

Data Visualization with R

Workshop Day 2

Making maps

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MONASH University

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Tuberculosis incidence

The TB data is from the WHO.

```
## # A tibble: 40,800 x 5
##   country     year age_group sex  count
##   <chr>       <dbl> <fct>    <chr> <dbl>
## 1 Afghanistan 1997 15-24     m      10
## 2 Afghanistan 1997 25-34     m      6
## 3 Afghanistan 1997 35-44     m      3
## 4 Afghanistan 1997 45-54     m      5
## 5 Afghanistan 1997 55-64     m      2
## 6 Afghanistan 1997 65-      m      0
## 7 Afghanistan 1997 15-24     f      38
## 8 Afghanistan 1997 25-34     f      36
## 9 Afghanistan 1997 35-44     f      14
## 10 Afghanistan 1997 45-54    f      8
## # ... with 40,790 more rows
```

What is a choropleth map?

Why use a choropleth map?



How do we get a map?

A polygon map of the world can be extracted from the maps package.

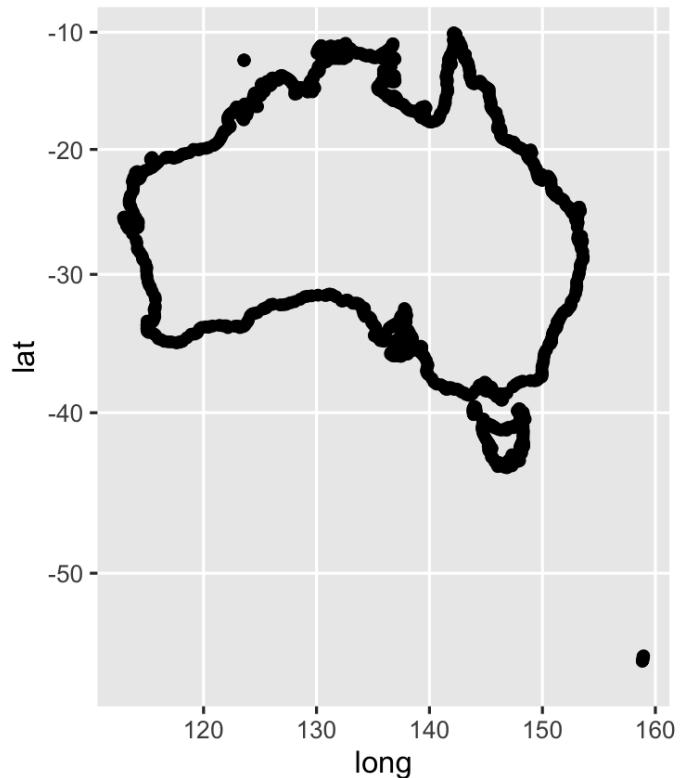
```
world_map <- map_data("world")  
world_map %>%  
  filter(region %in% c("Australia", "New Zealand")) %>%  
  DT::datatable(width=1150, height=100)
```

	long	lat	group	order	region	subregion
1	123.594528198242	-12.4256830215454	133	7115	Australia	Ashmore and Cartier Islands
2	123.595207214355	-12.4359369277954	133	7116	Australia	Ashmore and Cartier Islands
3	123.573150634766	-12.4341802597046	133	7117	Australia	Ashmore and Cartier Islands
4	123.572463989258	-12.4239253997803	133	7118	Australia	Ashmore and Cartier Islands
5	123.594528198242	-12.4256830215454	133	7119	Australia	Ashmore and Cartier Islands
6	158.878799438477	-54.7097625732422	139	7267	Australia	Macquarie Island
7	158.84521484375	-54.7492179870605	139	7268	Australia	Macquarie Island

