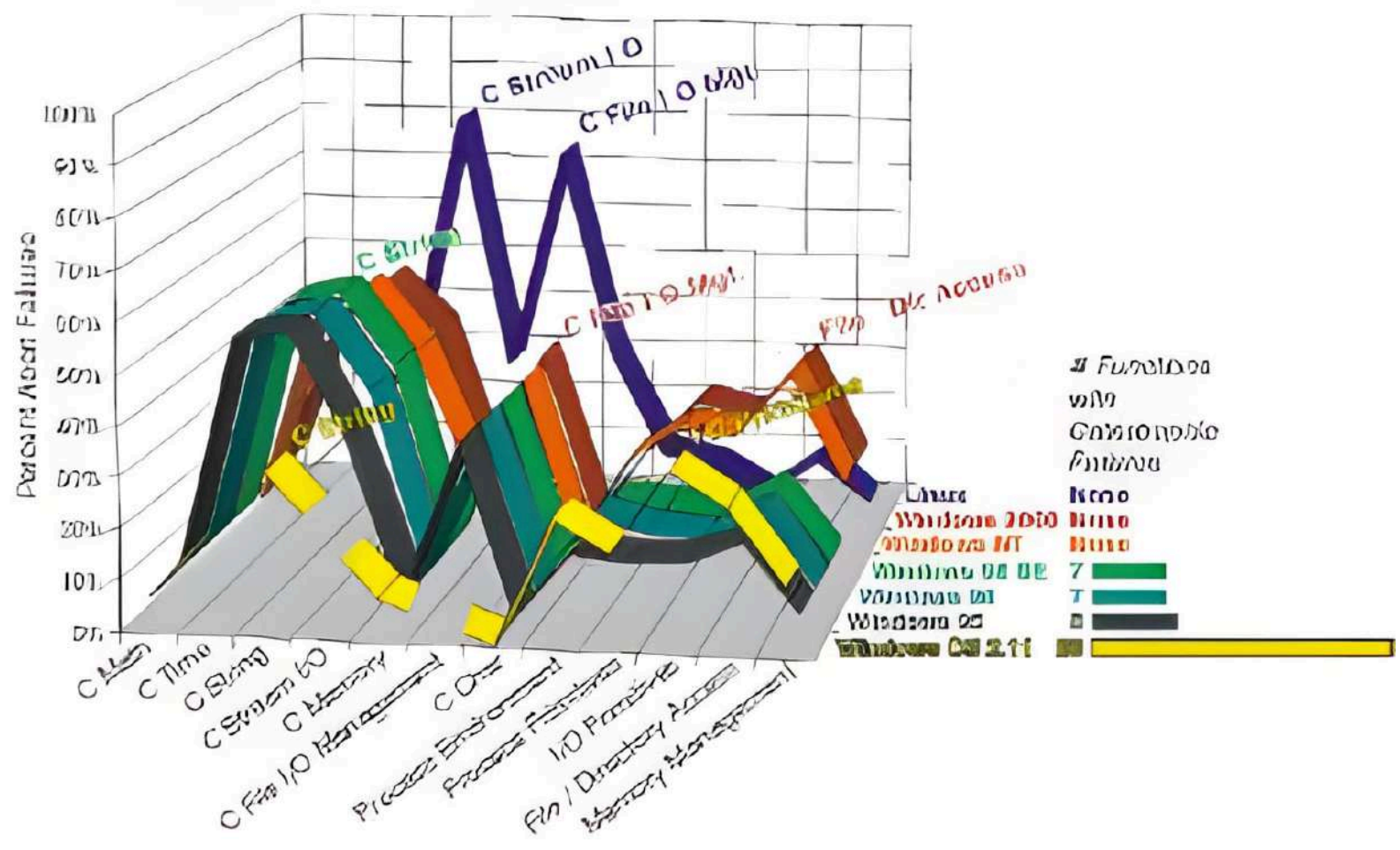
Just because you can doesn't mean you should!

Export von Bananen in Tonnen von 1994-2006



Dr. Kothmann
Berechnung 2006
Daten ZIM

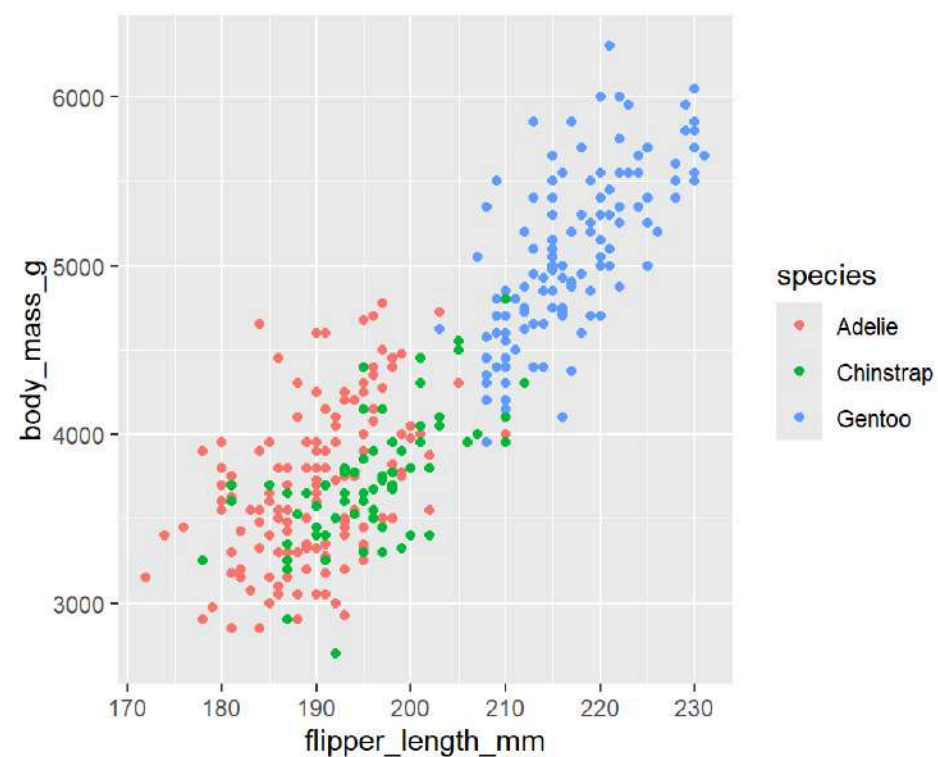
Percent Failures by Functional Group



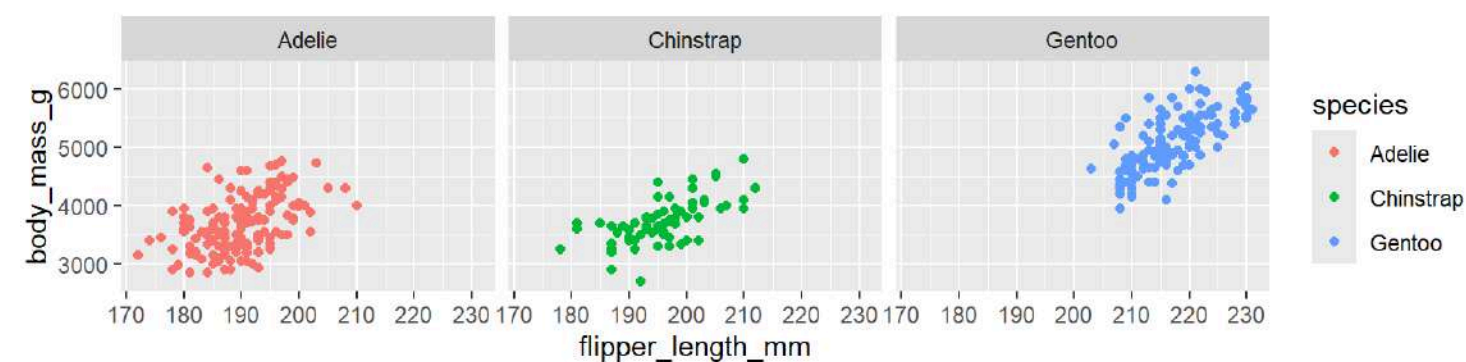
Facets • facet_wrap

- Split to subplots based on variable(s),
- Faceting in one dimension

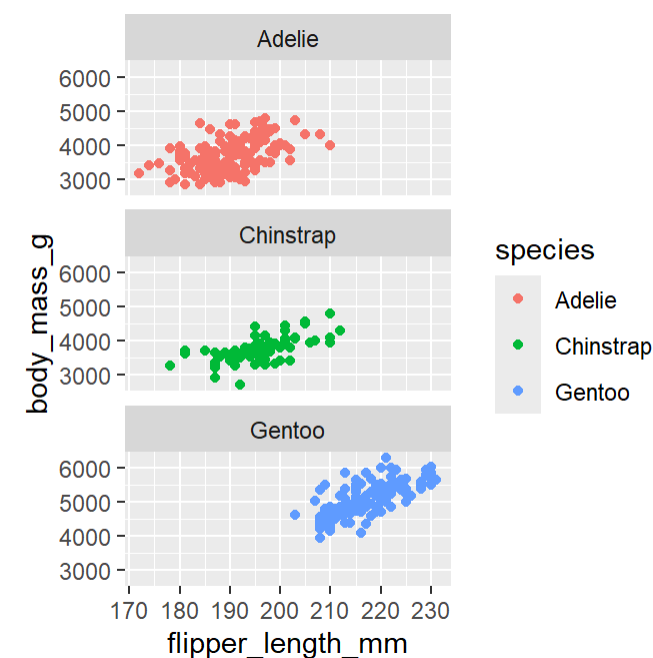
```
1 p <- ggplot(penguins) +
2   geom_point(aes(x=flipper_length_mm,
3                 y=body_mass_g,
4                 color=species))
5 p
```



```
1 p + facet_wrap(~species)
```



