

# UNIVERSITY OF CALIFORNIA, BERKELEY



## STAT 133 FALL 2024

Lecture Notes Taken in the Class

November 4<sup>th</sup>, 2024 – Monday

COURSE: MATH 133 – CONCEPTS IN COMPUTING WITH DATA

INSTRUCTOR: GASTON SANCHEZ

NAME: ALDAN OU

DATE: NOVEMBER 4<sup>TH</sup>, 2024

## ❖ GEOSPATIAL

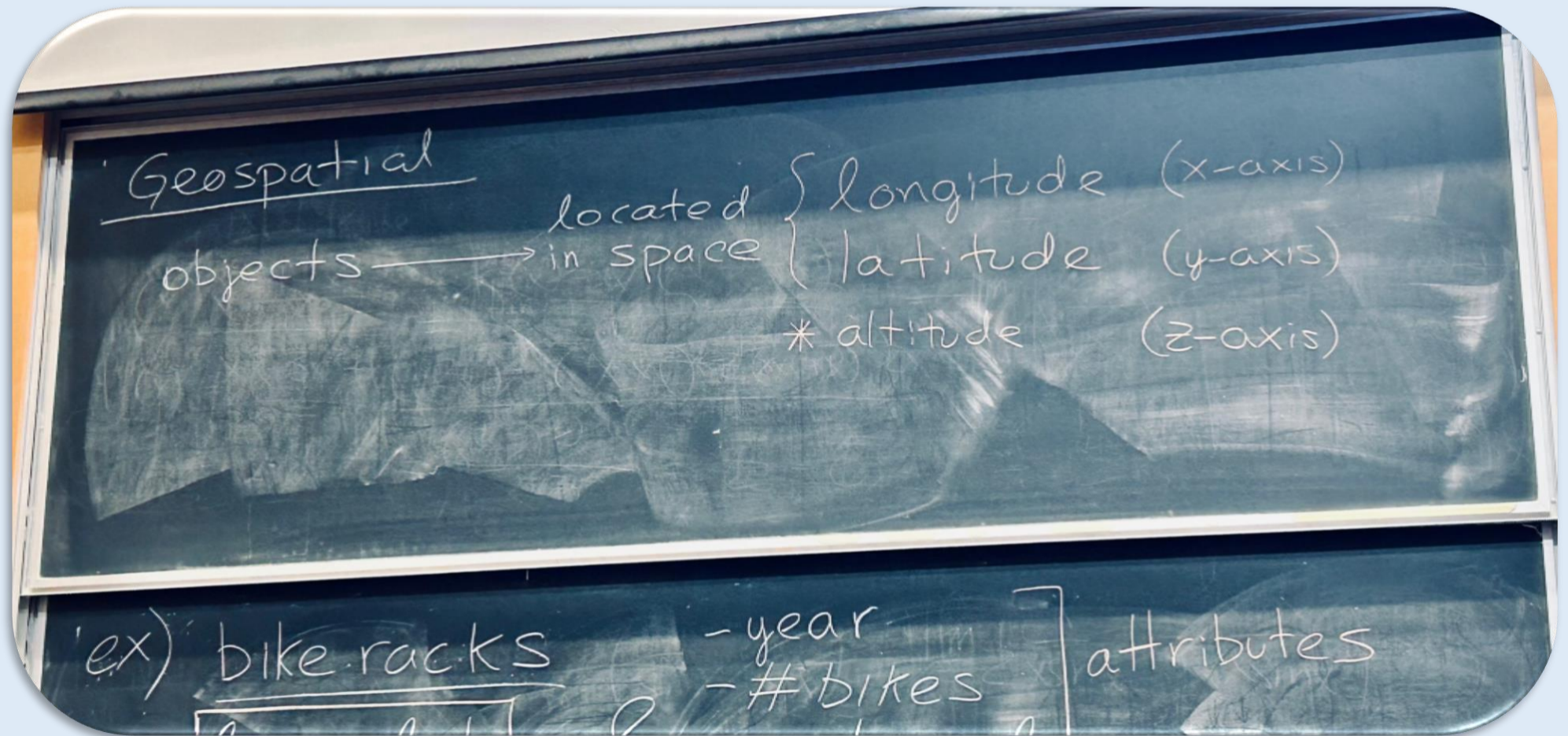
Geospatial refers to objects that are located in space using three primary coordinates.