Maps are basically groups of connected dots

These are the points, defining the country boundary for Australia

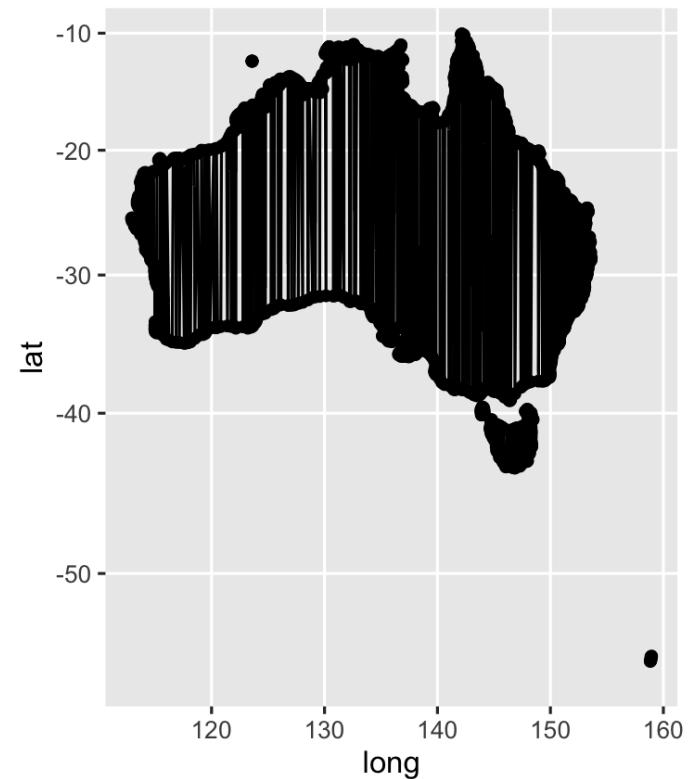
```
oz <- world_map %>%
  filter(region == "Australia")
ggplot(oz, aes(x = long, y = lat)) +
  geom_point() +
  coord_map()
```



Connect the dots

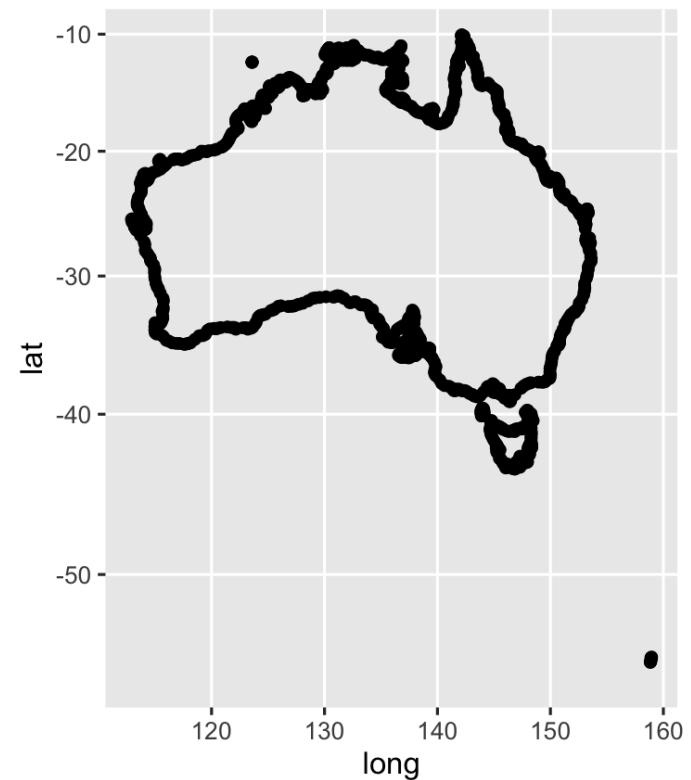
```
ggplot(oz, aes(x = long, y = lat,  
               group = group)) +  
  geom_point() +  
  geom_line() +  
  coord_map()
```

What happened?



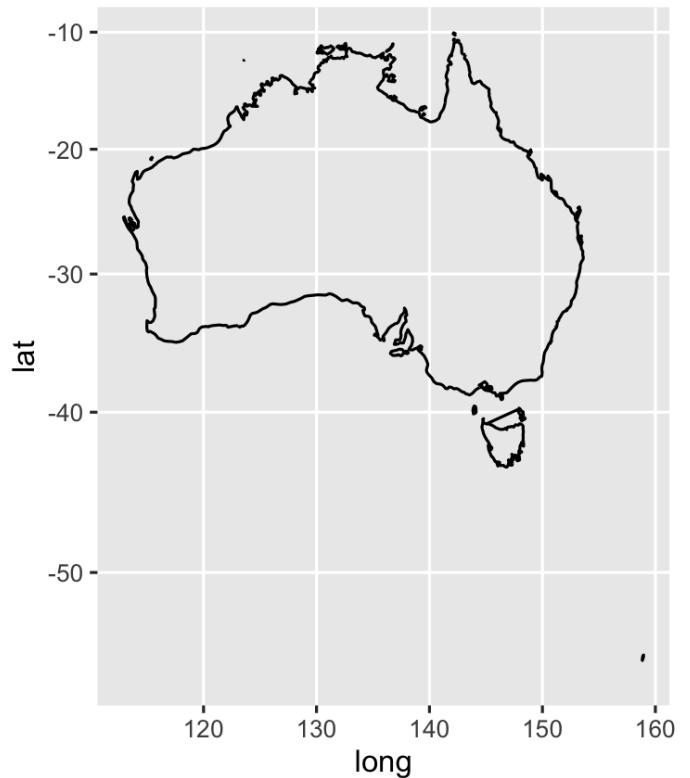
Connect the dots

```
ggplot(oz, aes(x = long, y = lat,  
               group = group)) +  
  geom_point() +  
  geom_path() +  
  coord_map()
```