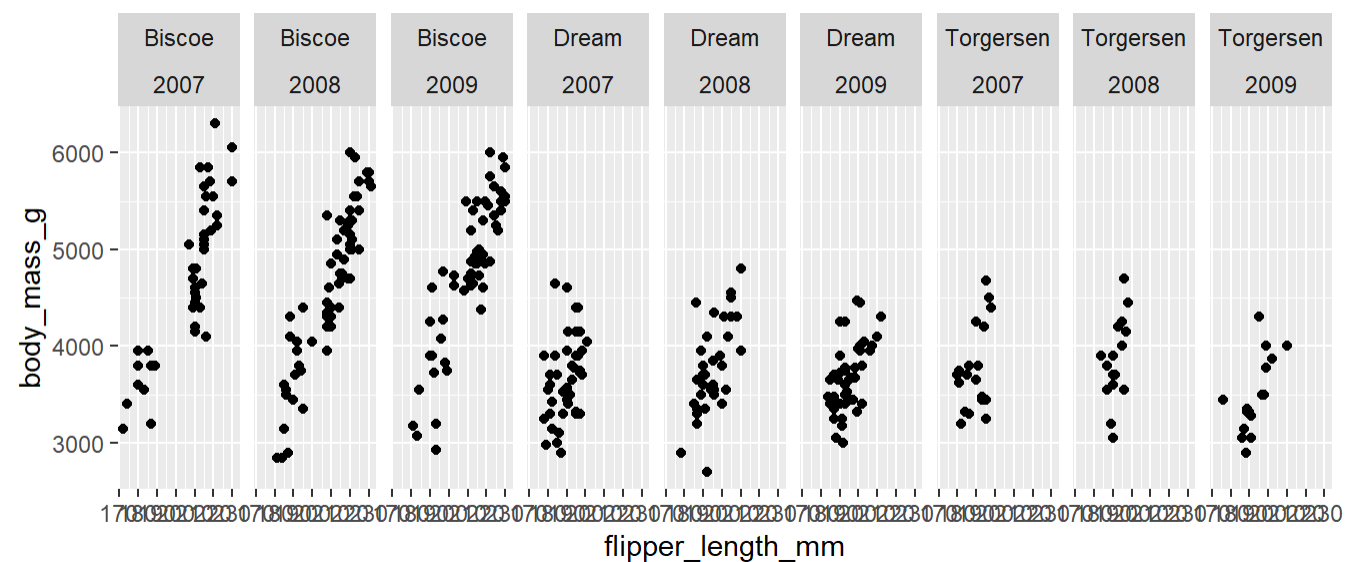
```
1 p + facet_wrap(~species, nrow=3)
```



Facets • facet_grid

- Faceting in two dimensions

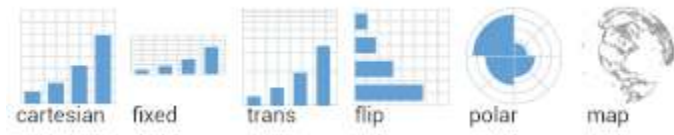
```
1 p <- ggplot(data = penguins, aes(x=flipper_length_mm,
2                                   y=body_mass_g)) +
3   geom_point()
4 p + facet_grid(~island+year)
```



```
1 p + facet_grid(island~year)
```



Coordinate Systems

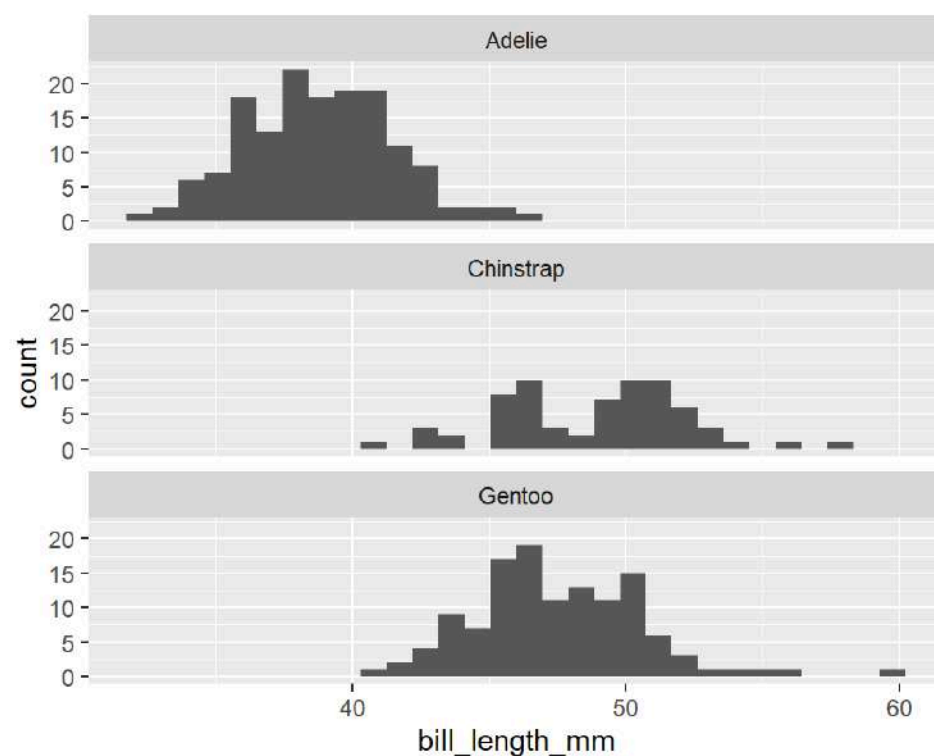


- `coord_cartesian(xlim=c(2,8))` for zooming in
- `coord_map` for controlling limits on maps
- `coord_polar` for polar coordinates

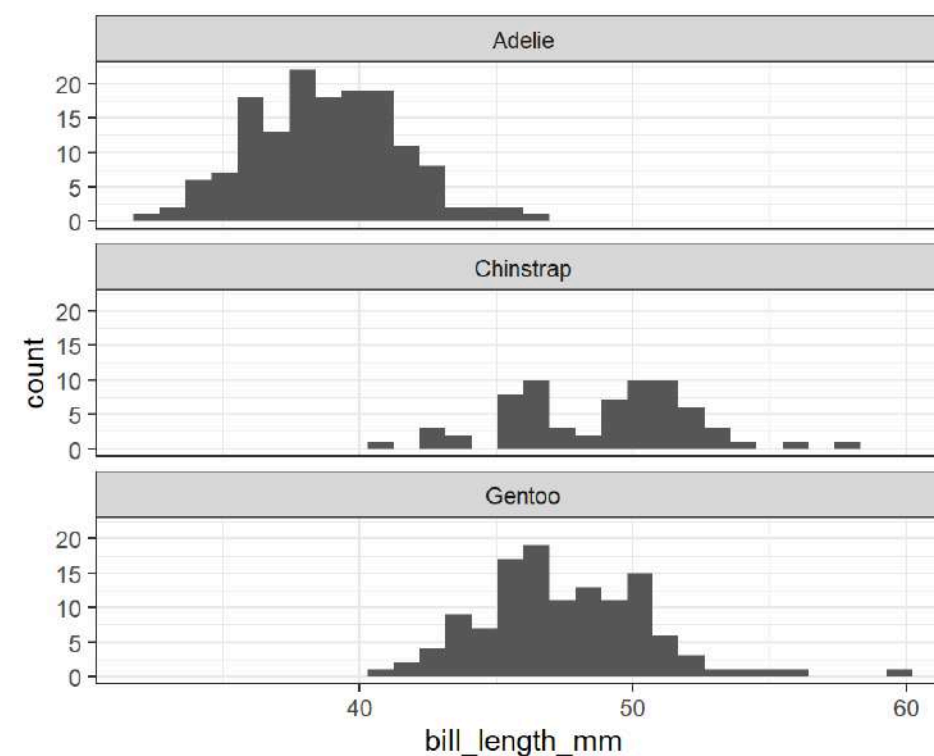
Theming

- Modify non-data plot elements/appearance
- Axis labels, panel colors, legend appearance etc
- Save a particular appearance for reuse
- `?theme`

```
1 ggplot(penguins, aes(x=bill_length_mm)) +
2   geom_histogram() +
3   facet_wrap(~species, ncol = 1) +
4   theme_grey()
```



```
1 ggplot(penguins, aes(x=bill_length_mm)) +
2   geom_histogram() +
3   facet_wrap(~species, ncol = 1) +
4   theme_bw()
```

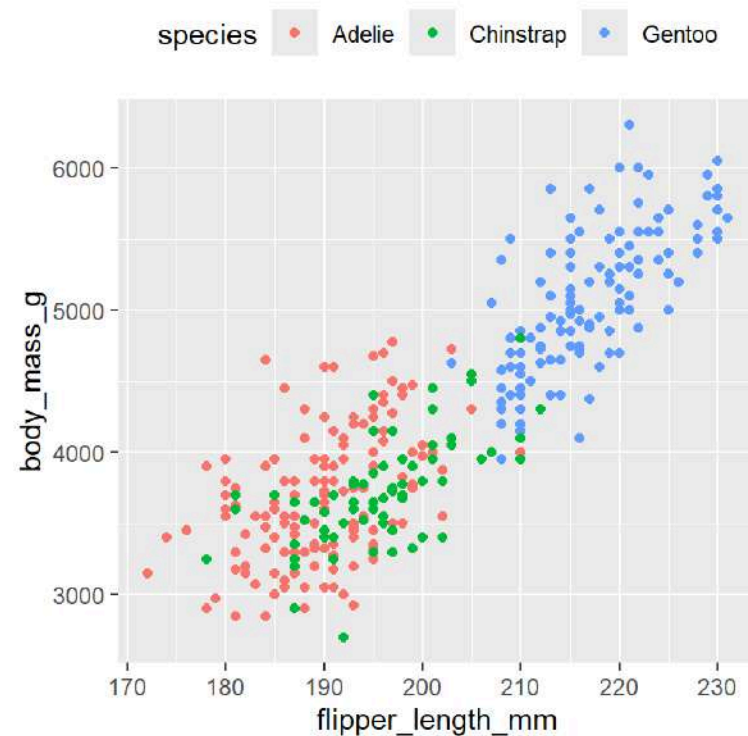


Theme • Legend

```
1 p <- ggplot(penguins) +
2   geom_point(aes(x=flipper_length_mm,
3                 y=body_mass_g,
4                 color=species))
```

