

Introduction to Vector Data Processing in R

Earth Observation Workshop, Basel

Michael Dorman

Geography and Environmental Development

dorman@post.bgu.ac.il

2018-02-26 - 2018-03-01



Ben-Gurion University of the Negev

Contents

- ▶ This tutorial is a practical introduction to working with **vector layers** in R using the **sf package**
- ▶ **Slides and code** available on [GitHub](#)
- ▶ **Data** available on [Dropbox](#)

Contents

- ▶ Part I: Intro to sf
 - ▶ Package structure
 - ▶ Creating layer from scratch
 - ▶ Extracting layer components
 - ▶ Interactive mapping
- ▶ Part II: Advanced sf
 - ▶ Creating layer from table
 - ▶ Creating layer from file
 - ▶ Plotting
 - ▶ Reprojection
 - ▶ Subsetting
 - ▶ Geometric calculations
 - ▶ Join by attributes
 - ▶ Spatial join
 - ▶ Visualization with ggplot2

Requirements

- ▶ Several packages **need to be installed** to run code examples -

```
install.packages("sf")
install.packages("mapview")
install.packages("units")
install.packages("dplyr")
devtools::install_github("tidyverse/ggplot2")
```

- ▶ Set to **working directory** with the **data** -

- ▶ cb_2015_us_state_5m - **US states (Shapefile)**
- ▶ hurmjrl020 - **Storm tracks (Shapefile)**

Part I: Intro to sf

Vector layers in R: package sf

- ▶ sf is a relatively new (2016-) R package for **handling vector layers in R**
- ▶ In the long-term, sf will replace rgdal (2003-), sp (2005-), and rgeos (2011-)
- ▶ The main innovation in sf is a complete implementation of the **Simple Features** standard
- ▶ Since 2003, Simple Features have been widely implemented in **spatial databases** (such as **PostGIS**), commercial GIS (e.g., **ESRI ArcGIS**) and forms the vector data basis for libraries such as GDAL
- ▶ When working with spatial databases, Simple Features are commonly specified as **Well Known Text (WKT)**
- ▶ A subset of simple features forms the **GeoJSON** standard

Vector layers in R: package sf

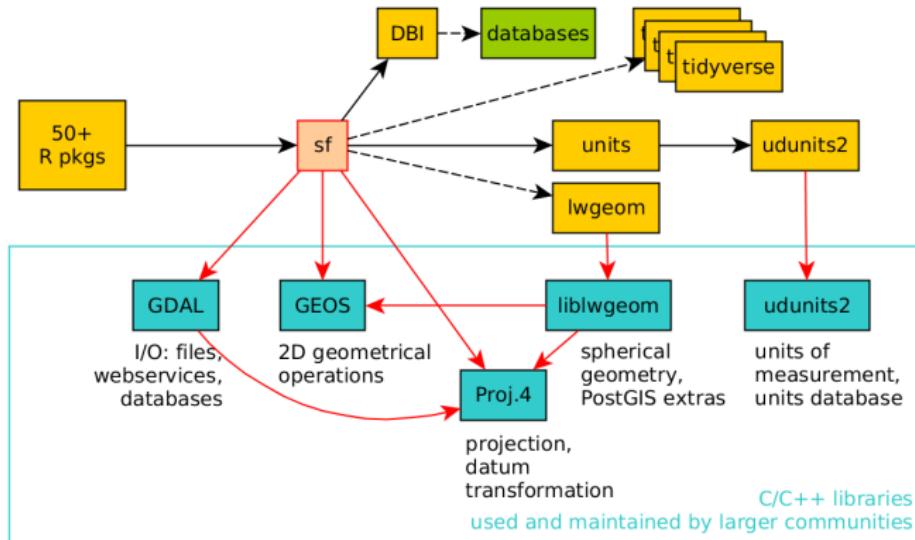


Figure 1: sf package dependencies¹

¹https://github.com/edzer/rstudio_conf

Vector layers in R: package sf

- ▶ The sf class extends the data.frame class to include a **geometry** column
- ▶ This is similar to the way that **spatial databases** are structured

## Simple feature collection with 100 features and 6 fields						
## geometry type: MULTIPOLYGON						
## dimension: XY						
## bbox: xmin: -84.32385 ymin: 33.88199 xmax: -75.45698 ymax: 36.58965						
## epsg (SRID): 4267						
## proj4string: +proj=longlat +datum=NAD27 +no_defs						
## precision: double (default; no precision model)						
## First 3 features:						
## BIR74 SID74 NWBIR74 BIR79 SID79 NWBIR79						
## 1	1091	1	10	1364	0	19
## 2	487	0	10	542	3	12
## 3	3188	5	208	3616	6	260

Simple feature

Simple feature geometry list-column (sf)

Simple feature geometry (sf)

Figure 2: Structure of an sf object²

²<https://cran.r-project.org/web/packages/sf/vignettes/sf1.html>

Vector layers in R: package sf

- The **sf** class is actually a hierarchical structure composed of three classes -
 - sf** - Vector **layer** object, a table (`data.frame`) with one or more attribute columns and one geometry column
 - sfc** - Geometric part of the vector layer, the **geometry column**
 - sfg** - **Geometry** of an individual simple feature

## Simple feature collection with 100 features and 6 fields					
## geometry type: MULTIPOLYGON					
## dimension: XY					
## bbox: xmin: -84.32385 ymin: 33.88199 xmax: -75.45698 ymax: 36.58965					
## epsg (SRID): 4267					
## proj4string: +proj=longlat +datum=NAD27 +no_defs					
## precision: double (default; no precision model)					
## First 3 features:					
## BIR74 SID74 NWBIR74 BIR79 SID79 NWBIR79					
## 1	1091	1	10	1364	0
## 2	487	0	10	542	3
## 3	3188	5	208	3616	6

Simple feature

Simple feature geometry list-column (sfc)

Simple feature geometry (sfg)

Creating layers from scratch

- ▶ As mentioned above, the main classes in the sf package are -
 - ▶ **sfg** - geometry
 - ▶ **sfc** - geometry column
 - ▶ **sf** - layer
- ▶ Let's create an object for each of these classes to learn more about them
- ▶ First load the sf package -

```
library(sf)
## Linking to GEOS 3.5.1, GDAL 2.2.2, proj.4 4.9.2
```

Geometry (`sfg`)

- ▶ Objects of class `sfg (geometry)` can be created using the appropriate function for each geometry type -
 - ▶ `st_point`
 - ▶ `st_multipoint`
 - ▶ `st_linestring`
 - ▶ `st_multilinestring`
 - ▶ `st_polygon`
 - ▶ `st_multipolygon`
 - ▶ `st_geometrycollection`
- ▶ From coordinates passed as -
 - ▶ **numeric vectors** - POINT
 - ▶ matrix objects - MULTIPOINT or LINESTRING
 - ▶ list objects - All other geometries

Geometry (sfg)

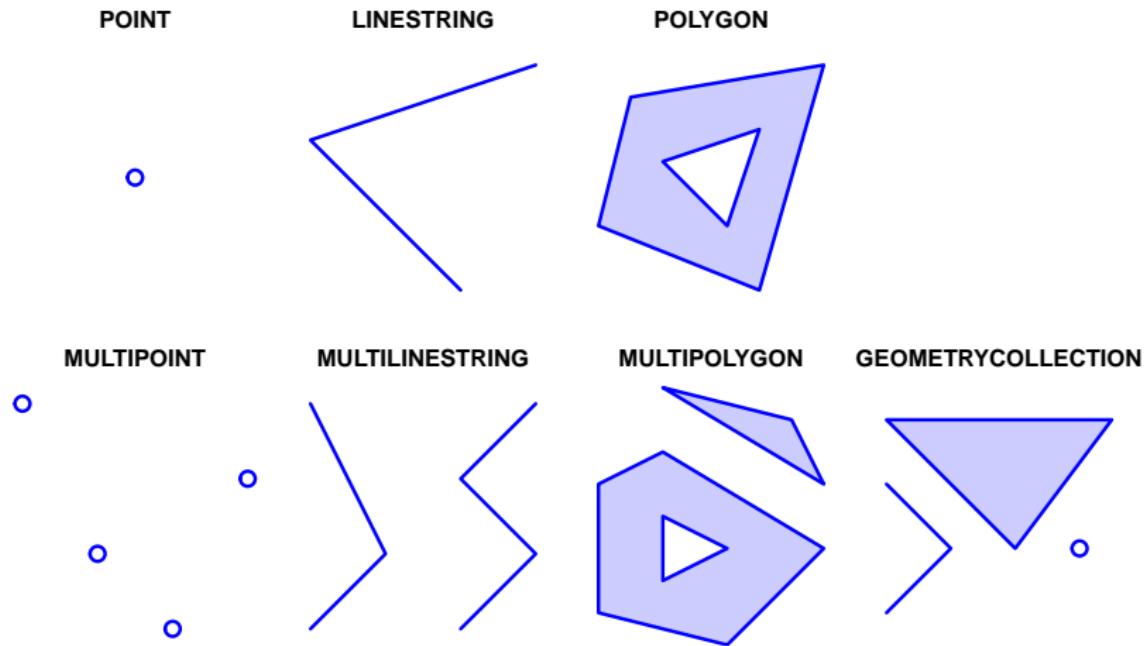


Figure 3: Simple feature geometry (sfg) types in package sf

Geometry (`sfg`)