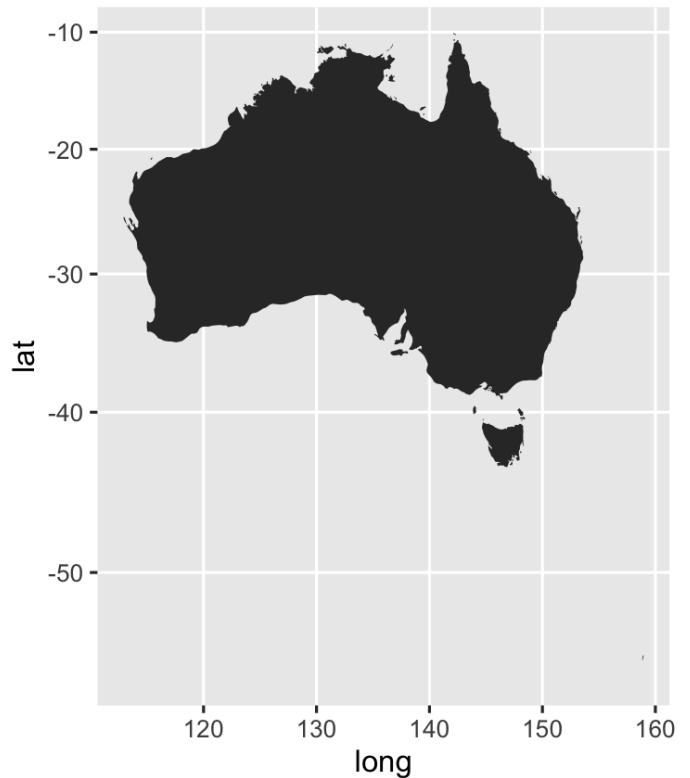
This map doesn't have states and territory connections.

```
ggplot(oz, aes(x = long, y = lat,  
               group = subregion)) +  
  geom_path() +  
  coord_map()
```



We can also plot the map using
geom_polygon, and fill with colour.

```
ggplot(oz, aes(x = long, y = lat,  
               group = group)) +  
  geom_polygon() +  
  coord_map()
```