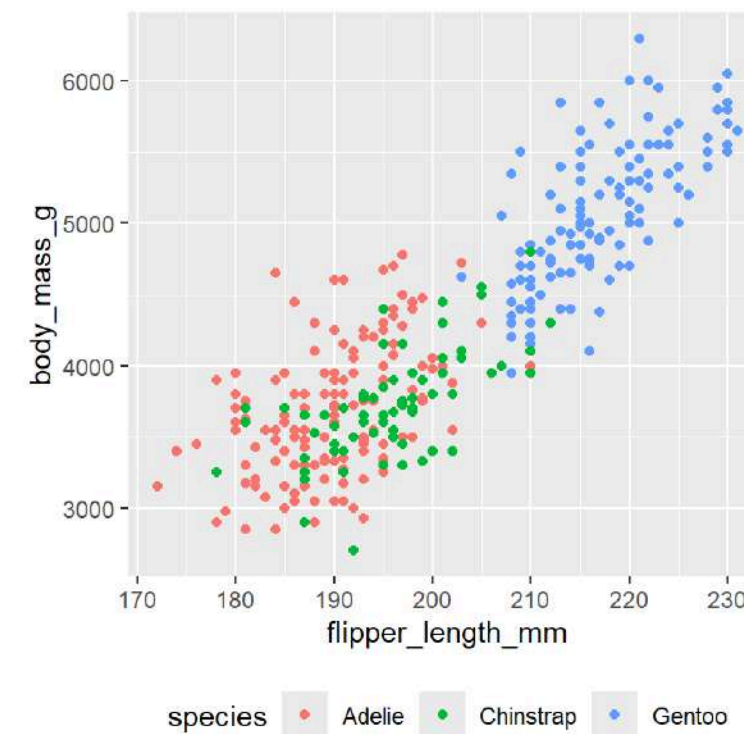
at top

```
1 p + theme(legend.position="top")
```



at bottom

```
1 p + theme(legend.position="bottom")
```

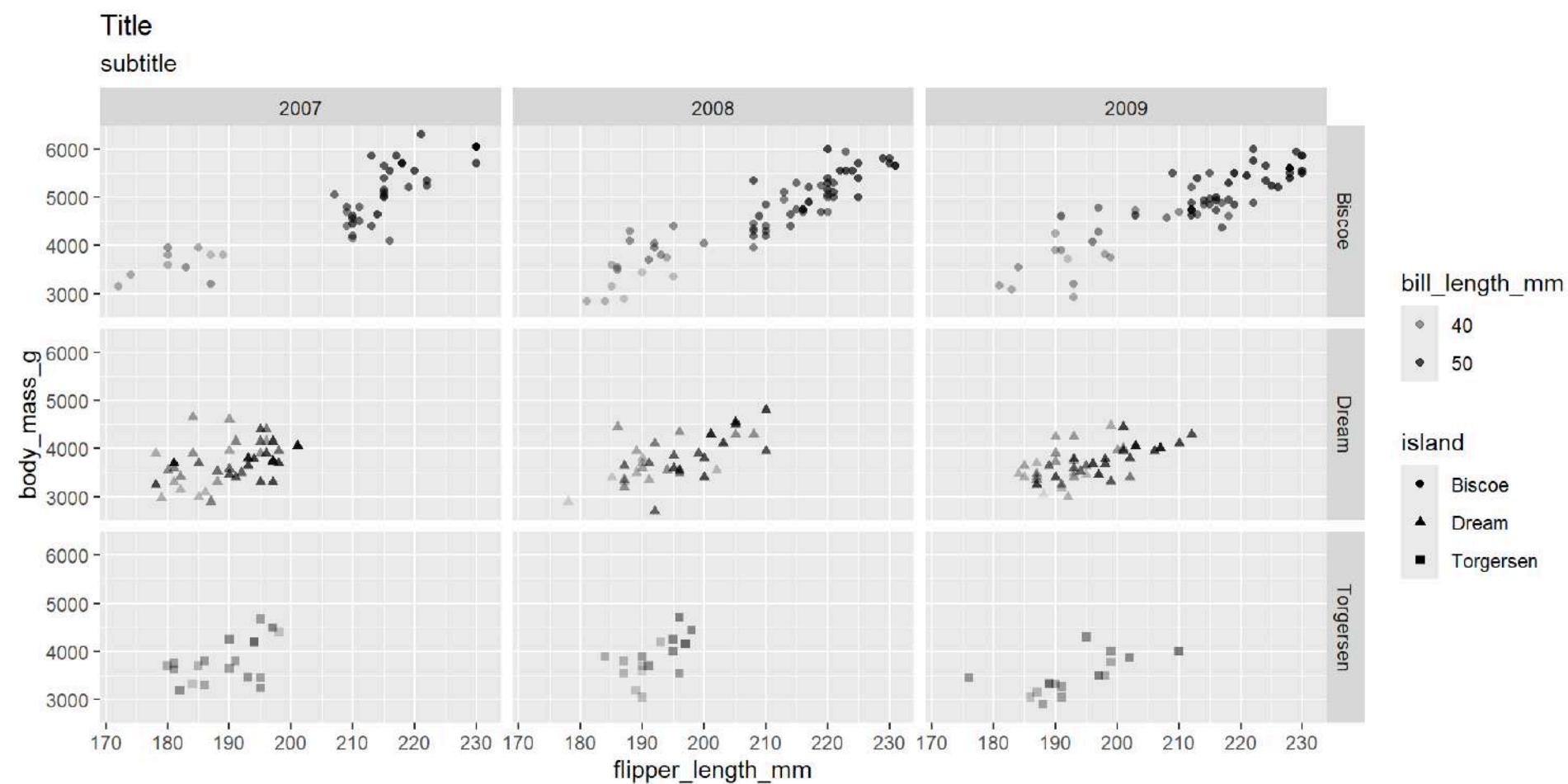


Theme • Text

```

1 p <- ggplot(penguins,
2             aes(x = flipper_length_mm,
3                 y = body_mass_g,
4                 alpha = bill_length_mm,
5                 shape = island)) +
6     geom_point() +
7     facet_grid(island~year) +
8     labs(title="Title",
9          subtitle="subtitle")
10 p

```

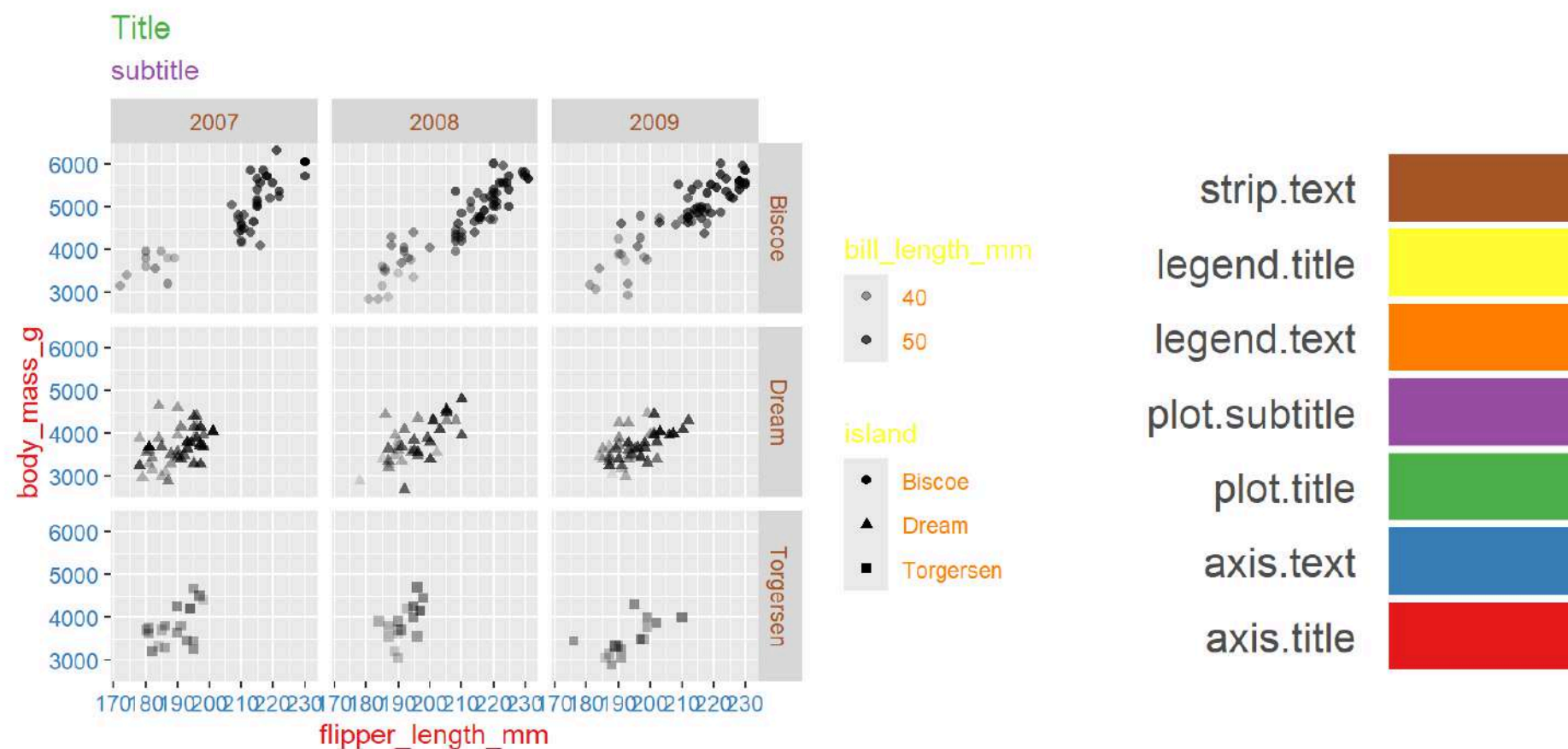


Theme • Text

```

1 p <- p + theme(
2   axis.title=element_text(color="#e41a1c"),
3   axis.text=element_text(color="#377eb8"),
4   plot.title=element_text(color="#4daf4a"),
5   plot.subtitle=element_text(color="#984ea3"),
6   legend.text=element_text(color="#ff7f00"),
7   legend.title=element_text(color="#ffff33"),
8   strip.text=element_text(color="#a65628")
9 )