- ▶ For example, we can create an object named `pnt1` representing a `POINT` geometry with `st_point` -

```
pnt1 = st_point(c(34.812831, 31.260284))
```

- ▶ Printing the object in the console gives the **WKT** representation -

```
pnt1
## POINT (34.81283 31.26028)
```

Geometry (**sfg**)

- ▶ Note the class definition of an **sfg** (**geometry**) object is actually composed of three parts -
 - ▶ XY - The dimensions type (XY, XYZ, XYM or XYZM)
 - ▶ POINT - The geometry type (POINT, MULTIPOLYGON, etc.)
 - ▶ **sfg** - The general class (**sfg** = Simple Feature Geometry)
- ▶ For example, the pnt1 object has geometry POINT and dimensions XY -

```
class(pnt1)
## [1] "XY"     "POINT"  "sfg"
```

Geometry column (**sfc**)

- ▶ Let's create a second object named `pnt2`, representing a different point -

```
pnt2 = st_point(c(34.798443, 31.243288))
```

- ▶ Geometry objects (`sfg`) can be *collected* into an **sfc** (**geometry column**) object
- ▶ This is done with function `st_sfc`

Geometry column (**sfc**)

- ▶ **Geometry column** objects contain a **Coordinate Reference System (CRS)** definition, specified with `crs`
- ▶ Two types of arguments are accepted -
 - ▶ A **PROJ.4** definition ("`+proj=longlat +datum=WGS84`")
 - ▶ An **EPSG** code (4326)
- ▶ Let's combine the two POINT geometries `pnt1` and `pnt2` into an **sfc (geometry column)** object `pnt` -

```
geom = st_sfc(pnt1, pnt2, crs = 4326)
```

Geometry column (**sfc**)

- ▶ Here is a summary of the resulting geometry column -

```
geom
```

```
## Geometry set for 2 features
## geometry type:  POINT
## dimension:      XY
## bbox:            xmin: 34.79844 ymin: 31.24329 xmax: 34.81283 ymax: 31.26028
## epsg (SRID):    4326
## proj4string:    +proj=longlat +datum=WGS84 +no_defs
## POINT (34.81283 31.26028)
## POINT (34.79844 31.24329)
```

Layer (`sf`)

- ▶ A `sfc` (**geometry column**) can be combined with non-spatial columns (*attributes*), resulting in an `sf` (**layer**) object
- ▶ In our case the two points in the `sfc` (**geometry column**) `pnt` represent the location of the two railway stations in Beer-Sheva
- ▶ Let's create a `data.frame` with a `name` column specifying **station name**
- ▶ Note that the **order** of attributes must match the order of the geometries

Layer (**sf**)

- ▶ Creating the attribute table -

```
dat = data.frame(  
  name = c("Beer-Sheva North", "Beer-Sheva Center")  
)
```

```
dat  
##           name  
## 1 Beer-Sheva North  
## 2 Beer-Sheva Center
```

Layer (**sf**)

- ▶ And combining the **attribute table** with the **geometry column** -

```
pnt = st_sf(dat, geom)
pnt
```

```
## Simple feature collection with 2 features and 1 field
## geometry type: POINT
## dimension: XY
## bbox: xmin: 34.79844 ymin: 31.24329 xmax: 34.81283 ymax: 31.26028
## epsg (SRID): 4326
## proj4string: +proj=longlat +datum=WGS84 +no_defs
##          name           geom
## 1 Beer-Sheva North POINT (34.81283 31.26028)
## 2 Beer-Sheva Center POINT (34.79844 31.24329)
```

Extracting layer components

- ▶ In the last few slides we -
 - ▶ Started from raw **coordinates**
 - ▶ Converted them to **geometry** objects (sfg)
 - ▶ Combined the geometries to a **geometry column** (sfc)
 - ▶ Added attributes to the geometry column to get a **layer** (sf)
- ▶ Sometimes we are interested in the opposite process
- ▶ We may need to extract the simpler components (**geometry**, **attributes**, **coordinates**) from an existing layer

Extracting layer components

- ▶ The **sfc (geometry column)** component can be extracted from an **sf (layer)** object using function **st_geometry** -

```
st_geometry(pnt)
```

```
## Geometry set for 2 features
## geometry type: POINT
## dimension: XY
## bbox:           xmin: 34.79844 ymin: 31.24329 xmax: 34.81283 ymax: 31.26028
## epsg (SRID):   4326
## proj4string:   +proj=longlat +datum=WGS84 +no_defs
## POINT (34.81283 31.26028)
## POINT (34.79844 31.24329)
```

Extracting layer components

- ▶ The non-spatial columns of an `sf` (**layer**), i.e. the **attribute table**, can be extracted from an `sf` object using function `st_set_geometry` and `NULL` -

```
st_set_geometry(pnt, NULL)
##                 name
## 1 Beer-Sheva North
## 2 Beer-Sheva Center
```

Extracting layer components

- ▶ The **coordinates** (matrix object) of **sf**, **sfc** or **sfg** objects can be obtained with function **st_coordinates** -

```
st_coordinates(pnt)
##           X           Y
## 1 34.81283 31.26028
## 2 34.79844 31.24329
```

Interactive mapping with mapview

- ▶ Function `mapview` is useful for inspecting spatial data -

```
library(mapview)  
mapview(pnt)
```

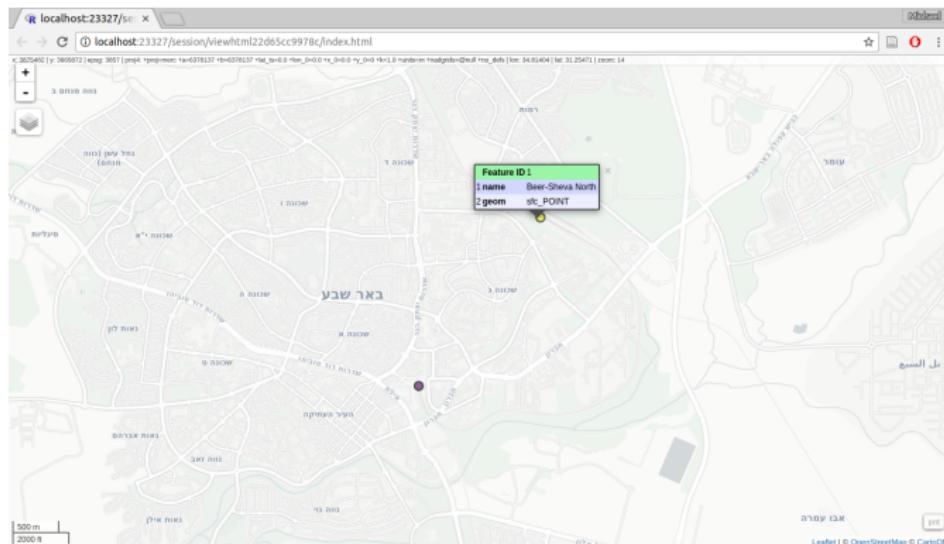


Figure 4: Intractive map of pnt layer

Part II: Advanced sf

Creating point layer from table

- ▶ A common way of creating a point layer is to transform a **table** which has **X and Y coordinate columns**
- ▶ Function `st_as_sf` can convert a table (`data.frame`) into a point layer (`sf`)
- ▶ In `st_as_sf` we specify -
 - ▶ `x` - The `data.frame` to be converted
 - ▶ `coords` - **Columns names** with the **coordinates (X, Y)**
 - ▶ `crs` - The **CRS** (NA if left unspecified)

Creating point layer from table

- ▶ For example, R has a built-in data.frame named `quakes`
- ▶ `quakes` gives the locations of **1000 earthquakes near Fiji**
- ▶ See `?quakes` for more information

```
head(quakes)
##      lat    long depth mag stations
## 1 -20.42 181.62   562 4.8       41
## 2 -20.62 181.03   650 4.2       15
## 3 -26.00 184.10     42 5.4       43
## 4 -17.97 181.66   626 4.1       19
## 5 -20.42 181.96   649 4.0       11
## 6 -19.68 184.31   195 4.0       12
```

Creating point layer from table

- ▶ The quakes table has two columns `long` and `lat` specifying the **longitude** and **latitude** of each earthquake
- ▶ We can **convert** the table into a point layer with `st_as_sf`

```
quakes_pnt = st_as_sf(  
  x = quakes,  
  coords = c("long", "lat"),  
  crs = 4326  
)
```

- ▶ Note that the order of `coords` corresponds to X-Y!