- Longitude (x-axis): measures east-west position
- Latitude (y-axis): Measures north-south position
- Altitude (z-axis): Measures height / elevation

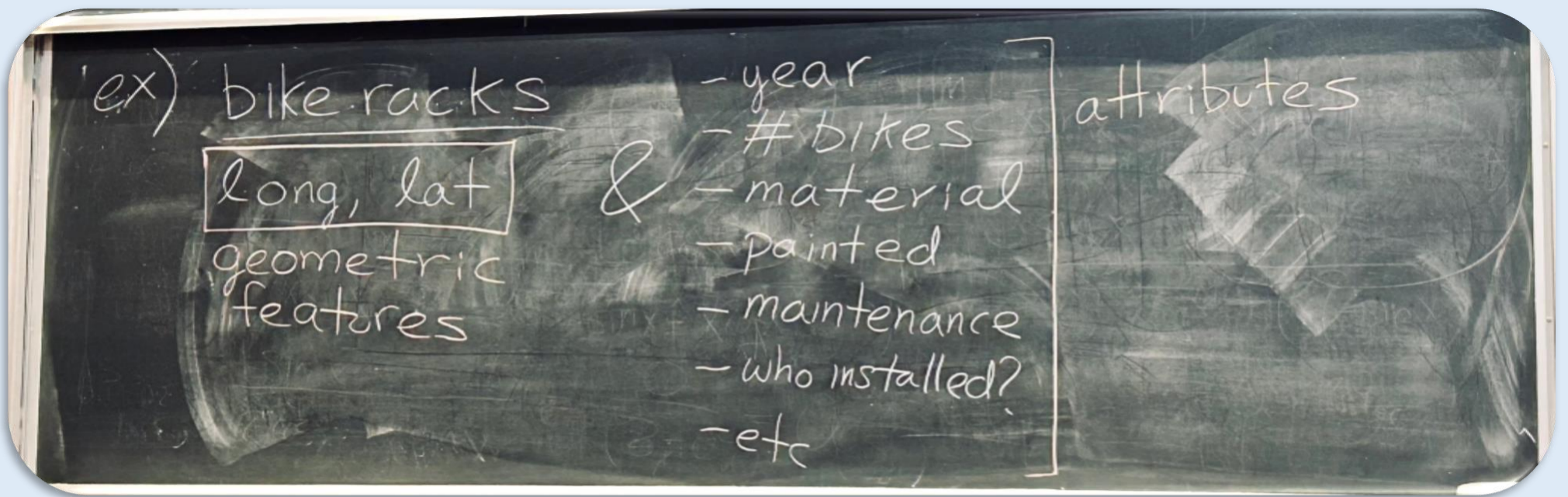
A common application of Geospatial is mapping storm and hurricane data.

1. Locations can be plotted using longitude and latitude coordinates;
2. Paths can be traced using connecting lines;
3. Data can be layered on different types of maps;
  - 1) Coastline maps
  - 2) Country maps
  - 3) Continental maps

The following photos are Prof. Sanchez's lecture notes that were taken in the class.