```

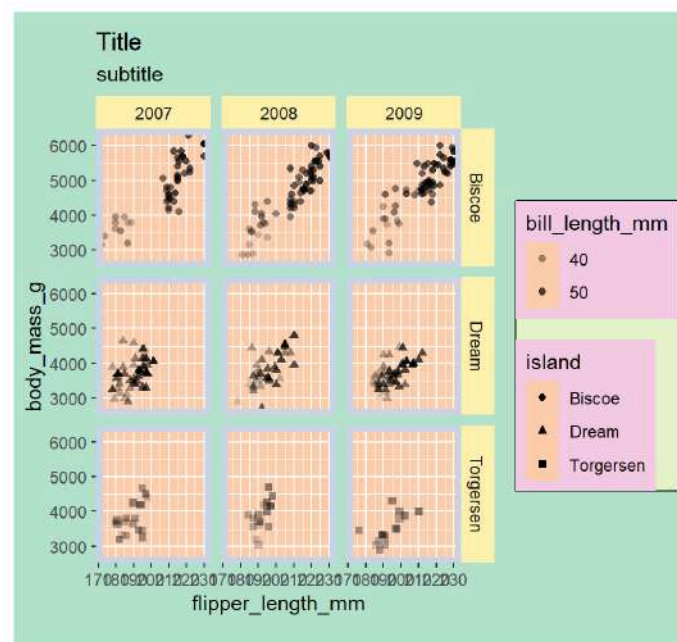


Theme • Rect

List of 5

```
$ fill      : NULL
$ colour    : NULL
$ linewidth : NULL
$ linetype  : NULL
$ inherit.blank: logi FALSE
- attr(*, "class")= chr [1:2] "element_rect" "element"
```

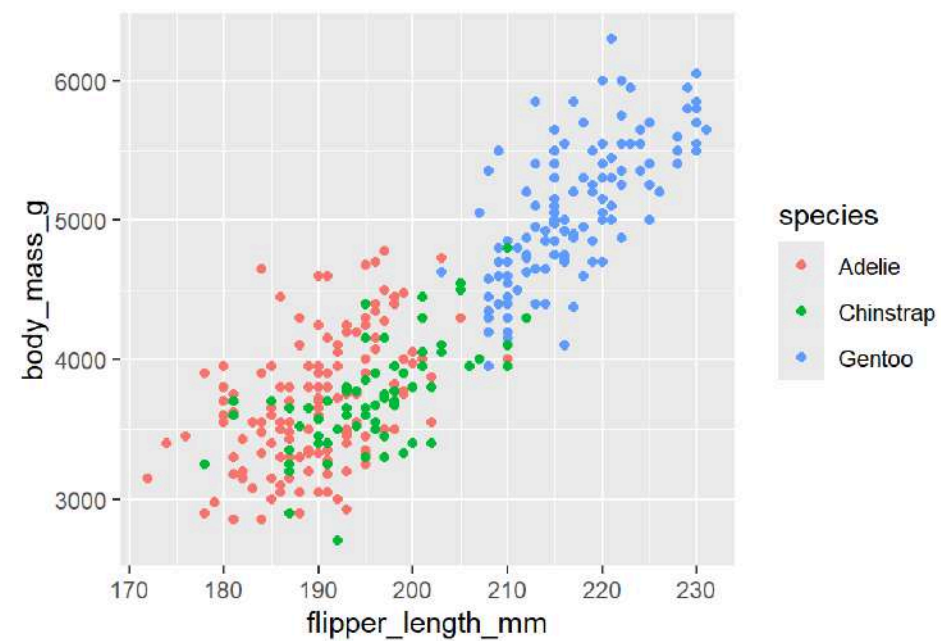
```
1 p <- p + theme(
2   plot.background=element_rect(fill="#b3e2cd"),
3   panel.background=element_rect(fill="#fdcdac"),
4   panel.border=element_rect(fill=NA,color="#cbd5e8",size=3),
5   legend.background=element_rect(fill="#f4cae4"),
6   legend.box.background=element_rect(fill="#e6f5c9"),
7   strip.background=element_rect(fill="#fff2ae")
8 )
```



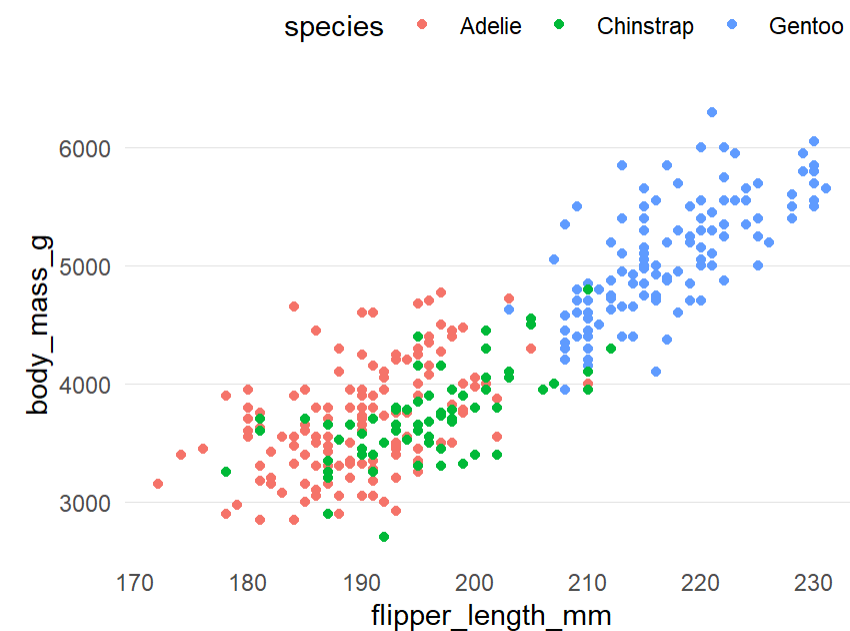
Theme • Reuse

```
1 newtheme <- theme_bw() + theme(
2   axis.ticks=element_blank(), panel.background=element_rect(fill="white"),
3   panel.grid.minor=element_blank(), panel.grid.major.x=element_blank(),
4   panel.grid.major.y=element_line(size=0.3,color="grey90"), panel.border=element_blank(),
5   legend.position="top", legend.justification="right"
6 )
```

```
1 p
```

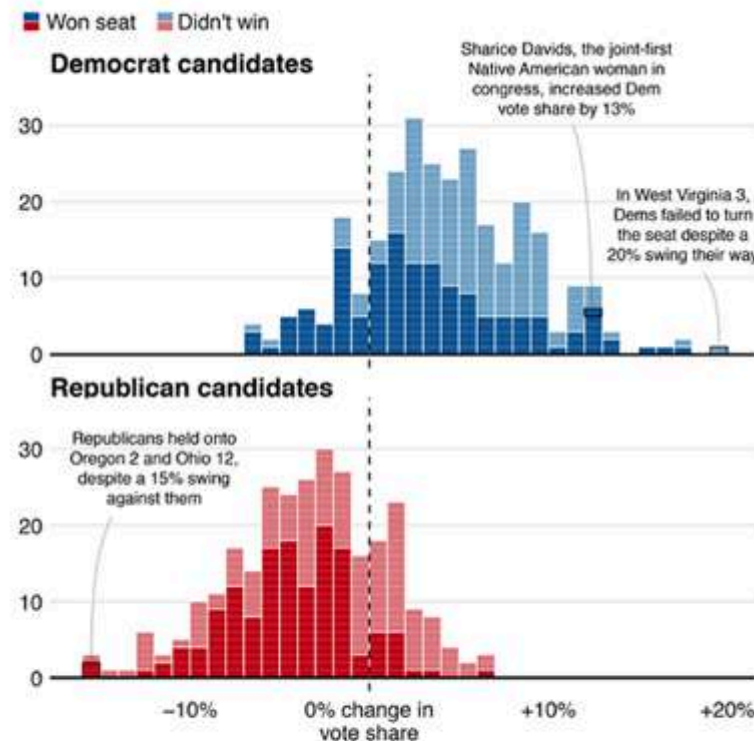


```
1 p + newtheme
```



Professional themes

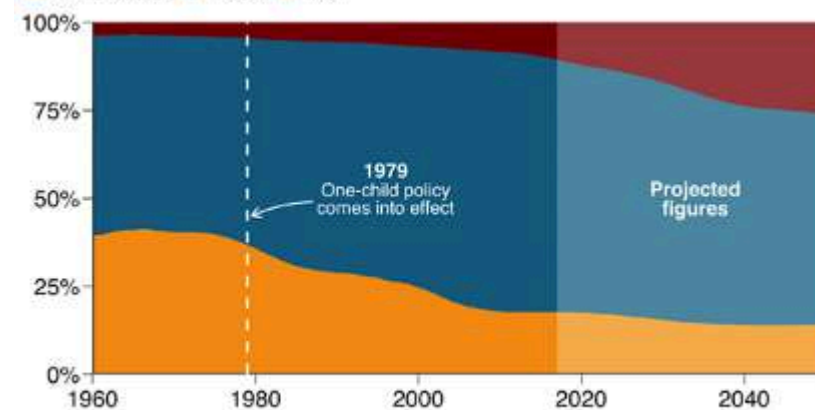
Blue wave



Breakdown of China's population by age group

Proportion of total population (1960-2050)