Creating point layer from table

quakes_pnt

```
## Simple feature collection with 1000 features and 3 fields
## geometry type: POINT
## dimension: XY
## bbox: xmin: 165.67 ymin: -38.59 xmax: 188.13 ymax: -10.72
## epsg (SRID): 4326
## proj4string: +proj=longlat +datum=WGS84 +no_defs
## First 10 features:
##   depth mag stations      geometry
## 1 562 4.8      41 POINT (181.62 -20.42)
## 2 650 4.2      15 POINT (181.03 -20.62)
## 3 42 5.4       43 POINT (184.1 -26)
## 4 626 4.1      19 POINT (181.66 -17.97)
## 5 649 4.0      11 POINT (181.96 -20.42)
## 6 195 4.0      12 POINT (184.31 -19.68)
## 7 82 4.8       43 POINT (166.1 -11.7)
## 8 194 4.4      15 POINT (181.93 -28.11)
## 9 211 4.7      35 POINT (181.74 -28.74)
## 10 622 4.3     19 POINT (179.59 -17.47)
```

Creating point layer from table

- ▶ When **plotting** an sf object we get multiple small maps, one for each attribute
- ▶ This can be useful to quickly examine the **types of spatial variation** in our data

```
plot(quakes_pnt)
```

Creating point layer from table

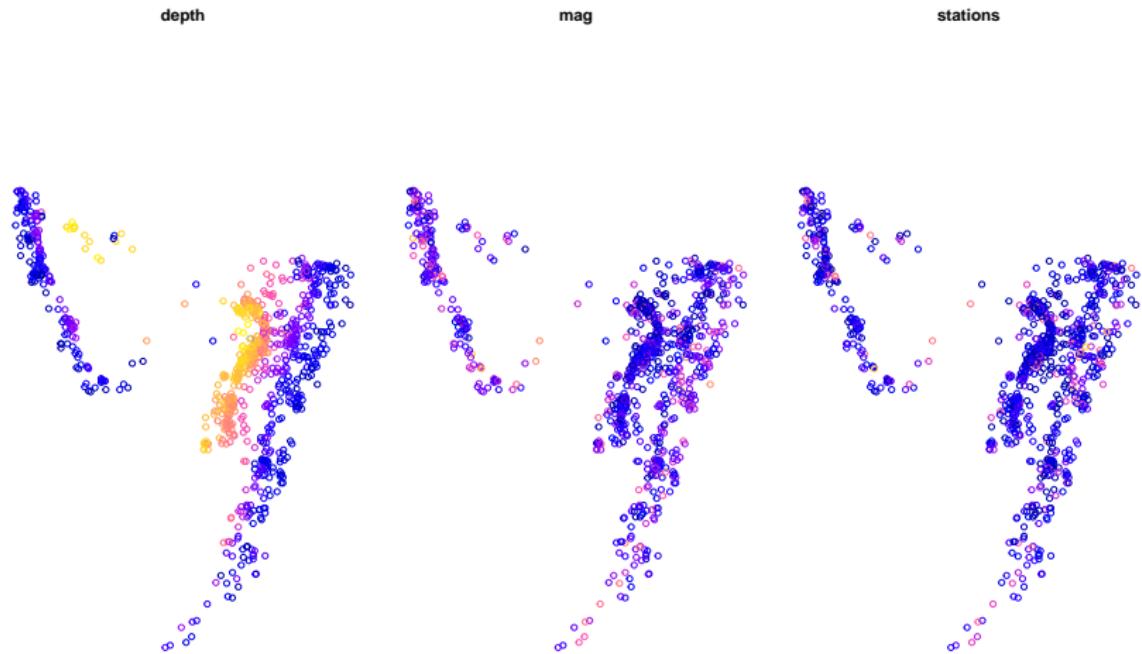


Figure 5: Point layer based on the quakes table

Reading layers into R

- ▶ Here are the driver names of some commonly used **vector layer formats** that can be read into R -
 - ▶ ESRI Shapefile
 - ▶ GeoJSON
 - ▶ GPX
 - ▶ KML
- ▶ Note that it is also possible to read / write to **spatial databases** such as -
 - ▶ PostgreSQL/PostGIS
 - ▶ SQLite/Spatialite
- ▶ For complete list of **available drivers** -

```
View(st_drivers(what = "vector"))
```

Reading layers into R

- ▶ In the following examples, we will use two vector layers -
 - ▶ **US state borders**
 - ▶ **Storm tracks**
- ▶ We will import both from **Shapefiles**
 - ▶ Download a ZIP archive with both layers from [here](#)
 - ▶ Extract the file contents into a new directory
 - ▶ Use `setwd` to change the **Working Directory**

```
setwd("C:/Tutorials/sf")
```

Reading layers into R

- ▶ Next we use `st_read` function to **read the layer**
- ▶ In case the Shapefile is located in the Working Directory we just need to specify the **name of the shp file**
- ▶ We can also specify `stringsAsFactors = FALSE` to **avoid conversion** of character to factor

```
states = st_read(  
  dsn = "cb_2015_us_state_5m.shp",  
  stringsAsFactors = FALSE  
)  
  
tracks = st_read(  
  dsn = "hurmjrl020.shp",  
  stringsAsFactors = FALSE  
)
```

Basic plotting

- ▶ states is a **polygonal** layer containing **US states borders**
- ▶ It has **49 features** and **1 attribute** -
 - ▶ state = State name

```
dim(states)
## [1] 49  2
```

```
plot(states)
```

Basic plotting

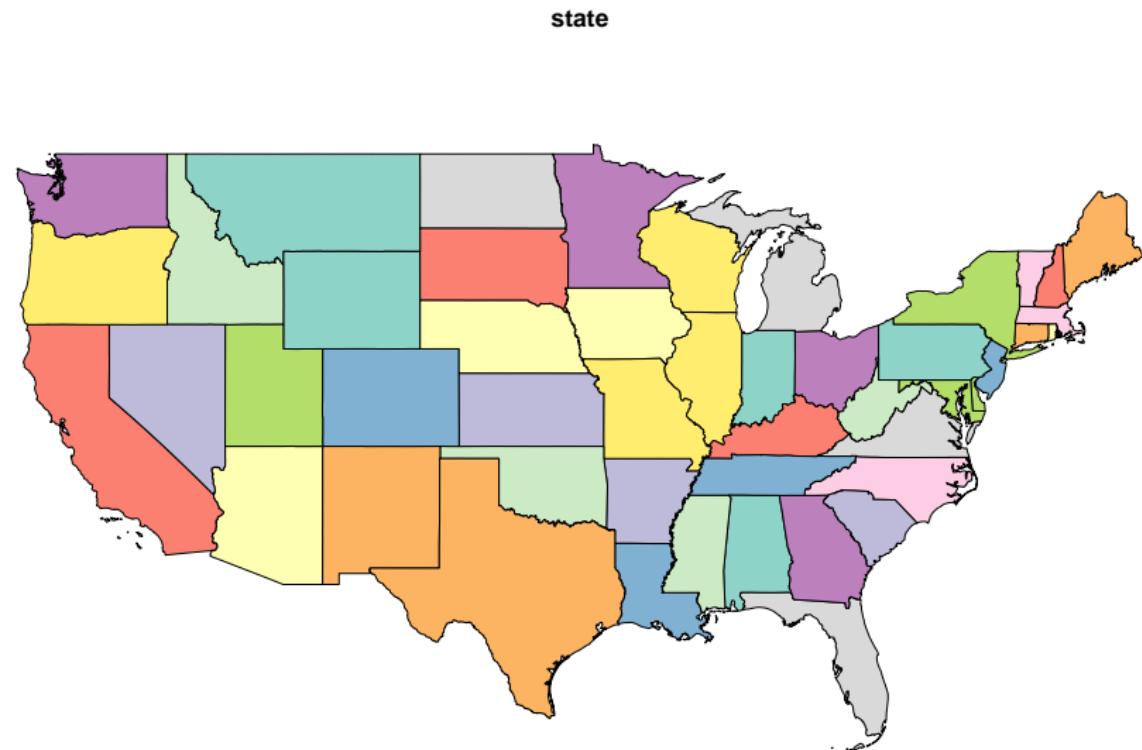


Figure 6: The states layer

Basic plotting

- ▶ tracks is a **line** layer with **storm trajectories**
- ▶ Each feature represents a storm **segment**
- ▶ It has **4056 features** and **7 attributes** -
 - ▶ btid = Track ID
 - ▶ year = Year
 - ▶ month = Month
 - ▶ day = Day
 - ▶ name = Storm name
 - ▶ wind_mph = Wind speed (miles / hour)
 - ▶ category = Storm category (e.g. H5 = Category 5 Hurricane)

```
dim(tracks)
## [1] 4056      8
```

```
plot(tracks)
```