Using a map theme makes the result look more map like

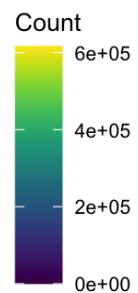
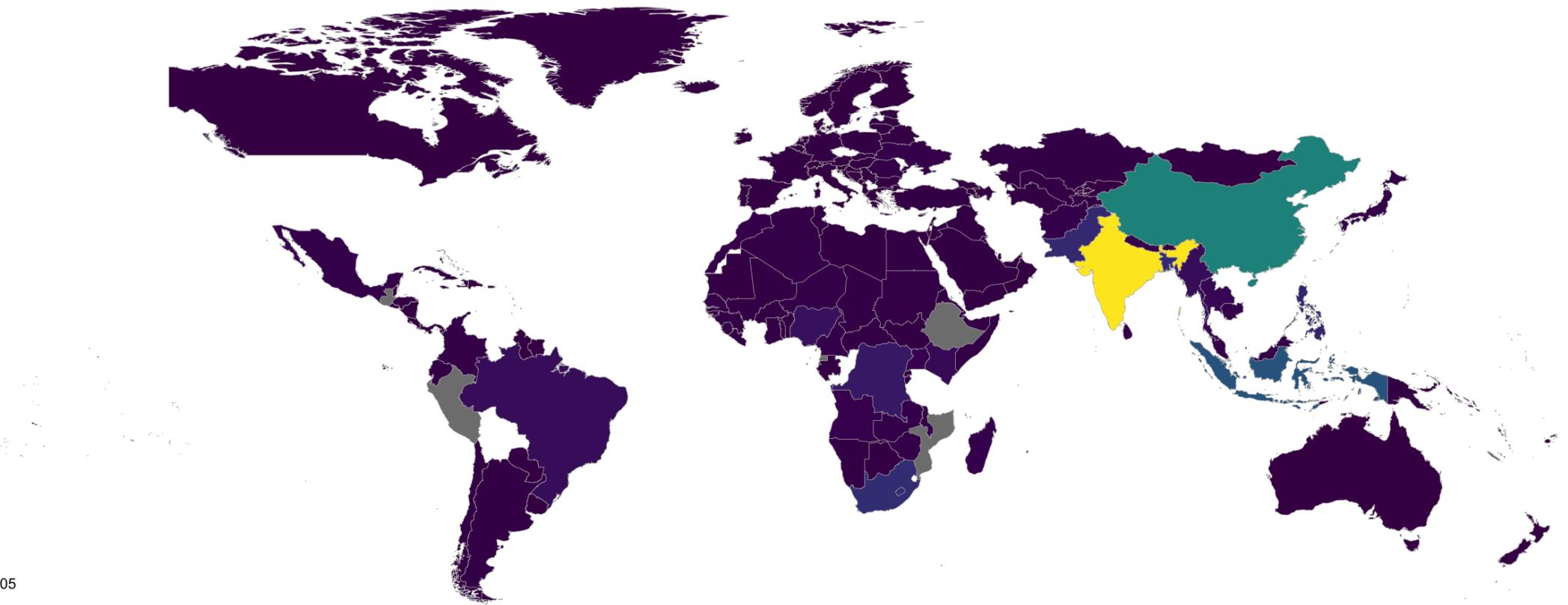
```
ggplot(oz, aes(x = long, y = lat,  
               group = group)) +  
  geom_polygon() +  
  coord_map() +  
  theme_map()
```



Let's make a choropleth map of
tuberculosis

Aggregate counts across sex and age group for 2012

```
tb_2012 <- tb %>%
  filter(year == 2012) %>%
  rename(region = country) %>%
  group_by(region) %>%
  summarise(count = sum(count))
ggplot(tb_2012, aes(map_id = region)) +
  geom_map(aes(fill = count), map = world_map,
           color="grey70", size = 0.1, na.rm = TRUE) +
  expand_limits(x = world_map$long, y = world_map$lat) +
  scale_fill_viridis("Count") +
  theme_map()
```



What happened to the USA? UK?



Check the name matching

```
wm_names <- world_map %>%  
  select(region) %>%  
  distinct()  
  
tb_names <- tb %>%  
  filter(year == 2012) %>%  
  select(country) %>%  
  distinct()  
  
tb_miss_from_wm <- anti_join(tb_names, wm_names,  
                                by=c("country" = "region"))  
  
DT::datatable(tb_miss_from_wm, width = 1150, height = 100)
```

Show 10 entries

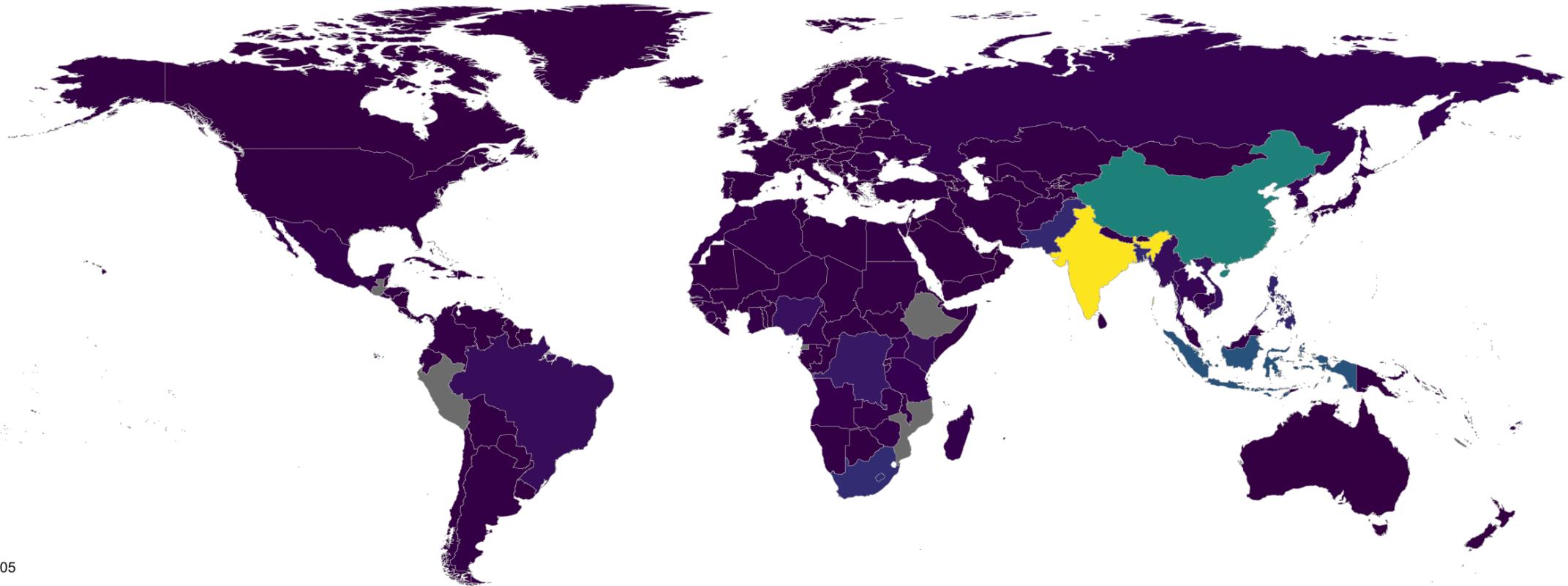
Search:

	country
1	Antigua and Barbuda
2	Bolivia (Plurinational State of)
3	British Virgin Islands
4	Brunei Darussalam

```
tb_fixed <- tb %>%
  mutate(region=recode(country,
    "United States of America" = "USA",
    "United Kingdom of Great Britain and Northern Ireland" = "UK",
    "Russian Federation" = "Russia",
    "Viet Nam" = "Vietnam",
    "Venezuela (Bolivarian Republic of)" = "Venezuela",
    "Bolivia (Plurinational State of)" = "Bolivia",
    "Czechia" = "Czech Republic",
    "Iran (Islamic Republic of)" = "Iran",
    "Iran (Islamic Republic of)" = "Laos",
    "Democratic People's Republic of Korea" = "North Korea",
    "Republic of Korea" = "South Korea",
    "United Republic of Tanzania" = "Tanzania",
    "Congo" = "Republic of Congo"))
```

TRY AGAIN!

```
tb_2012 <- tb_fixed %>%
  filter(year == 2012) %>%
  group_by(region) %>%
  summarise(count = sum(count))
ggplot(tb_2012, aes(map_id = region)) +
  geom_map(aes(fill = count), map = world_map,
           color = "grey70", size = 0.1, na.rm = TRUE) +
  expand_limits(x = world_map$long, y = world_map$lat) +
  scale_fill_viridis("Count") +
  theme_map()
```