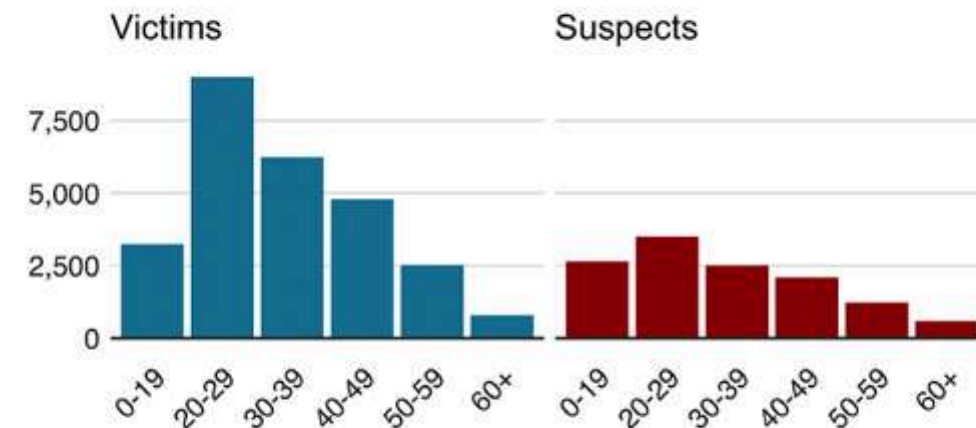
■ 0-14 years ■ 15-64 ■ 65+



Source: The World Bank

Homophobic hate crimes are mainly committed by young people on young people

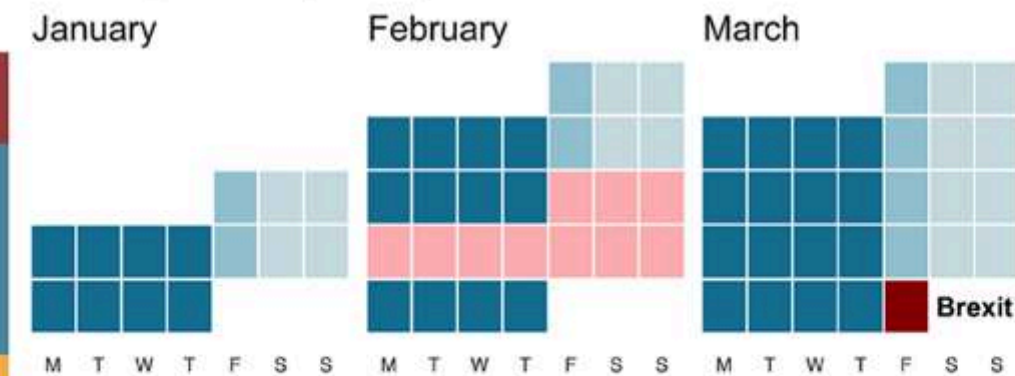
Number in each age group 2014 - 2017



Source: BBC Freedom of Information requests to UK police forces

The Commons has 36 normal working days until Brexit

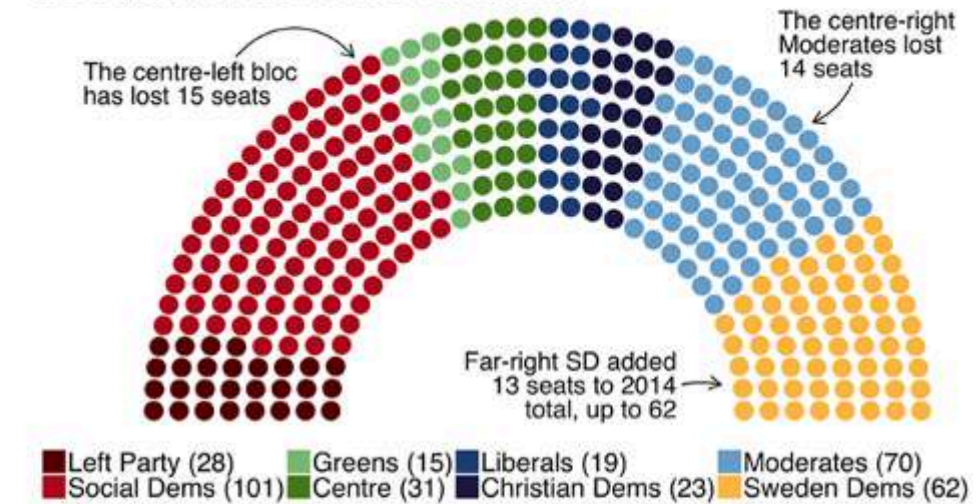
■ Monday to Thursday ■ Friday ■ Weekend ■ Recess



Note: The House of Commons sometimes sits on Fridays to debate individual MPs' bills

Source: Parliament

Results of the 2018 election

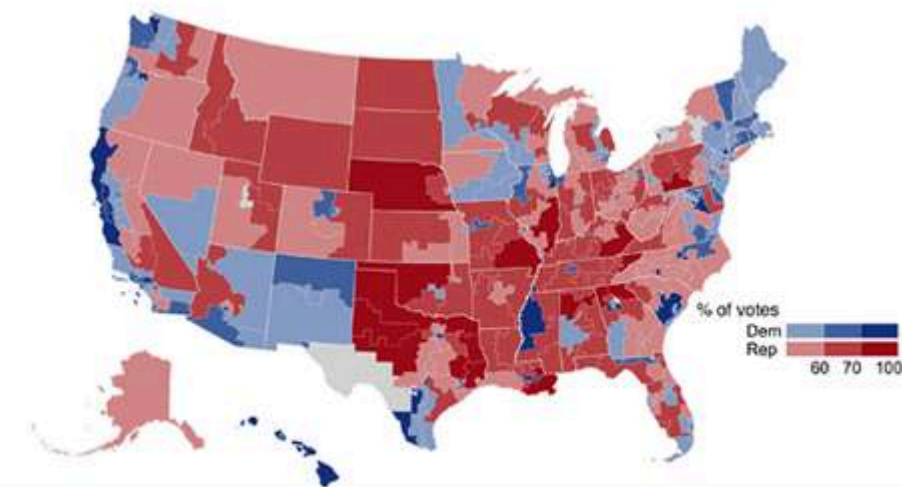


Source: Reuters

BBC

Democrats take the House

Dem 232 218 to win Rep 198



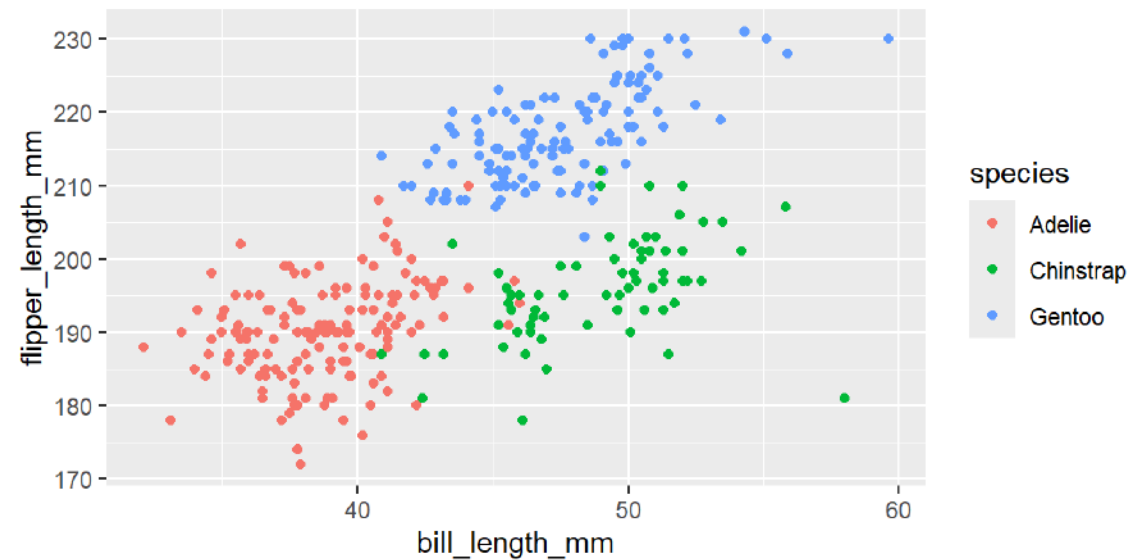
Source: AP. Grey districts are undeclared

BBC

How BBC works with R graphics

Saving plots

```
1 p <- ggplot(penguins, aes(x=bill_length_mm, y=flipper_length_mm, color=species)) +
2   geom_point()
3 p
```



- **ggplot2** package offers a convenient function

```
1 ggsave("plot.png", p, height=5, width=7, units="cm", dpi=200)
2 # Note that default units in png is pixels while in ggsave it's inches
```