Example: Bike Racks

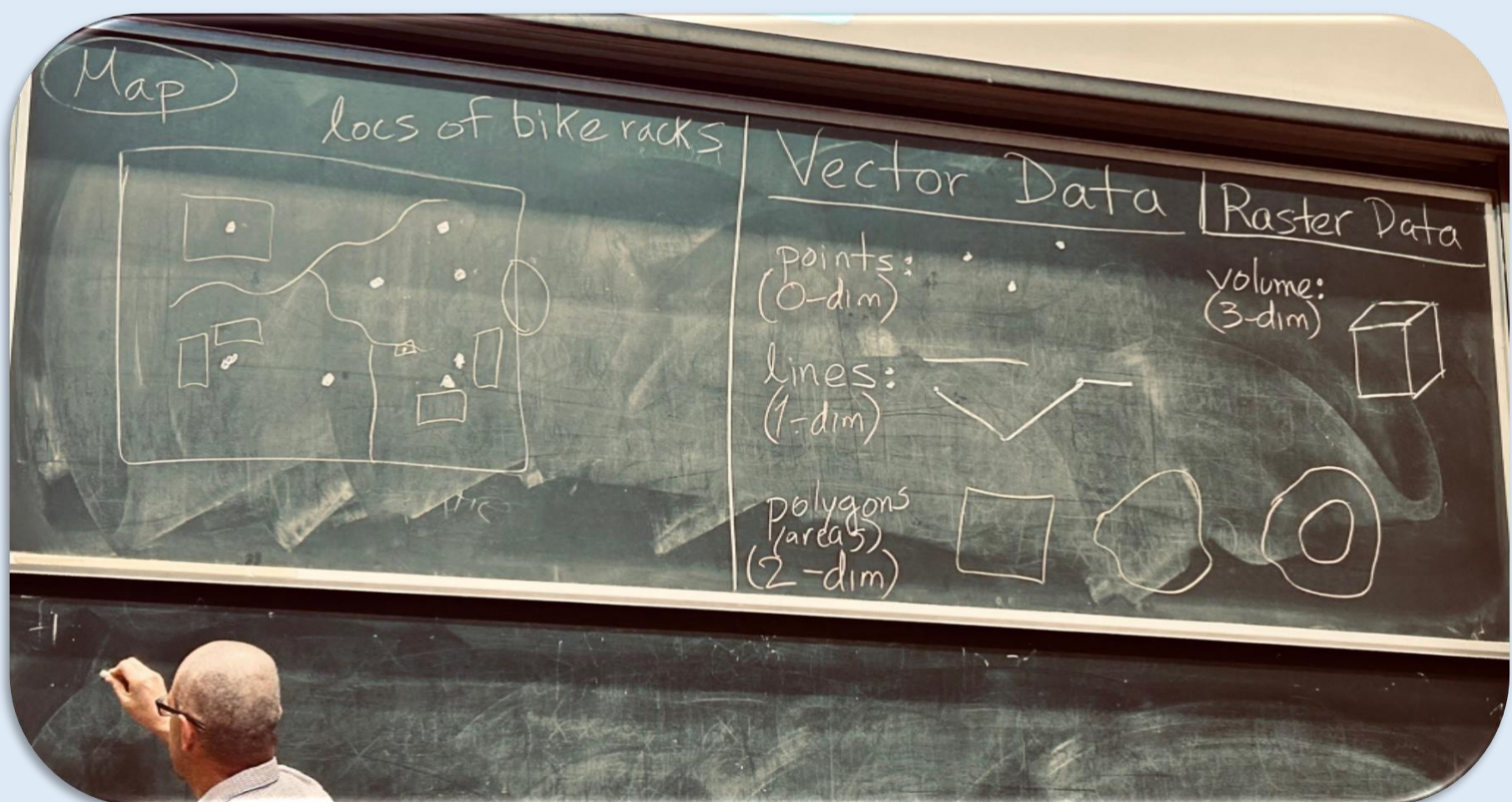


▪ How does this example illustrate the key aspects of geospatial data?

- 1) The first one is the **Spatial Component**. The geometric features (longitude and latitude) that specify where objects are located in physical space.
- 2) The second one is the **Descriptive Component**. The attributes that provide additional information about the object at those locations. Such attributes include several non-spatial characteristics: year of installation, number of bikes, material used, paint condition, and maintenance history...

## ❖ VECTOR DATA

The following photo – Vector Data, is Prof. Sanchez’s lecture notes that were taken in the class.



## ❖ R STUDIO CODING

Now, it's time to do some R coding.

```
{r}
# Loading libraries
library(tidyverse)
library(sf)
library(rnaturalearth)
```

The R package “sf” provides powerful tools for handling geospatial data with several advantages.

- Fast reading and writing of geospatial data
- Enhanced plotting performance
- Compatibility with data frames
- Consistent function naming

```
{r}
# Our first "vector data" set (informal)
storms # data set comes from dplyr
```