Count

6e+05

4e+05

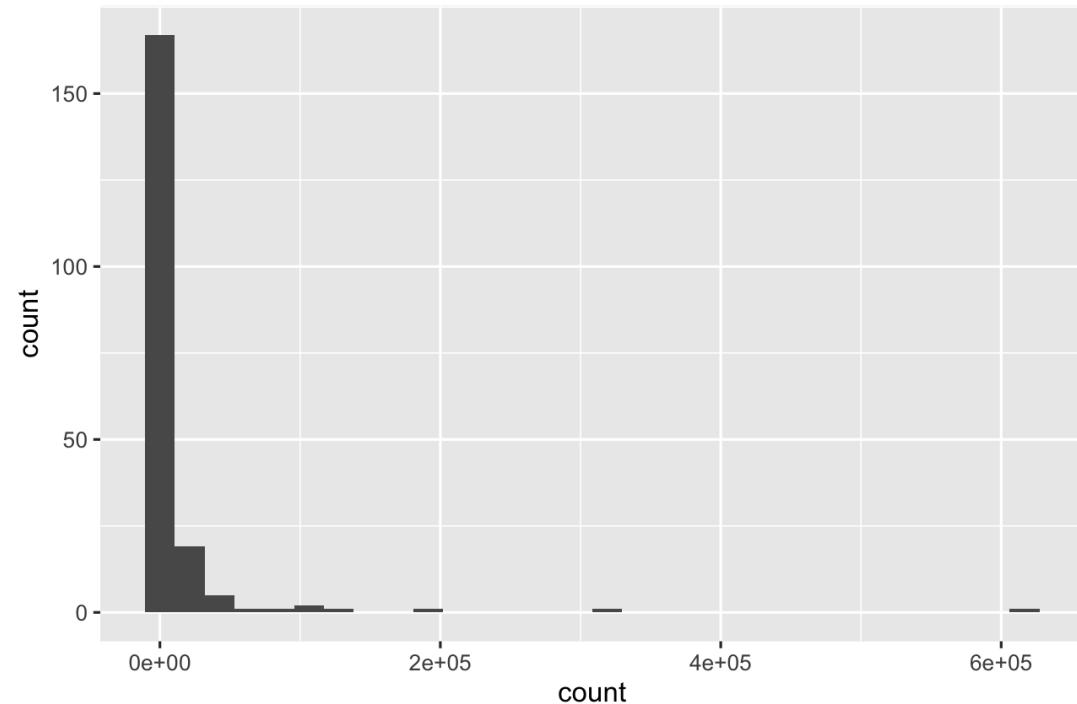
2e+05

0e+00

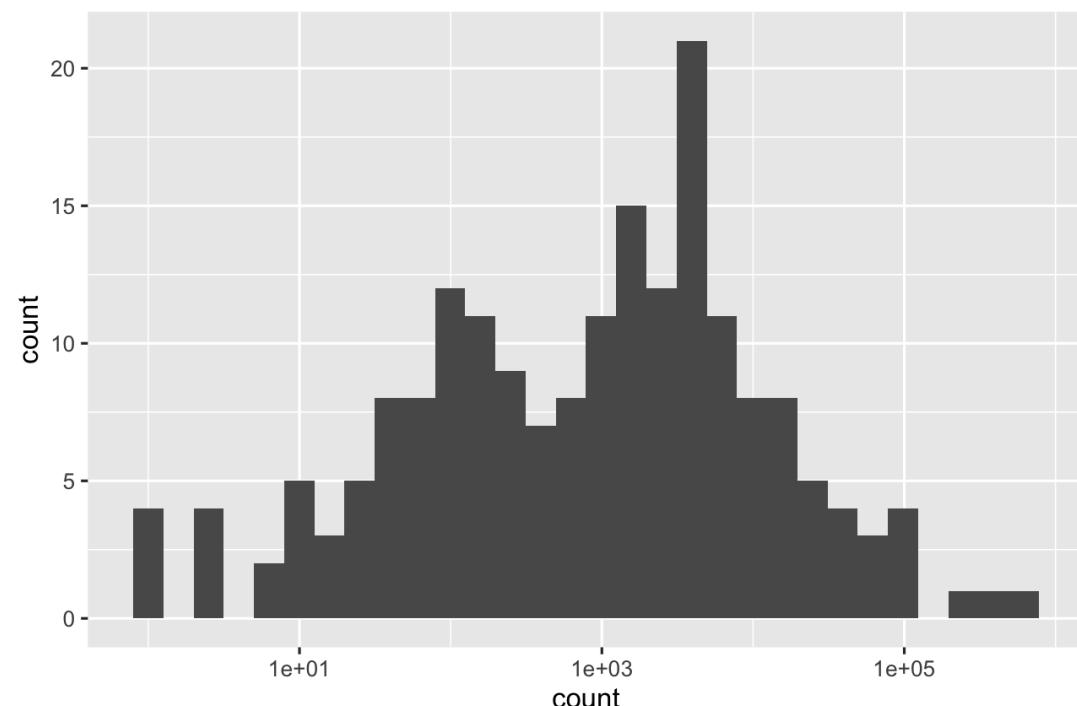
Counts are typically skewed

may be best to symmetrise

```
ggplot(tb_2012, aes(x = count)) +  
  geom_histogram()
```