Basic plotting

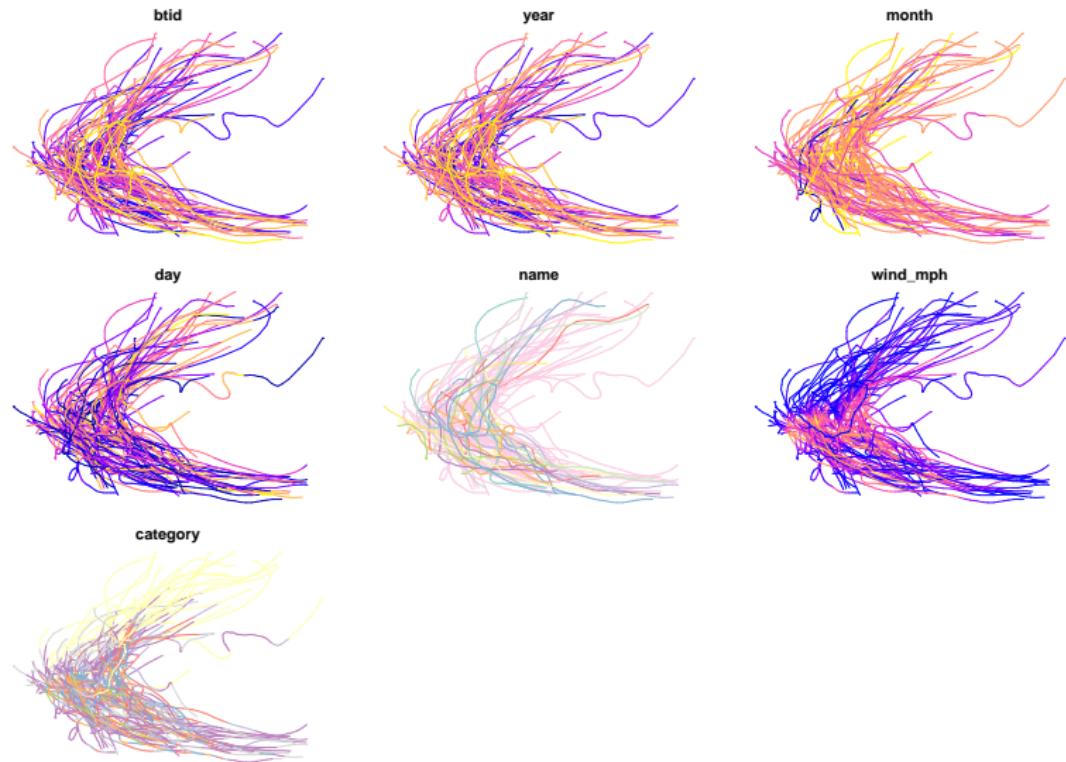


Figure 7: The tracks layer

Basic plotting

- ▶ When we are plotting an sfc **geometry column**, the plot only displays the geometric shape
- ▶ We can use **basic graphical parameters** to control the appearance, such as -
 - ▶ col - Fill color
 - ▶ border - Outline color
 - ▶ pch - Point shape
 - ▶ cex - Point size
- ▶ For example, to draw **states outline in grey** -

```
plot(st_geometry(states), border = "grey")
```

Basic plotting

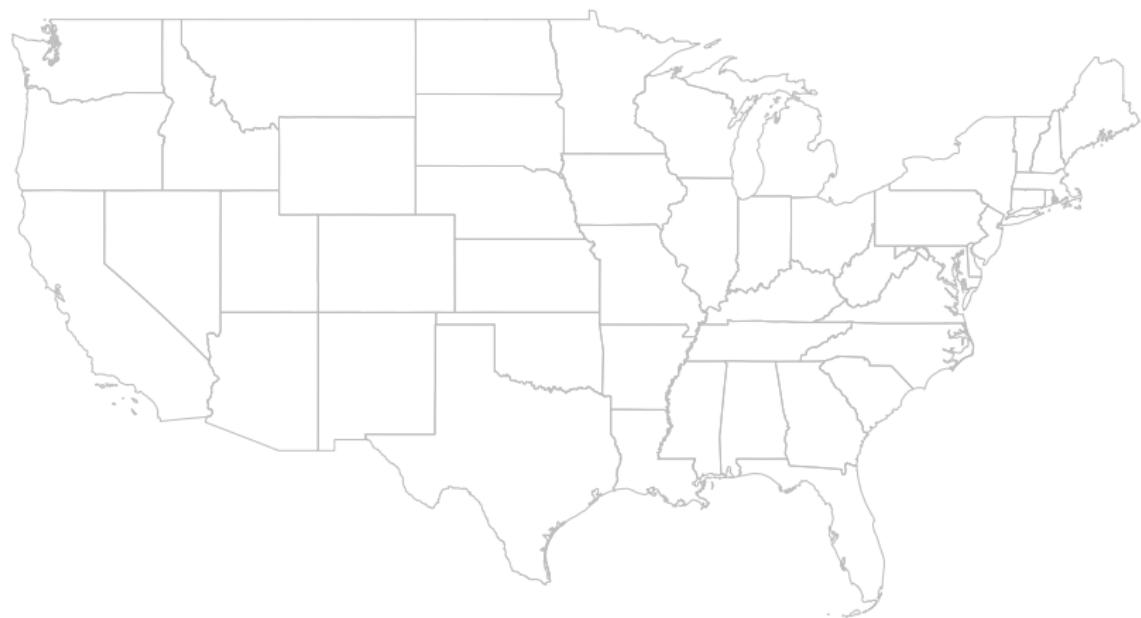


Figure 8: Basic plot of `sfc` object

Basic plotting

- ▶ Additional vector layers can be **drawn** in an **existing graphical window** with add=TRUE
- ▶ For example, the following expressions draw **both** tracks and states
- ▶ Note that the second expression uses add=TRUE

```
plot(st_geometry(states), border = "grey")
plot(st_geometry(tracks), col = "red", add = TRUE)
```

Basic plotting



Figure 9: Using add=TRUE in plot

Interactive map with mapview

```
mapview(tracks, zcol = "wind_mph", legend = TRUE)
```

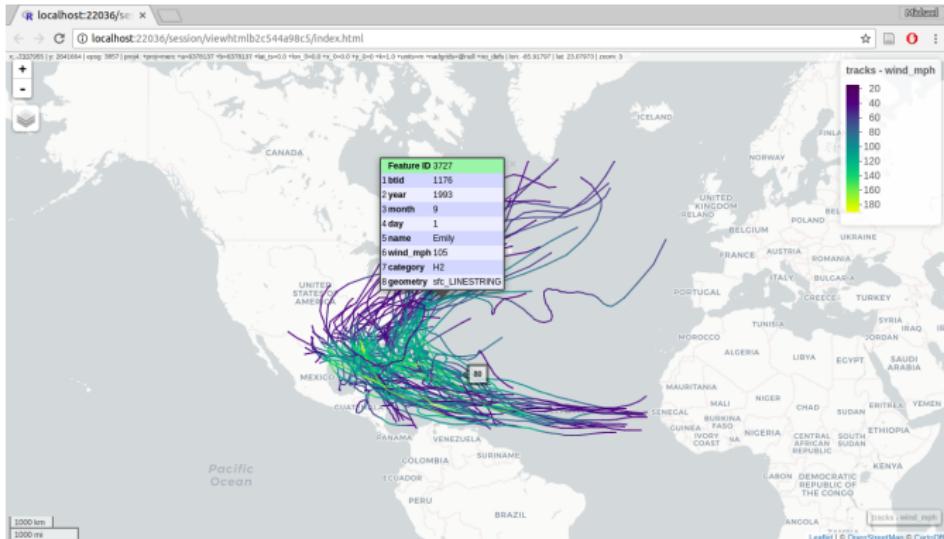


Figure 10: Intractive map of tracks layer

Reprojection

- ▶ **Reprojection** is an important part of spatial analysis workflow since as we often need to -
 - ▶ Transform several layers into the same projection
 - ▶ Switch between un-projected and projected data
- ▶ A vector layer can be reprojected with `st_transform`
- ▶ `st_transform` has **two parameters** -
 - ▶ `x` - The **layer** to be reprojected
 - ▶ `crs` - The **target CRS**
- ▶ The CRS can be **specified** in one of two ways -
 - ▶ A **PROJ.4** definition ("`+proj=longlat +datum=WGS84`")
 - ▶ An **EPSG** code (4326)