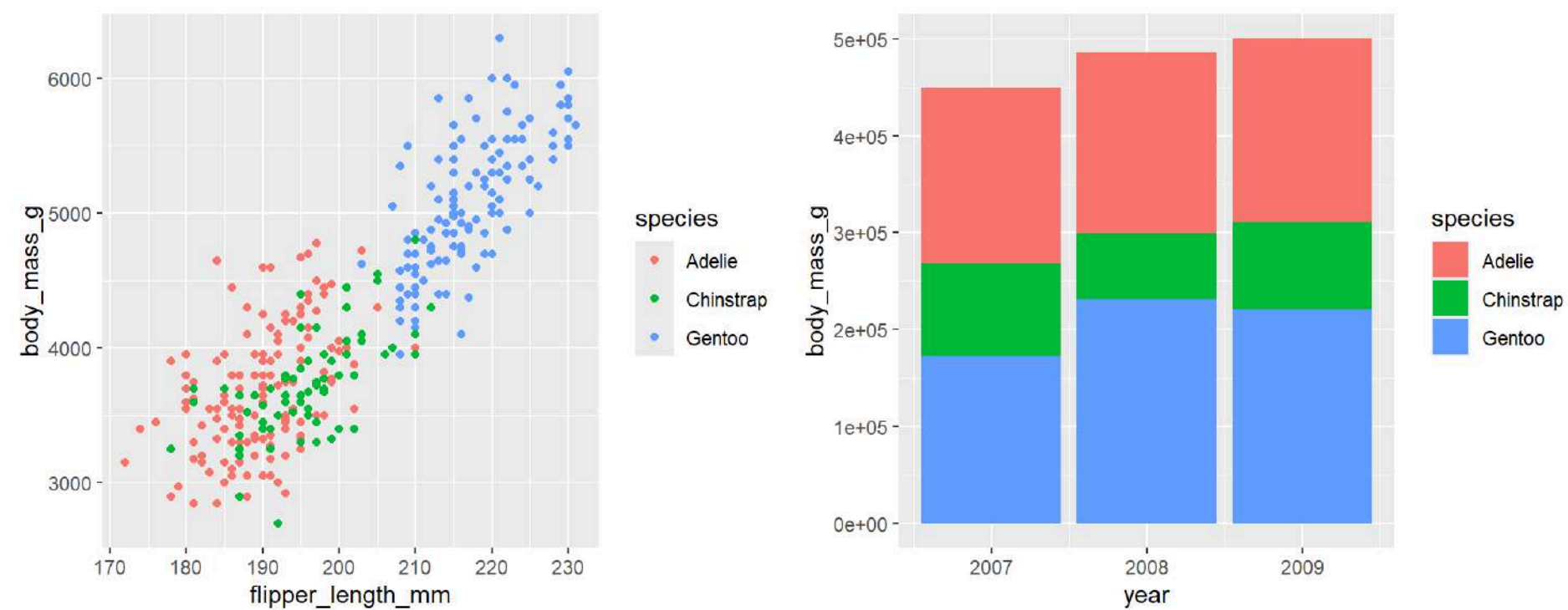
- **ggplot2** plots can be saved just like base plots

```
1 png("plot.png", height=5, width=7, units="cm", res=200)
2 print(p)
3 dev.off()
```

Combining Plots

```
1 p <- ggplot(penguins, aes(x=flipper_length_mm, y=body_mass_g, color=species)) + geom_point()
2 q <- ggplot(penguins, aes(x=year, y=body_mass_g, fill=species)) + geom_bar(stat="identity")
```

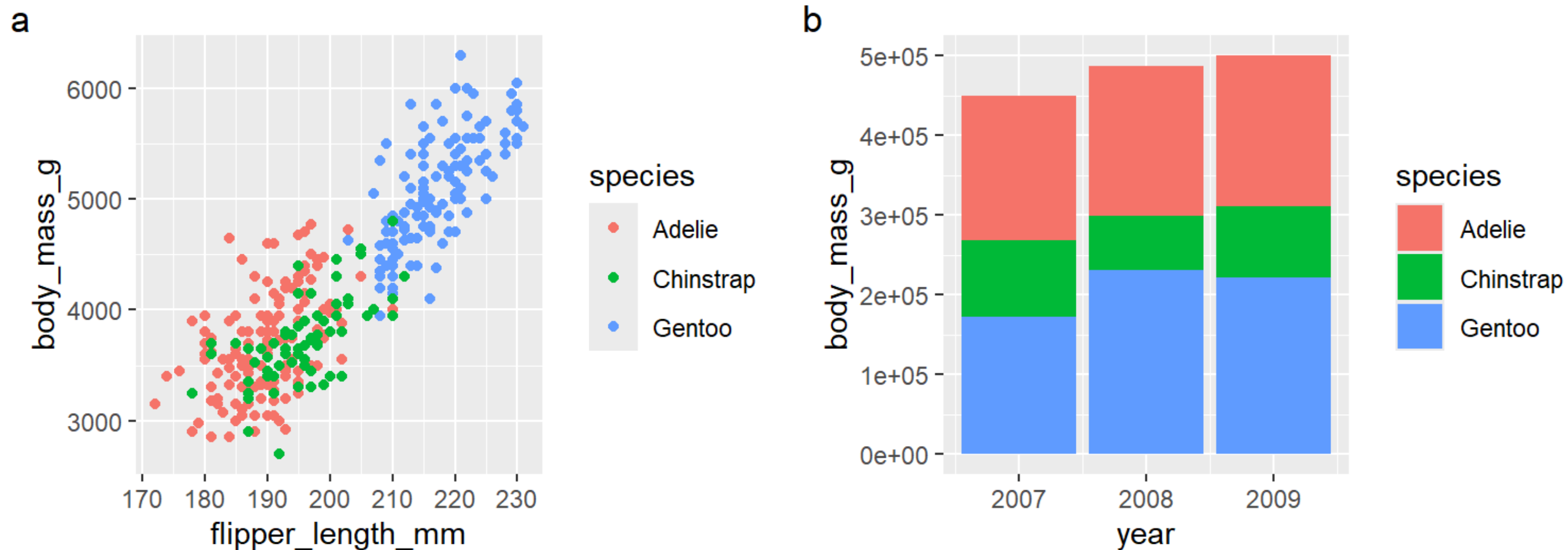
```
1 patchwork::wrap_plots(p,q)
```



Combining Plots

```
1 p <- ggplot(penguins, aes(x=flipper_length_mm, y=body_mass_g, color=species)) + geom_point()
2 q <- ggplot(penguins, aes(x=year, y=body_mass_g, fill=species)) + geom_bar(stat="identity")
```

```
1 patchwork::wrap_plots(p,q) +
2   plot_annotation(tag_levels = 'a')
```



[patchwork documentation.](#)

Extensions

- **patchwork**: Combining plots
- **ggrepel**: Text labels including overlap control
- **ggforce**: Circles, splines, hulls, voronoi etc
- **ggpmisc**: Miscellaneous features
- **ggthemes**: Set of extra themes
- **ggthemr**: More themes
- **ggsci**: Color palettes for scales
- **ggmap**: Dedicated to mapping
- **ggraph**: Network graphs
- **ggiraph**: Converting ggplot2 to interactive graphics

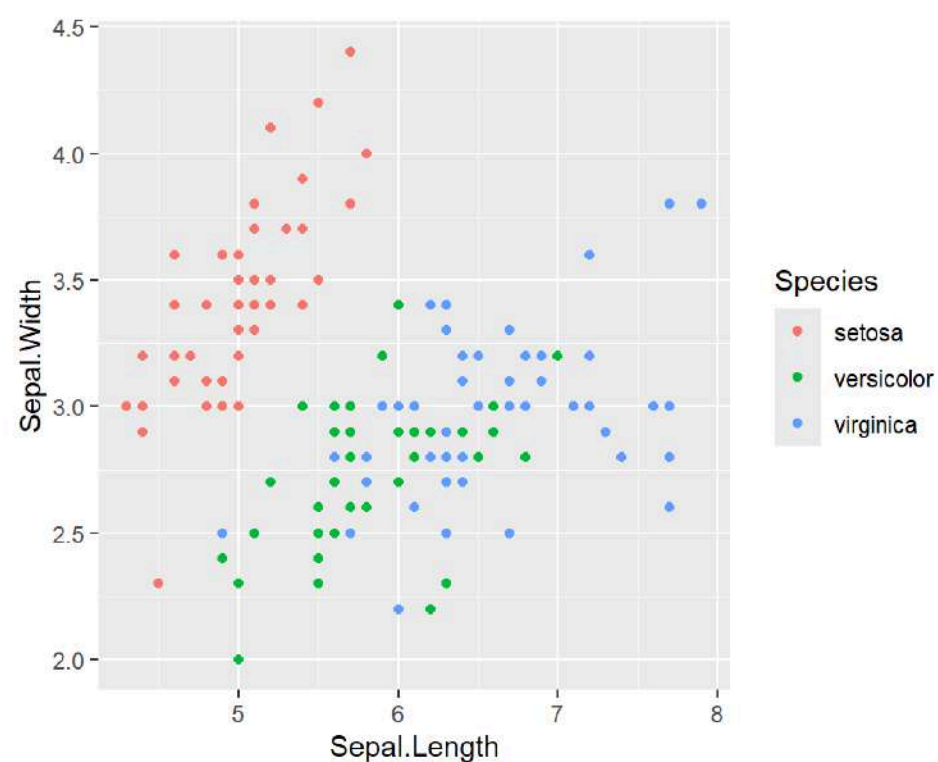
A collection of ggplot extension packages: <https://exts.ggplot2.tidyverse.org/>.

Curated list of ggplot2 links: <https://github.com/erikgahner/awesome-ggplot2>.