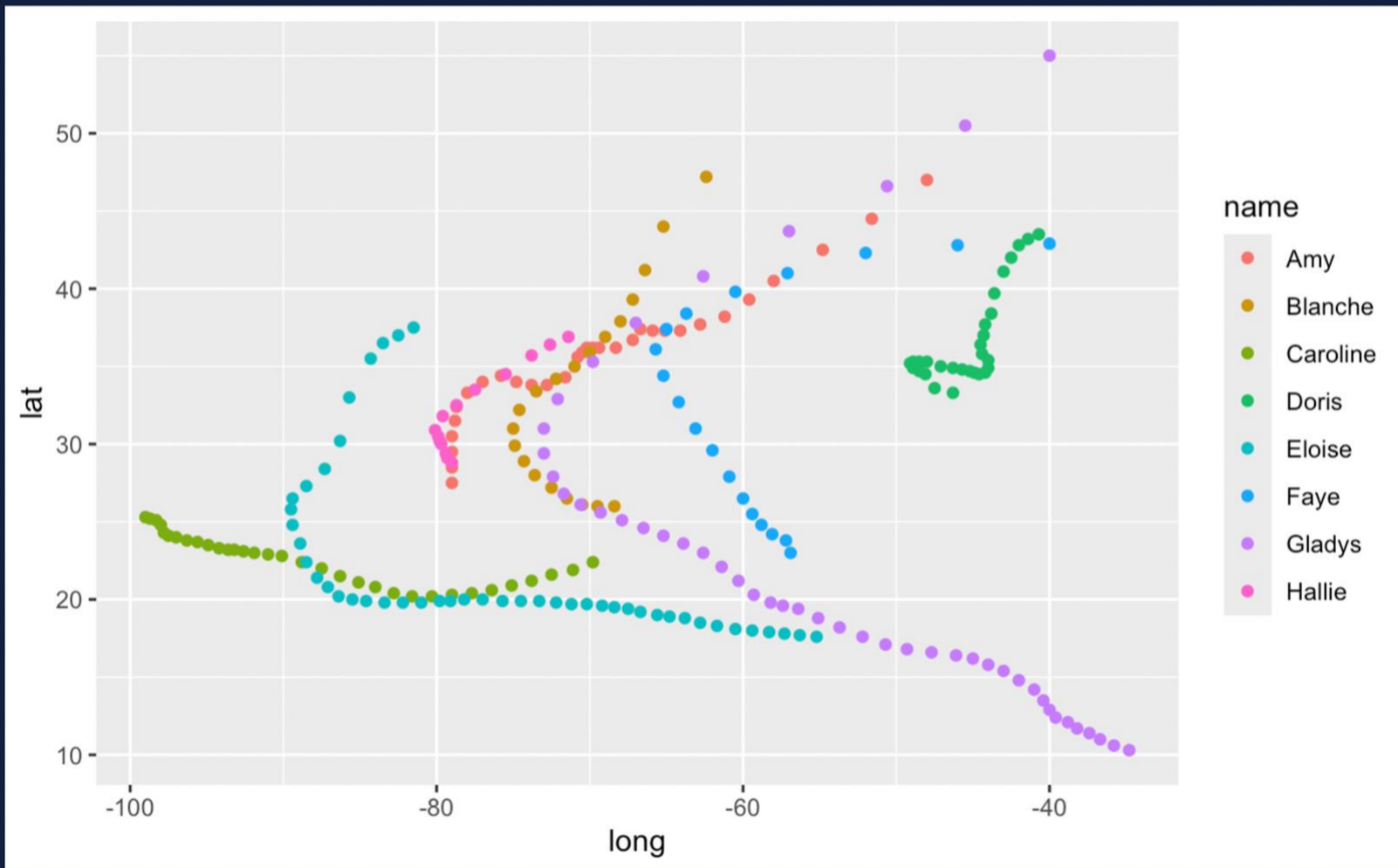
A tibble: 19,537 × 13

name <chr>	year <dbl>	month <dbl>	day <int>	hour <dbl>	lat <dbl>	long <dbl>	status <fctr>	catego... <dbl>	wind <int>
Amy	1975	6	27	0	27.5	-79.0	tropica...	NA	25
Amy	1975	6	27	6	28.5	-79.0	tropica...	NA	25
Amy	1975	6	27	12	29.5	-79.0	tropica...	NA	25
Amy	1975	6	27	18	30.5	-79.0	tropica...	NA	25
Amy	1975	6	28	0	31.5	-78.8	tropica...	NA	25
Amy	1975	6	28	6	32.4	-78.7	tropica...	NA	25
Amy	1975	6	28	12	33.3	-78.0	tropica...	NA	25
Amy	1975	6	28	18	34.0	-77.0	tropica...	NA	30
Amy	1975	6	29	0	34.4	-75.8	tropica...	NA	35
Amy	1975	6	29	6	34.0	-74.8	tropica...	NA	40