Reprojection

- ▶ In the following code section we are reprojecting both the states and tracks layers
- ▶ The **target CRS** is the *US National Atlas Equal Area* projection (EPSG=2163)

```
states = st_transform(states, 2163)
tracks = st_transform(tracks, 2163)
```

Reprojection

state

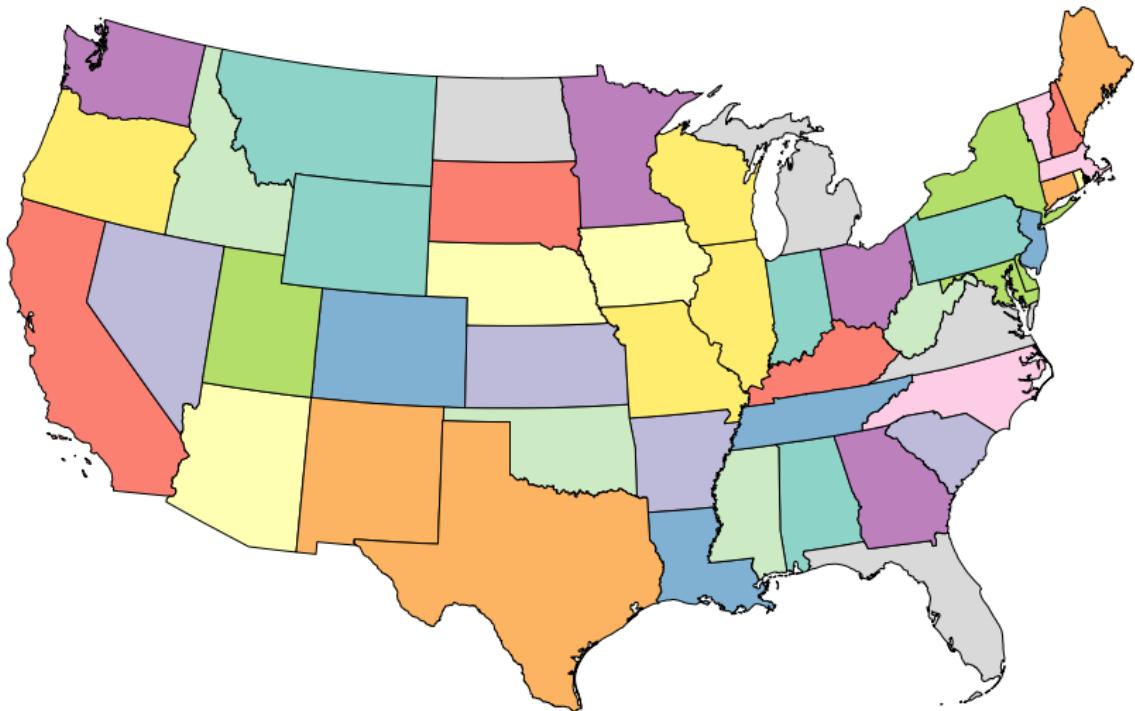


Figure 11: Reprojected states layer

Subsetting

- ▶ **Subsetting** (filtering) of features in an sf vector layer is done in exactly the same way as filtering rows in a data.frame
- ▶ For example, the following expression filters the tracks layer to keep only those storms **which took place after 1949**

```
tracks = tracks[tracks$year > 1949, ]
```

- ▶ As another example, we can create a tracks_h5 layer with all segments that belong to storms which **reached the "H5" category**

```
strong = unique(tracks$btid[tracks$category == "H5"])
tracks_h5 = tracks[tracks$btid %in% strong, ]
```

Geometric calculations

Geometric operations on vector layers can conceptually be divided into **three groups** according to their output -

- ▶ **Numeric** values: Functions that summarize geometrical properties of -
 - ▶ A **single layer** (e.g. area, length)
 - ▶ A **pair of layers** (e.g. distance)
- ▶ **Logical** values: Functions that evaluate whether a certain condition holds true, regarding -
 - ▶ A **single layer** (e.g. geometry is valid)
 - ▶ A **pair of layers** (e.g. feature A intersects feature B)
- ▶ **Spatial** layers: Functions that create a new layer based on -
 - ▶ A **single layer** (e.g. centroids)
 - ▶ A **pair of layers** (e.g. intersection area)

Numeric

- ▶ There are several functions to calculate **numeric geometric properties** of vector layers in package sf -
 - ▶ `st_length`
 - ▶ `st_area`
 - ▶ `st_distance`
 - ▶ `st_bbox`
 - ▶ `st_dimension`

Numeric

- ▶ For example, we can calculate the area of each feature in the states layer (i.e. each state) using `st_area` -

```
states$area = st_area(states)
states$area[1:3]
## Units: m^2
## [1] 134050489381 295336534486 137732515283
```

- ▶ The result is an object of class `units` -

```
class(states$area)
## [1] "units"
```

Numeric

- ▶ We can convert measurements to different units with `set_units` from package `units` -

```
library(units)

states$area = set_units(states$area, km^2)
states$area[1:3]
## Units: km^2
## [1] 134050.5 295336.5 137732.5
```

- ▶ Inspecting result -

```
plot(states[, "area"])
```

Numeric

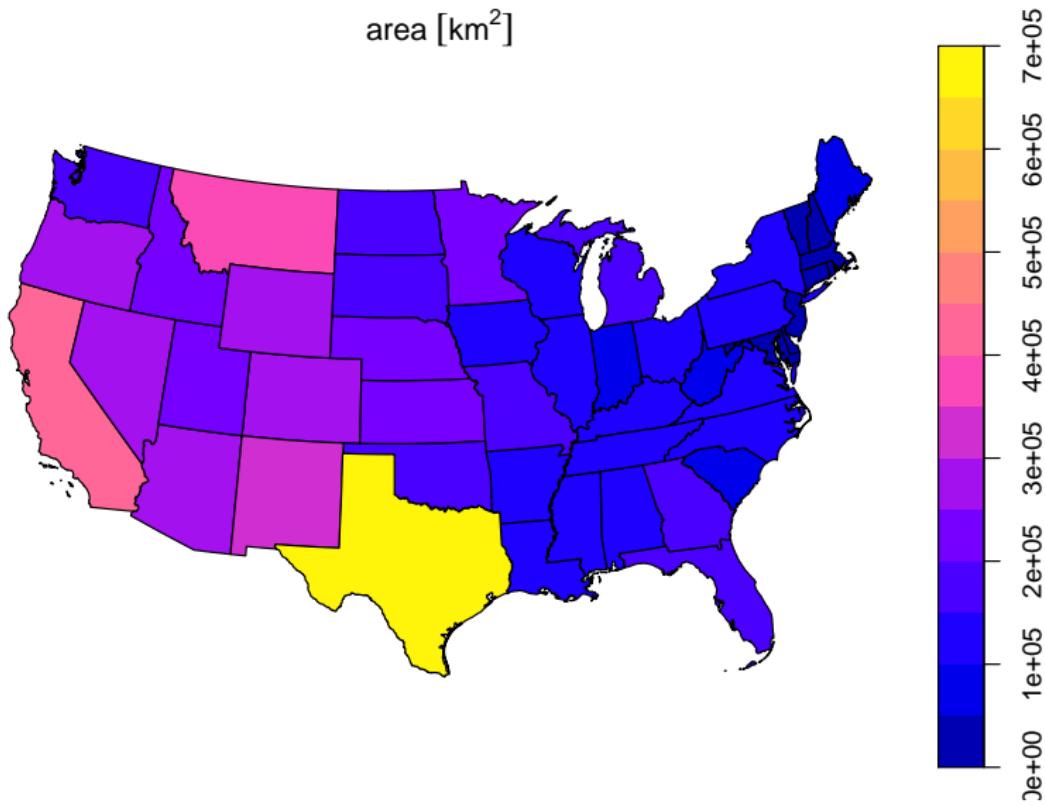


Figure 12: Calculated area attribute

Logical

- ▶ Given two layers, x and y, the following **logical geometric functions** check whether each feature in x maintains the specified **relation** with each feature in y -
 - ▶ st_intersects
 - ▶ st_disjoint
 - ▶ st_touches
 - ▶ st_crosses
 - ▶ st_within
 - ▶ st_contains
 - ▶ st_overlaps
 - ▶ st_covers
 - ▶ st_covered_by
 - ▶ st_equals
 - ▶ st_equals_exact

Logical

- ▶ When specifying `sparse=FALSE` the functions return a **logical matrix**
- ▶ Each **element** i, j in the matrix is TRUE when $f(x[i], y[j])$ is TRUE
- ▶ For example, this creates a matrix of **intersection** relations between states -

```
int = st_intersects(states, states, sparse = FALSE)
```

```
int[1:4, 1:4]
##      [,1]  [,2]  [,3]  [,4]
## [1,] TRUE FALSE FALSE FALSE
## [2,] FALSE TRUE FALSE  TRUE
## [3,] FALSE FALSE  TRUE FALSE
## [4,] FALSE  TRUE FALSE  TRUE
```