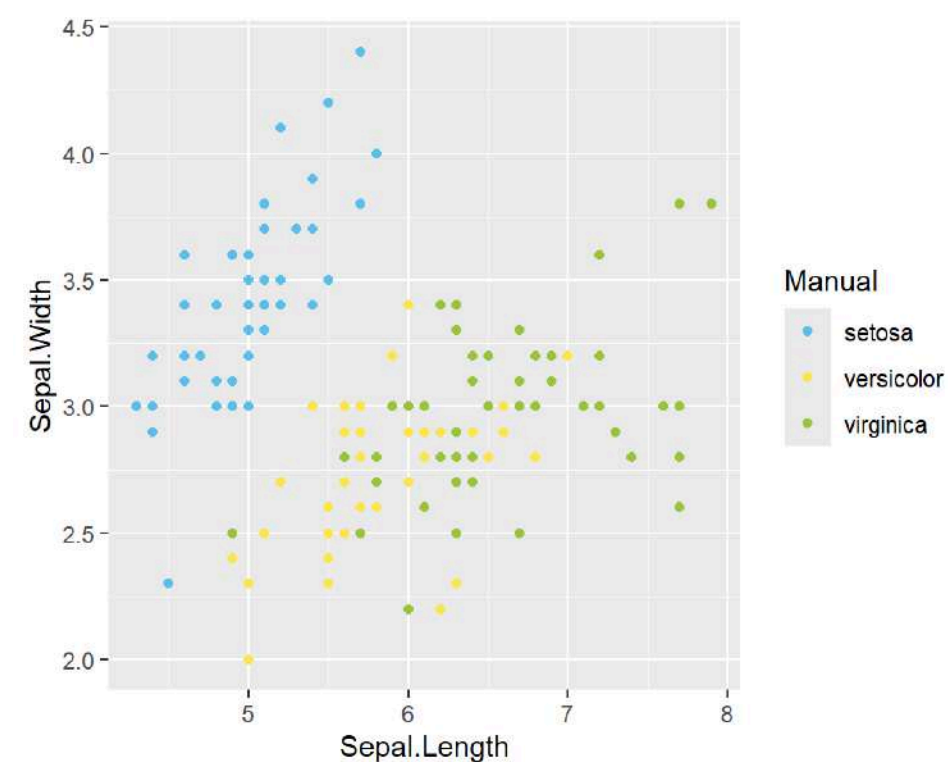
Scales • Discrete Colors

- scales: position, color, fill, size, shape, alpha, linetype
- syntax: `scale_<aesthetic>_<type>`

```
1 p <- ggplot(iris)+geom_point(aes(x=Sepal.Length,
2                               y=Sepal.Width,color=Species))
3 p
```



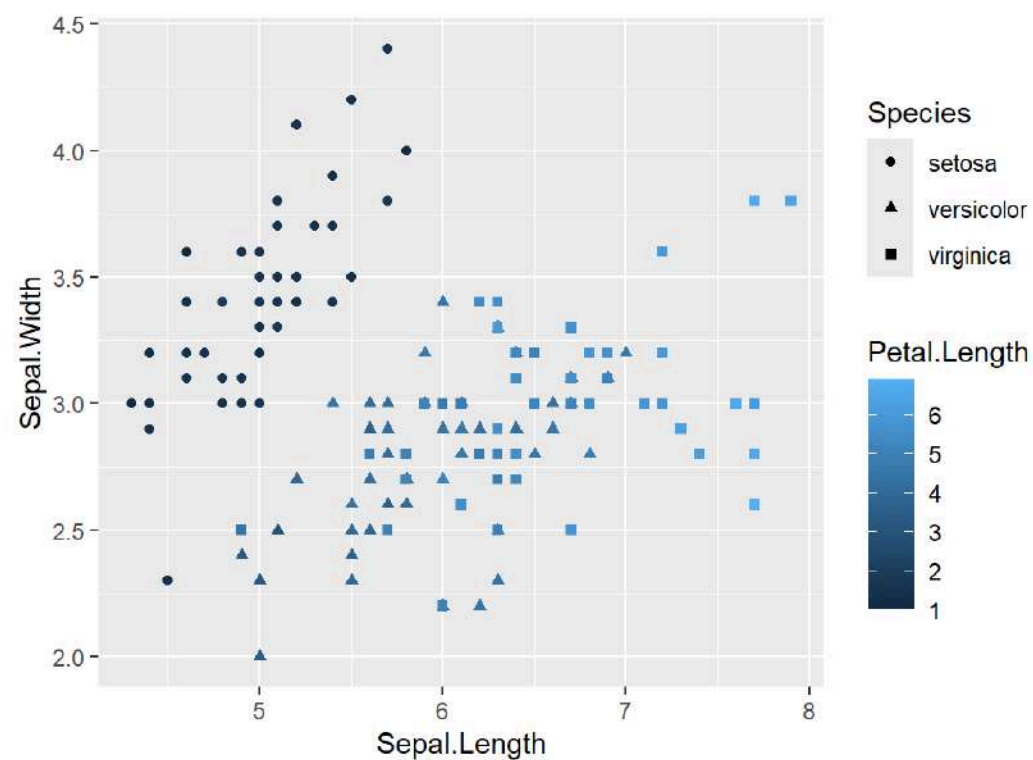
```
1 p + scale_color_manual(
2   name="Manual",
3   values=c("#5BC0EB", "#FDE74C", "#9BC53D"))
```



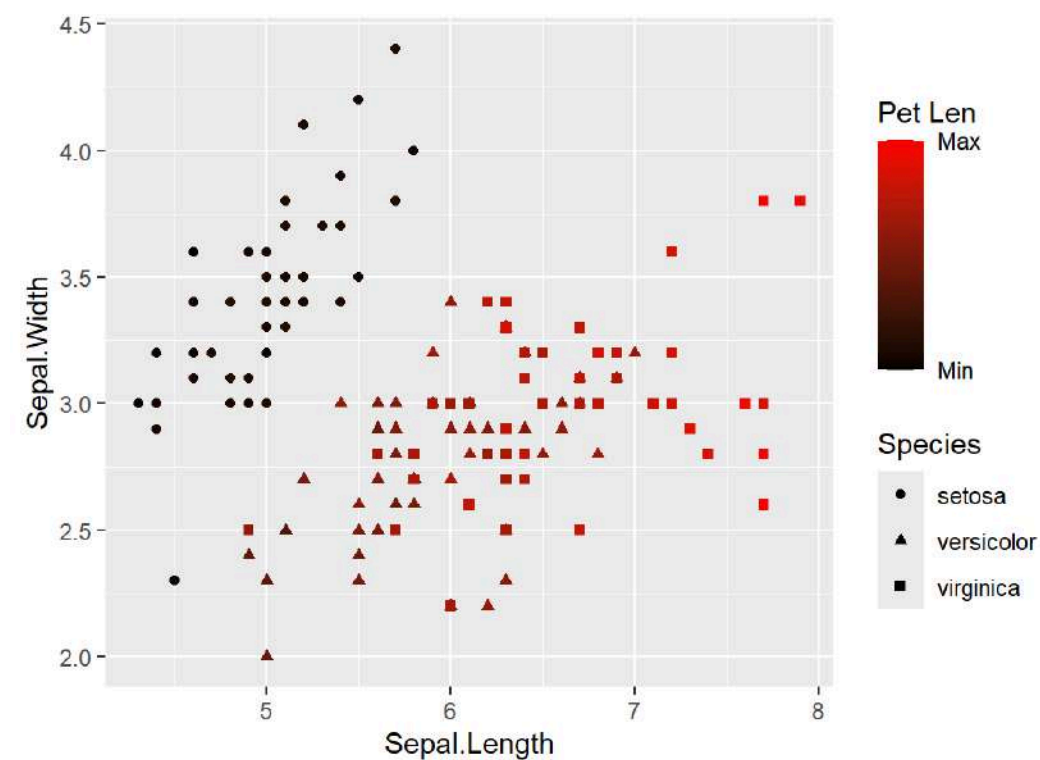
Scales • Continuous Colors

- In RStudio, type `scale_`, then press **TAB**

```
1 p <- ggplot(iris)+
2   geom_point(aes(x=Sepal.Length,
3                 y=Sepal.Width,
4                 shape=Species,color=Petal.Length))
5 p
```

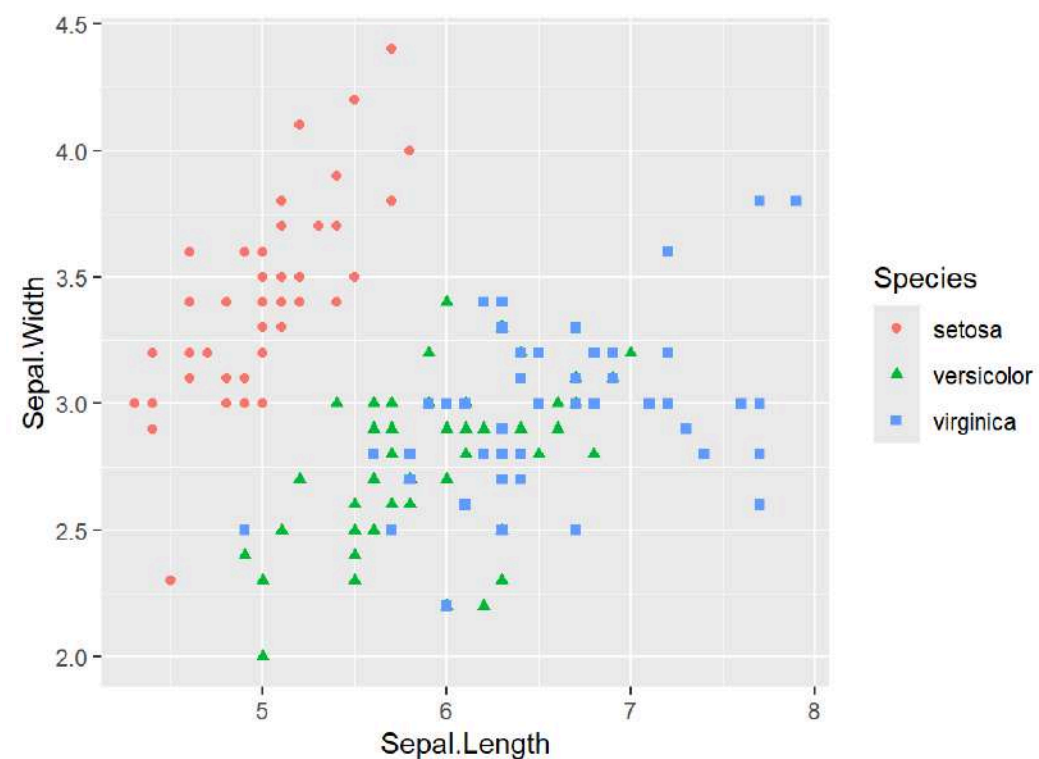


```
1 p +
2   scale_color_gradient(name="Pet Len",
3                       breaks=range(iris$Petal.Length),
4                       labels=c("Min", "Max"),
5                       low="black",high="red")
```