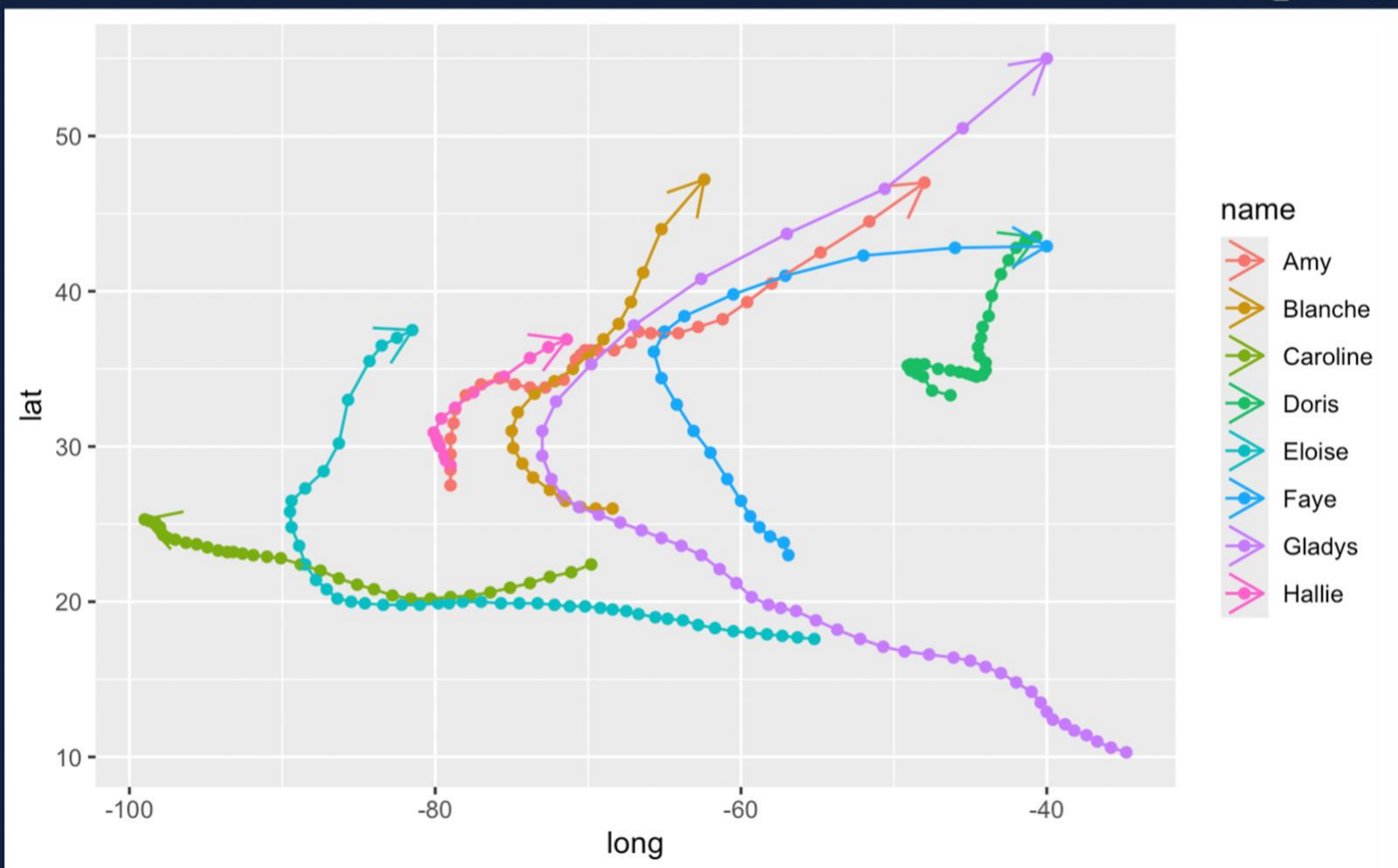
1–10 of 19,537 rows | 1–10 of 13 columns

Previous  2 3 4 5 6 ... 100 Next

```
{r}  
storms75 <- storms %>% filter(year == 1975)  
storms75 %>%  
  ggplot(aes(x = long, y = lat, color = name)) +  
  geom_point()
```



```
{r}  
storms75 %>%  
  ggplot(aes(x = long, y = lat, color = name)) +  
  geom_point() +  
  geom_path(arrow = arrow(length = unit(0.2, units = "in")))
```