Logical



Figure 13: Intersection relations between states features

Spatial

- ▶ sf provides common **geometry-generating** functions applicable to **individual** geometries, such as -
 - ▶ st_centroid
 - ▶ st_buffer
 - ▶ st_sample
 - ▶ st_convex_hull
 - ▶ st_voronoi

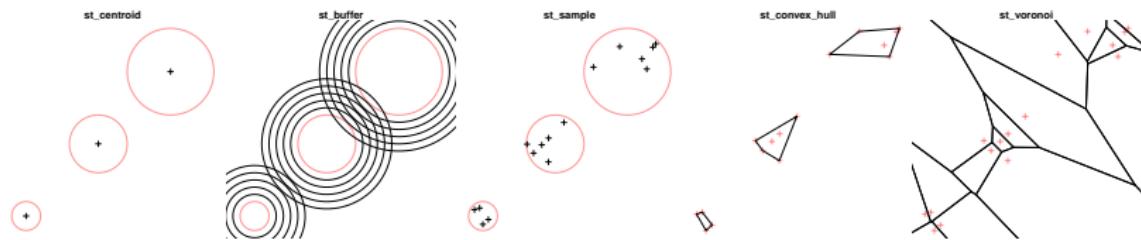


Figure 14: Geometry-generating operations on individual layers

Spatial

- ▶ For example, the following expression uses `st_centroid` to create a layer of **state centroids** -

```
states_ctr = st_centroid(states)
```

- ▶ They can be **plotted** as follows -

```
plot(  
  st_geometry(states),  
  border = "grey"  
)  
  
plot(  
  st_geometry(states_ctr),  
  col = "red", pch = 3,  
  add = TRUE  
)
```

Spatial



Figure 15: State centroids

Spatial

- ▶ Other **geometry-generating** functions work on **pairs** of input geometries -
 - ▶ `st_intersection`
 - ▶ `st_difference`
 - ▶ `st_sym_difference`
 - ▶ `st_union`

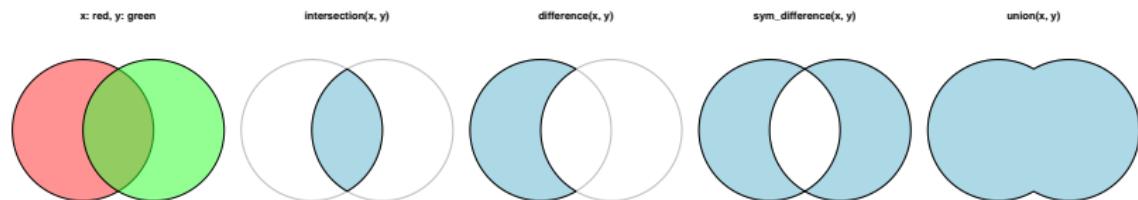


Figure 16: Geometry-generating operations on pairs of layers

Spatial

- ▶ For example, to calculate **total tracks length per state** we can use `st_intersection` to ‘split’ the tracks layer by State

-

```
tracks_int = st_intersection(tracks, states)
```

Spatial

- ▶ The result is a new line layer split by state borders and including a state attribute -

```
plot(  
  st_geometry(states),  
  border = "grey"  
)  
plot(  
  tracks_int[, "state"],  
  lwd = 3,  
  add = TRUE  
)
```

Spatial



Figure 17: Intersection result

Spatial

- ▶ The resulting layer has mixed LINESTRING and MULTILINESTRING geometries (Why?)

```
class(tracks_int$geometry)
## [1] "sfc_GEOMETRY" "sfc"
```

- ▶ To calculate line length we need to convert it to MULTILINESTRING -

```
tracks_int = st_cast(tracks_int, "MULTILINESTRING")
```

- ▶ Verifying the conversion succeeded -

```
class(tracks_int$geometry)
## [1] "sfc_MULTILINESTRING" "sfc"
```

Spatial

- ▶ Let's add a **storm track length** attribute called `length` -

```
tracks_int$length = st_length(tracks_int)
tracks_int$length = set_units(tracks_int$length, km)
```

Join layer with table

- ▶ Next we aggregate attribute table of tracks_int by state, to find the sum of length values -

```
library(dplyr)

track_lengths =
  tracks_int %>%
  st_set_geometry(NULL) %>%
  group_by(state) %>%
  summarize(length = sum(length)) %>%
  as.data.frame
```

Join layer with table

- ▶ The result is a `data.frame` with **total length** of storm tracks **per state** -

```
head(track_lengths)
##           state      length
## 1      Alabama 2272.1309 km
## 2      Arkansas 1000.0448 km
## 3 Connecticut  290.7856 km
## 4    Delaware   58.2246 km
## 5     Florida 2932.1240 km
## 6    Georgia 1505.7847 km
```

Join layer with table

- ▶ Next, we can **join** the aggregated table back to the states layer -

```
states = merge(  
  states,  
  track_lengths,  
  by = "state",  
  all.x = TRUE  
)
```

Join layer with table

- ▶ Here is how the states layer looks like after the join -

```
head(states)
```

```
## Simple feature collection with 6 features and 3 fields
## geometry type: MULTIPOLYGON
## dimension: XY
## bbox: xmin: -2031905 ymin: -1562374 xmax: 2295610 ymax: 67481.2
## epsg (SRID): 2163
## proj4string: +proj=laea +lat_0=45 +lon_0=-100 +x_0=0 +y_0=0 +a=6370997 +b
##           state      area     length          geometry
## 1    Alabama 134050.49 km^2 2272.1309 km MULTIPOLYGON (((1150023 -15...
## 2    Arizona 295336.53 km^2       NA km MULTIPOLYGON (((-1386136 -1...
## 3   Arkansas 137732.52 km^2 1000.0448 km MULTIPOLYGON (((482001 -928...
## 4 California 409701.73 km^2       NA km MULTIPOLYGON (((-1717278 -1...
## 5 Colorado 269373.19 km^2       NA km MULTIPOLYGON (((-786661.9 -...
## 6 Connecticut 12912.34 km^2 290.7856 km MULTIPOLYGON (((2156197 -83...
```

Join layer with table

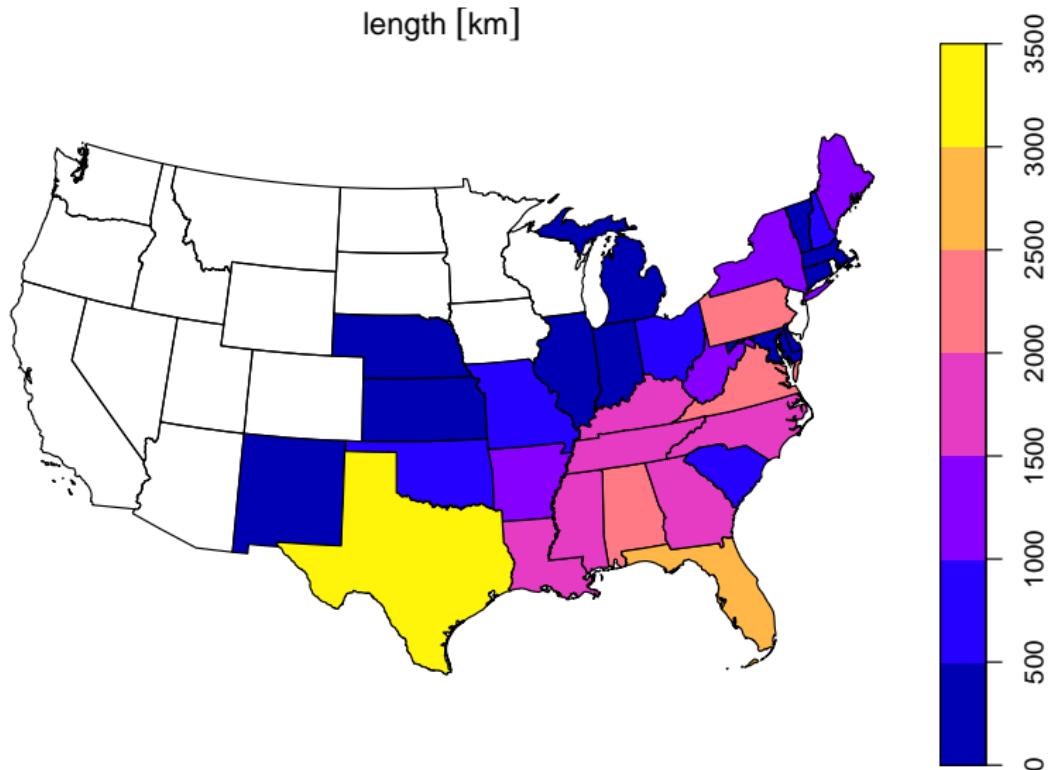


Figure 18: Total track length per state

Join layer with table

- ▶ Finally, we **divide** total track length by state area
- ▶ This gives us track **density** per state -

```
states$track_density = states$length / states$area
```