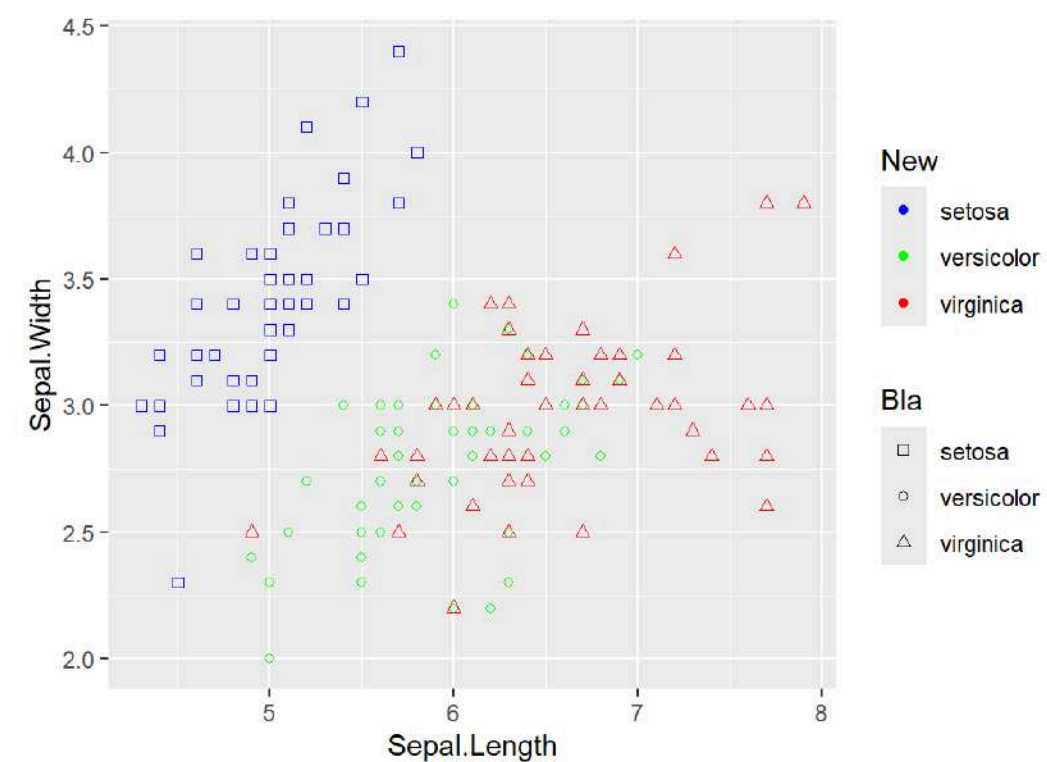


Scales • Shape

```
1 p <- ggplot(iris)+
2   geom_point(aes(x=Sepal.Length,
3                 y=Sepal.Width,
4                 shape=Species,color=Species))
5 p
```



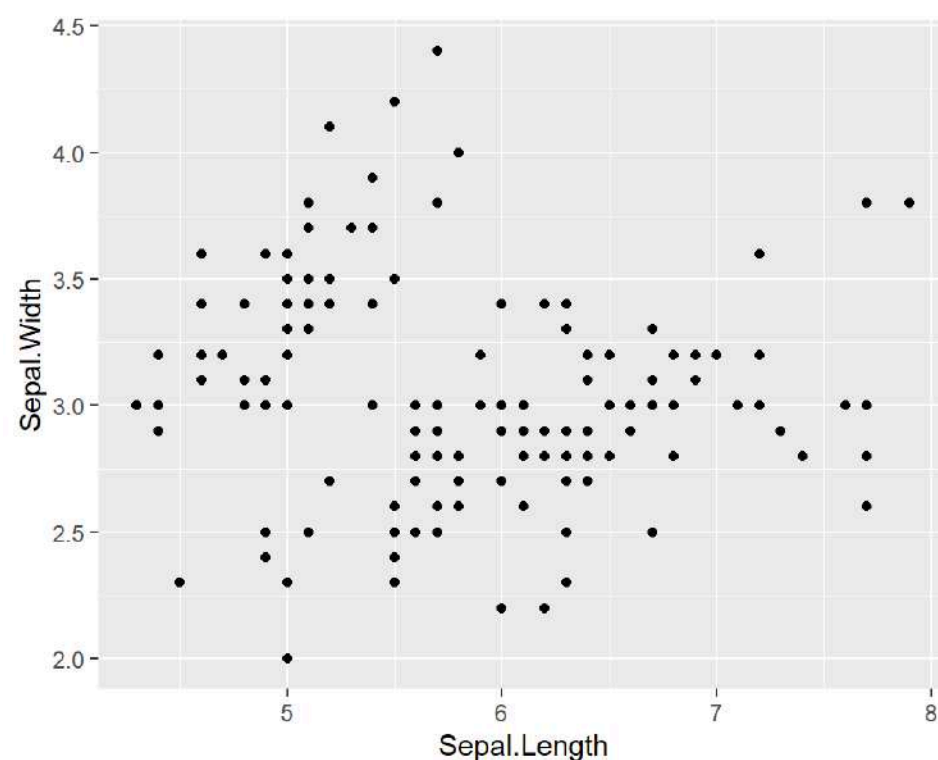
```
1 p +
2   scale_color_manual(name="New",
3                     values=c("blue","green","red"))+
4   scale_shape_manual(name="Bla",values=c(0,1,2))
```



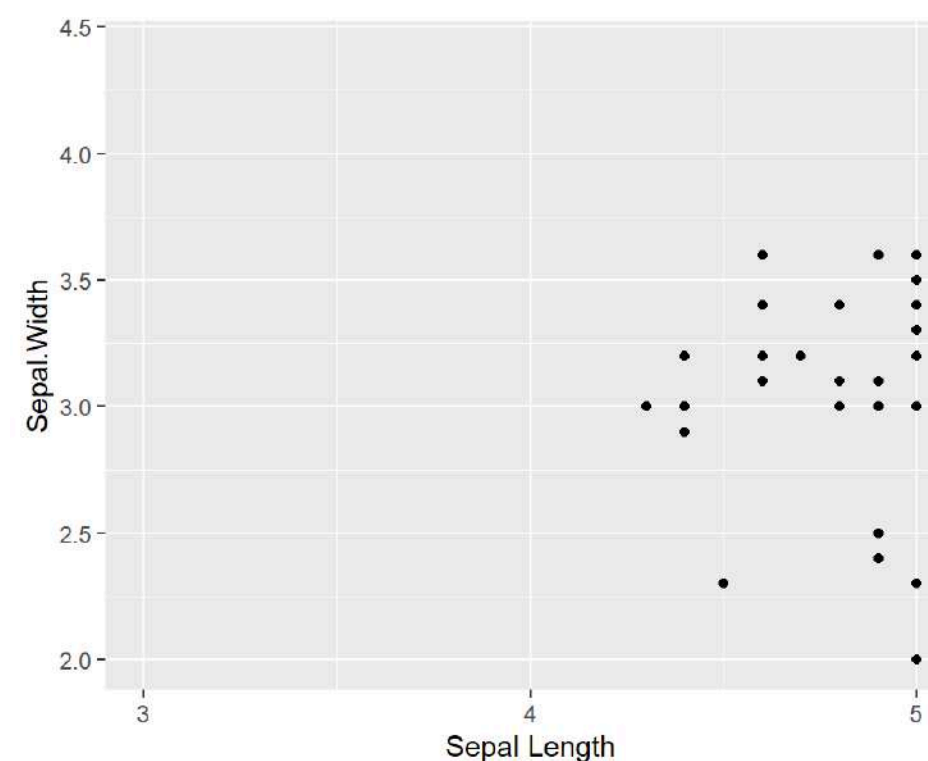
Scales • Axes

- scales: x, y
- syntax: `scale_<axis>_<type>`
- arguments: name, limits, breaks, labels

```
1 p <- ggplot(iris)+geom_point(
2   aes(x=Sepal.Length,y=Sepal.Width))
3 p
```

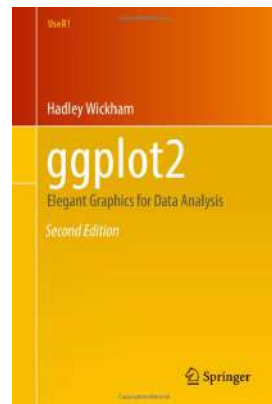


```
1 p + scale_x_continuous(name="Sepal Length",
2   breaks=seq(1,8),limits=c(3,5))
```

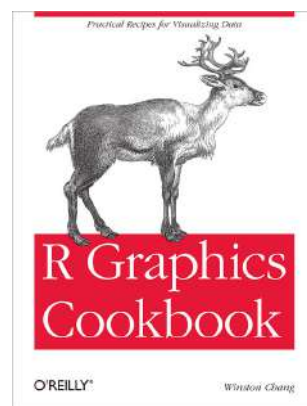


Help

- **ggplot2 book**



- **The R cookbook**



- **ggplot2 official reference**
- **RStudio cheatsheet**
- **r-statistics ggplot2 cheatsheet**
- **StackOverflow**
- **Blogs, R-Bloggers, Cedric Scherer etc.**

Thank you! Questions?



Acknowledgements:

- SLUBI • 3Bs • Slides adapted from [RaukR](#) • [GPL-3 License](#)