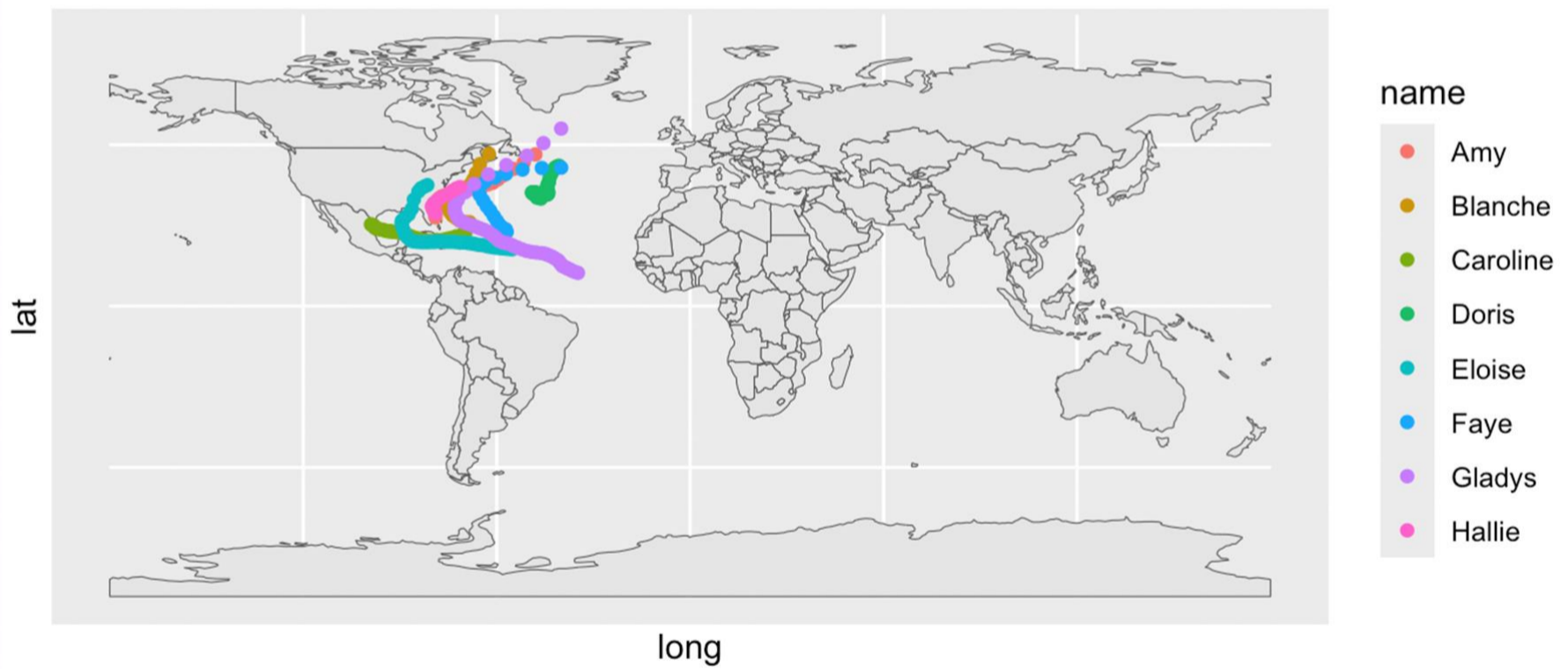


## ❖ MAPS

```
{r}  
world_map = ne_countries(returnclass = "sf")  
class(world_map)
```

```
[1] "sf"          "data.frame"
```

```
{r}  
ggplot() +  
  geom_sf(data = world_map) +  
  geom_point(data = storms75,  
            aes(x = long, y = lat, color = name))
```



```
{r}  
ggplot() +  
  geom_sf(data = world_map) +  
  coord_sf(xlim = c(-110, 0), ylim = c(0, 60)) +  
  geom_point(data = storms75,  
            aes(x = long, y = lat, color = name, size = wind,  
                alpha = 0.5)) +  
  geom_path(data = storms75,  
           aes(x = long, y = lat, color = name))
```