- ▶ Plotting the layer shows the new `track_density` attribute -

```
plot(states)
```

Join layer with table

track_density [1/km]

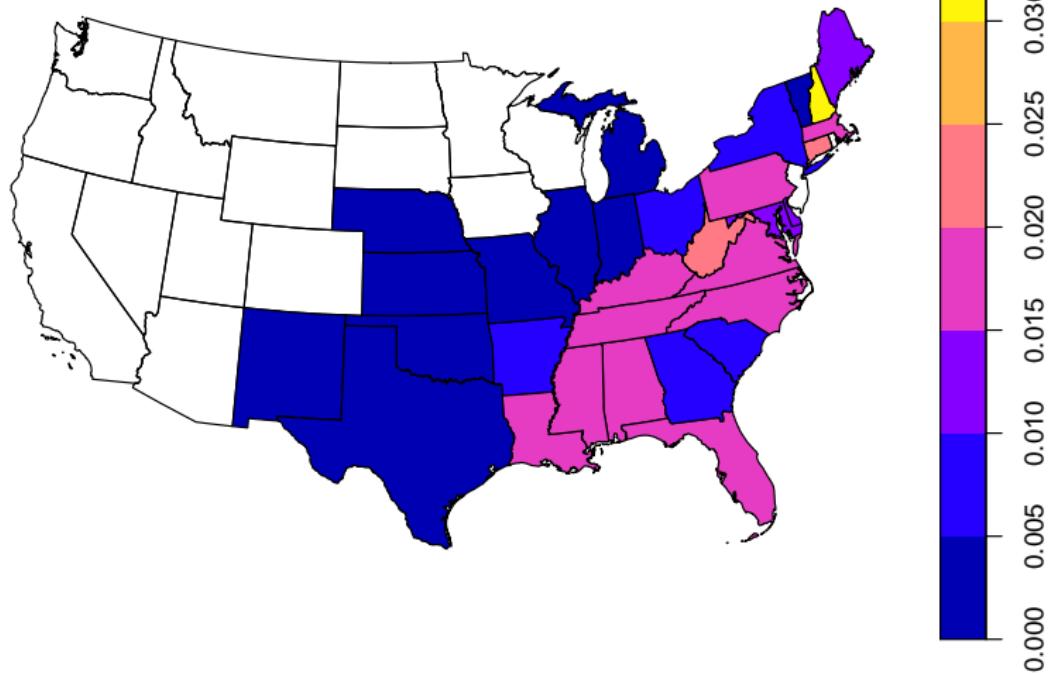


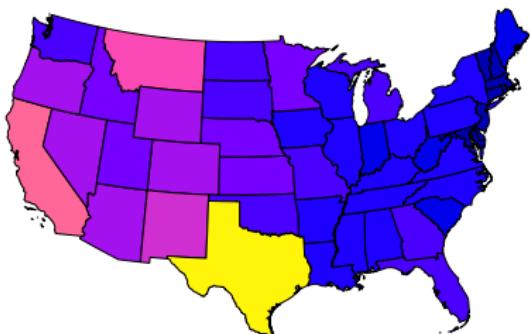
Figure 19: Track density per state

Join layer with table

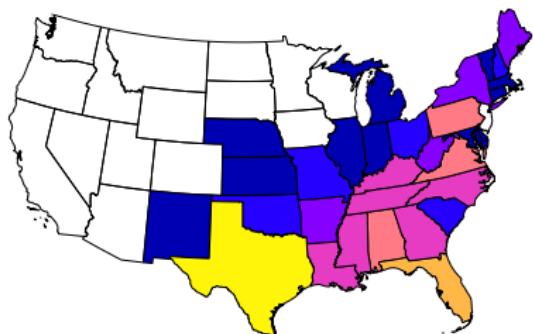
state



area



length



track_density

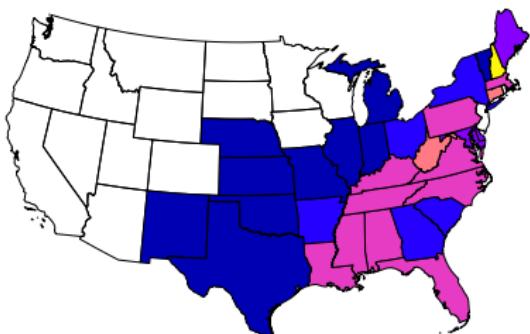


Figure 20: States layer with calculated attributes

Spatial Join

- ▶ One of the most useful operators in spatial analysis is the **spatial join**
- ▶ Function `st_join` accepts -
 - ▶ Two layers x and y
 - ▶ Geometric type join (default `st_intersects`)
 - ▶ Join type left (Left TRUE, Inner FALSE)
- ▶ For example, the following expression joins the `state` attribute of the state each track segment intersects -

```
tracks_states = st_join(  
    states[, "state"],  
    tracks[, "name"]  
)
```

- ▶ The new layer has all states geometries, duplicated when necessary, with all respective tracks attributes for each state/track segment intersection

Spatial Join

tracks_states

```
## Simple feature collection with 397 features and 2 fields
## geometry type:  MULTIPOLYGON
## dimension:      XY
## bbox:            xmin: -2031905 ymin: -2116976 xmax: 2516374 ymax: 732352.2
## epsg (SRID):   2163
## proj4string:    +proj=laea +lat_0=45 +lon_0=-100 +x_0=0 +y_0=0 +a=6370997 +b
## First 10 features:
##       state      name           geometry
## 1   Alabama     Easy  MULTIPOLYGON (((1150023 -15...
## 1.1 Alabama    Easy  MULTIPOLYGON (((1150023 -15...
## 1.2 Alabama    Easy  MULTIPOLYGON (((1150023 -15...
## 1.3 Alabama    Easy  MULTIPOLYGON (((1150023 -15...
## 1.4 Alabama    Easy  MULTIPOLYGON (((1150023 -15...
## 1.5 Alabama    Hilda MULTIPOLYGON (((1150023 -15...
## 1.6 Alabama    Hilda MULTIPOLYGON (((1150023 -15...
## 1.7 Alabama    Eloise MULTIPOLYGON (((1150023 -15...
## 1.8 Alabama    Eloise MULTIPOLYGON (((1150023 -15...
## 1.9 Alabama Frederic MULTIPOLYGON (((1150023 -15...
```

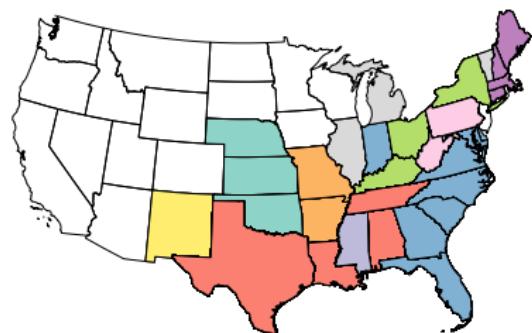
Spatial Join

- ▶ When using `left=FALSE` only geometries that have a match are returned

```
left = st_join(  
  states[, "state"],  
  tracks[, "name"]  
)  
inner = st_join(  
  states[, "state"],  
  tracks[, "name"],  
  left = FALSE  
)
```

Spatial Join

Left join (default)



Inner join

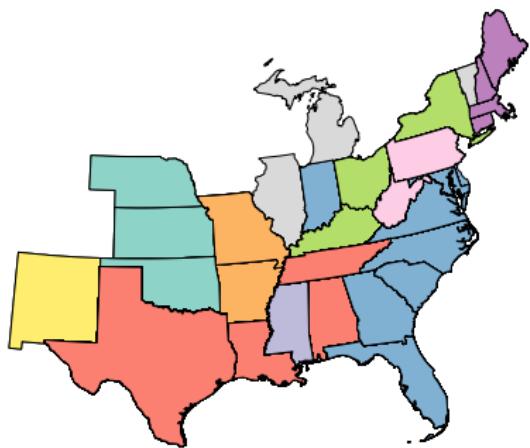


Figure 21: Left vs. Inner spatial join result

Spatial Join

- ▶ In the next example we visualize the states affected by each "H5" storm track
- ▶ First, we union the segments per storm

```
tracks_h5_agg = aggregate(  
  tracks_h5[, "name"],  
  data.frame(id = tracks_h5$btid),  
  unique  
)
```