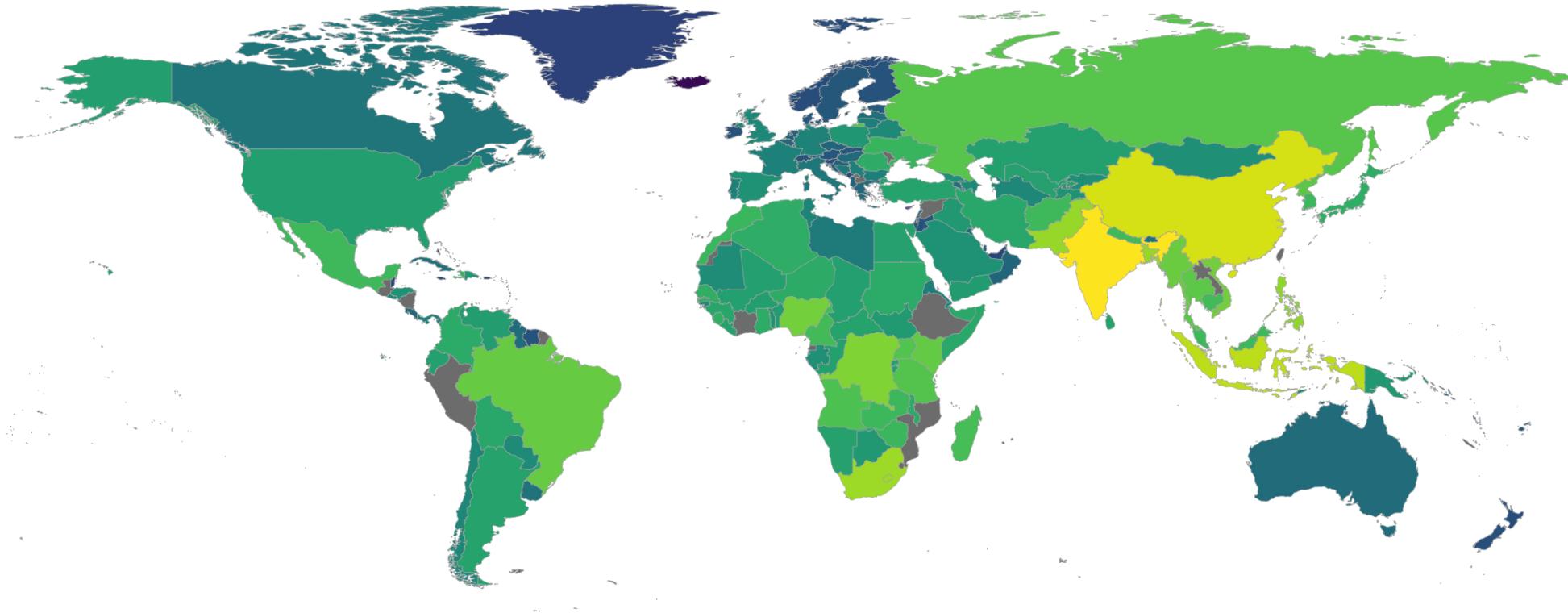


```
ggplot(tb_2012, aes(x = count)) +  
  geom_histogram() +  
  scale_x_log10()
```



geom_polygon can also be used

```
tb_2012_map <- world_map %>% left_join(tb_2012)
ggplot(tb_2012_map, aes(x = long, y = lat, group=group)) +
  geom_polygon(aes(fill = count),
               color="grey70", size = 0.1, na.rm = TRUE) +
  expand_limits(x = world_map$long*1.1, y = world_map$lat*1.1) +
  scale_fill_viridis("Count", trans = "log10") +
  theme_map()
```



Count

Count Range	Color
1 - 100	Dark Purple
100 - 1000	Dark Teal
1000 - 10,000	Dark Green
10,000+ (approx.)	Bright Yellow

Resources

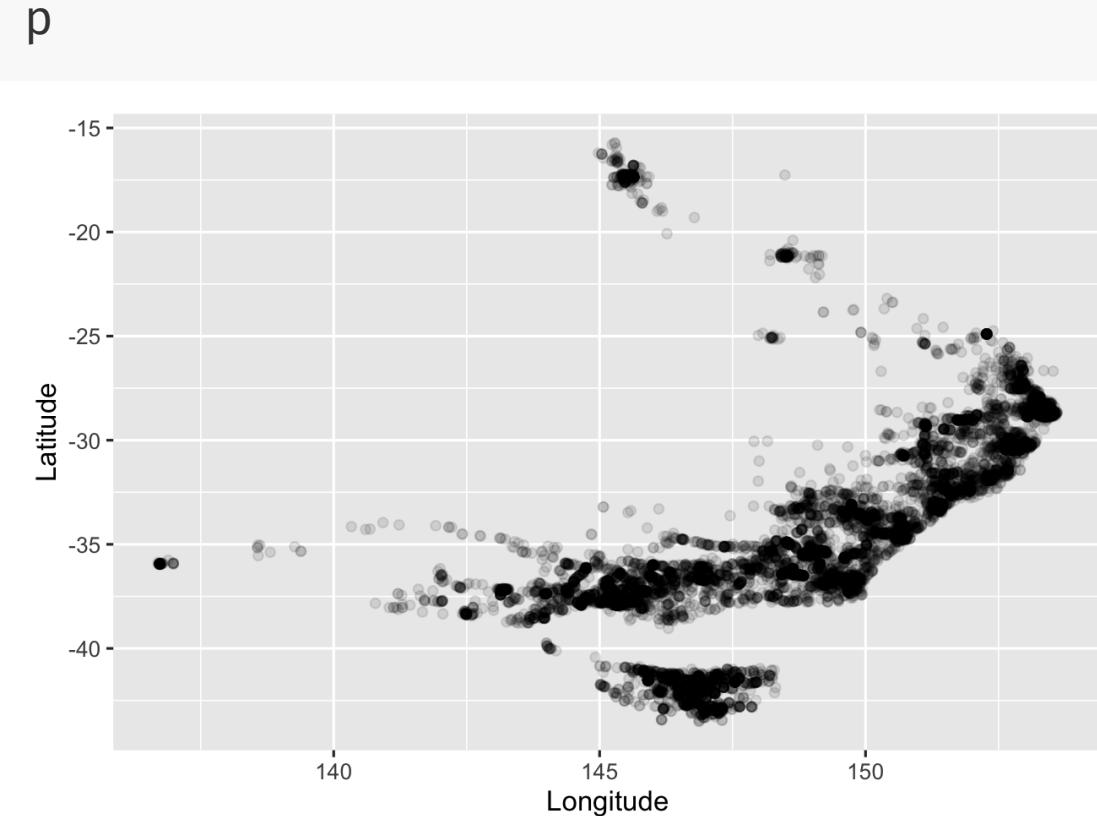
Where to go to get more help on maps

- ozmaps package: <https://github.com/mdsumner/ozmaps>,
<https://mdsumner.github.io/ozmaps/>
- eechidna package <https://docs.ropensci.org/eechidna/>
- <https://www.littlemissdata.com/blog/maps>
- <https://www.r-spatial.org/r/2018/10/25/ggplot2-sf.html>
- <https://www.paulamoraga.com/book-geospatial/sec-spatialdataandCRS.html>

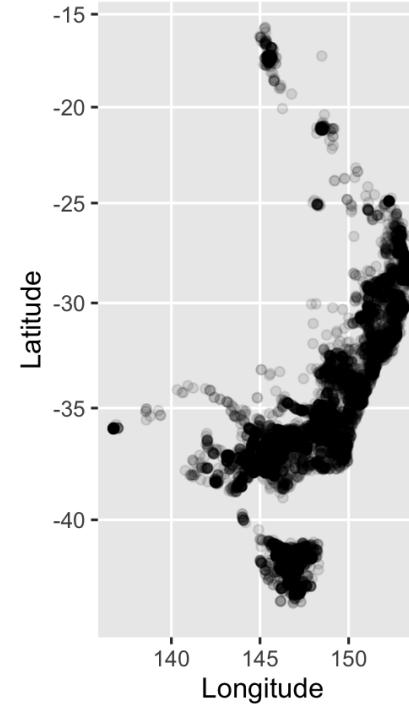
Note: The biggest change when converting spatial objects from a mapping software to a data analysis project is "thinning" the map to make it smaller and efficient to work with.

Reading google maps and overlating data

```
load(here:::here("data/platypus.rda"))
p <- ggplot(platypus) +
  geom_point(aes(x = Longitude, y = Latitude),
             alpha = 0.1)
```



```
p + coord_map()
```



Extract Open Street Map using ggmap

```
oz_bbox <- c(112.9, # min long  
           -45, # min lat  
           159, # max long  
           -10) # max lat  
  
oz_map <- get_map(location = oz_bbox, source = "osm")  
save(oz_map, file=here::here("data/oz_map.rda"))
```

```
load(here::here("data/oz_map.rda"))
ggmap(oz_map) +
  geom_point(data = platypus,
             aes(x = Longitude, y = Latitude),
             alpha = 0.1, colour = "orange")
  theme_map()
```





</> Open day2-exercise-02.Rmd

15:00

Session Information

```
devtools::session_info()
```

```
## - Session info --
##   setting  value
##   version  R version 4.0.1 (2020-06-06)
##   os        macOS Catalina 10.15.6
##   system   x86_64, darwin17.0
##   ui        X11
##   language (EN)
##   collate  en_AU.UTF-8
##   ctype    en_AU.UTF-8
##   tz       Australia/Melbourne
##   date     2020-07-29
##
## - Packages --
##   package      * version    date lib source
## 
```

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