Spatial Join

```
tracks_h5_agg
```

```
## Simple feature collection with 8 features and 2 fields
## Attribute-geometry relationship: 0 constant, 1 aggregate, 1 identity
## geometry type: MULTILINESTRING
## dimension: XY
## bbox: xmin: -285061.9 ymin: -3250141 xmax: 7729104 ymax: 2534017
## epsg (SRID): 2163
## proj4string: +proj=laea +lat_0=45 +lon_0=-100 +x_0=0 +y_0=0 +a=6370997 +b
##           id      name          geometry
## 1   864    Donna MULTILINESTRING ((2389447 -...
## 2   869    Carla MULTILINESTRING ((2570017 -...
## 3   922   Beulah MULTILINESTRING ((2349914 -...
## 4   939  Camille MULTILINESTRING ((1920845 -...
## 5  1050    Allen MULTILINESTRING ((2807307 -...
## 6  1139     Hugo MULTILINESTRING ((3876948 -...
## 7  1166   Andrew MULTILINESTRING ((1703214 -...
## 8  1319     Ivan MULTILINESTRING ((3484779 -...
```

Spatial Join

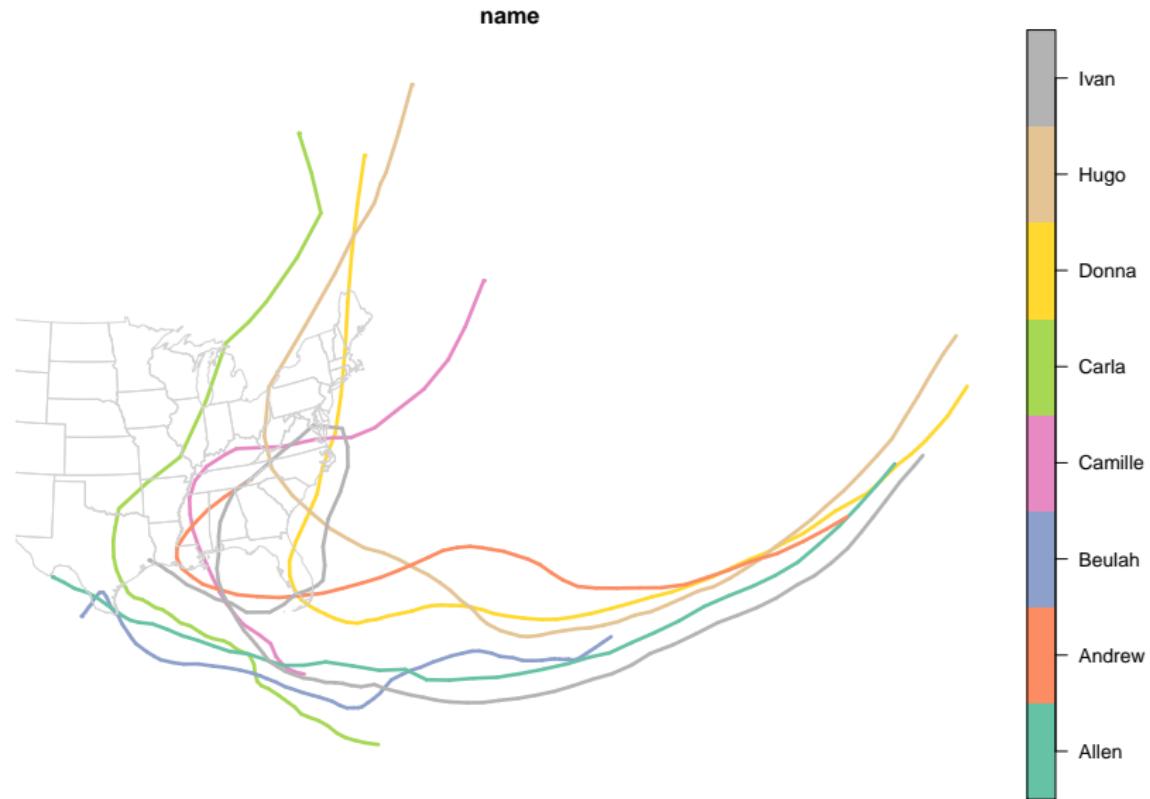


Figure 22: "H5" storm tracks

Spatial Join

- ▶ Second, we join the aggregated tracks to the states layer
- ▶ As a result, the states features are duplicated per tracks_h5_agg feature they intersect

```
states_tracks_h5 = st_join(  
    states,  
    tracks_h5_agg,  
    left = FALSE  
)
```

Spatial Join

```
head(states_tracks_h5)
```

```
## Simple feature collection with 6 features and 6 fields
## geometry type:  MULTIPOLYGON
## dimension:      XY
## bbox:           xmin: 482001 ymin: -2086803 xmax: 2295610 ymax: 67481.2
## epsg (SRID):   2163
## proj4string:   +proj=laea +lat_0=45 +lon_0=-100 +x_0=0 +y_0=0 +a=6370997 +b
##                 state          area      length track_density id name
## 1      Alabama 134050.49 km^2 2272.1309 km 0.016949814 1/km 1166 Andrew
## 1.1    Alabama 134050.49 km^2 2272.1309 km 0.016949814 1/km 1319 Ivan
## 3      Arkansas 137732.52 km^2 1000.0448 km 0.007260775 1/km  869 Carla
## 6     Connecticut 12912.34 km^2  290.7856 km 0.022519971 1/km  864 Donna
## 9      Florida 151838.27 km^2 2932.1240 km 0.019310836 1/km  864 Donna
## 9.1    Florida 151838.27 km^2 2932.1240 km 0.019310836 1/km 1166 Andrew
##                 geometry
## 1  MULTIPOLYGON (((1150023 -15...
## 1.1 MULTIPOLYGON (((1150023 -15...
## 3  MULTIPOLYGON (((482001 -928...
## 6  MULTIPOLYGON (((2156197 -83...
## 9  MULTIPOLYGON (((1953691 -20...
## 9.1 MULTIPOLYGON (((1953691 -20...
```

Mapping with ggplot2

- ▶ The development version of ggplot2 supports plotting sf layers using function `geom_sf`

```
devtools::install_github("tidyverse/ggplot2")
```

- ▶ For example -

```
library(ggplot2)

ggplot() +
  geom_sf(
    data = states
  ) +
  geom_sf(
    data = tracks_h5_agg,
    aes(colour = name)
  )
```

Mapping with ggplot2

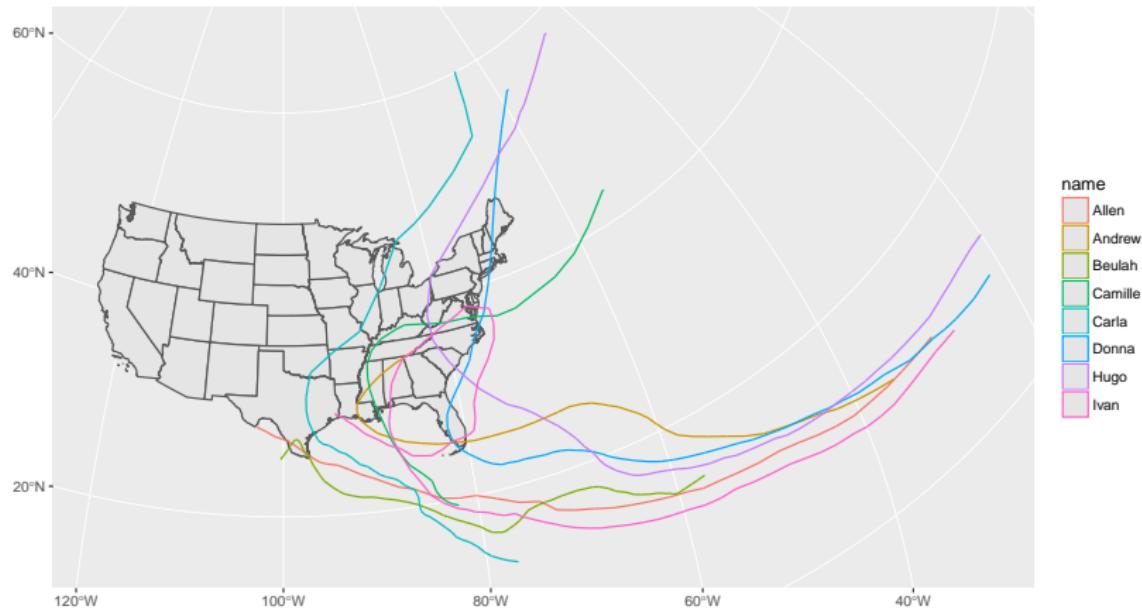


Figure 23: Mapping sf layer with ggplot2

Mapping with ggplot2

```
ggplot() +
  geom_sf(
    data = states,
    fill = NA, colour = "black", size = 0.1
  ) +
  geom_sf(
    data = states_tracks_h5,
    fill = "red", colour = NA, size = 0.3, alpha = 0.2
  ) +
  facet_wrap(~ name) +
  theme_bw()
```

Mapping with ggplot2

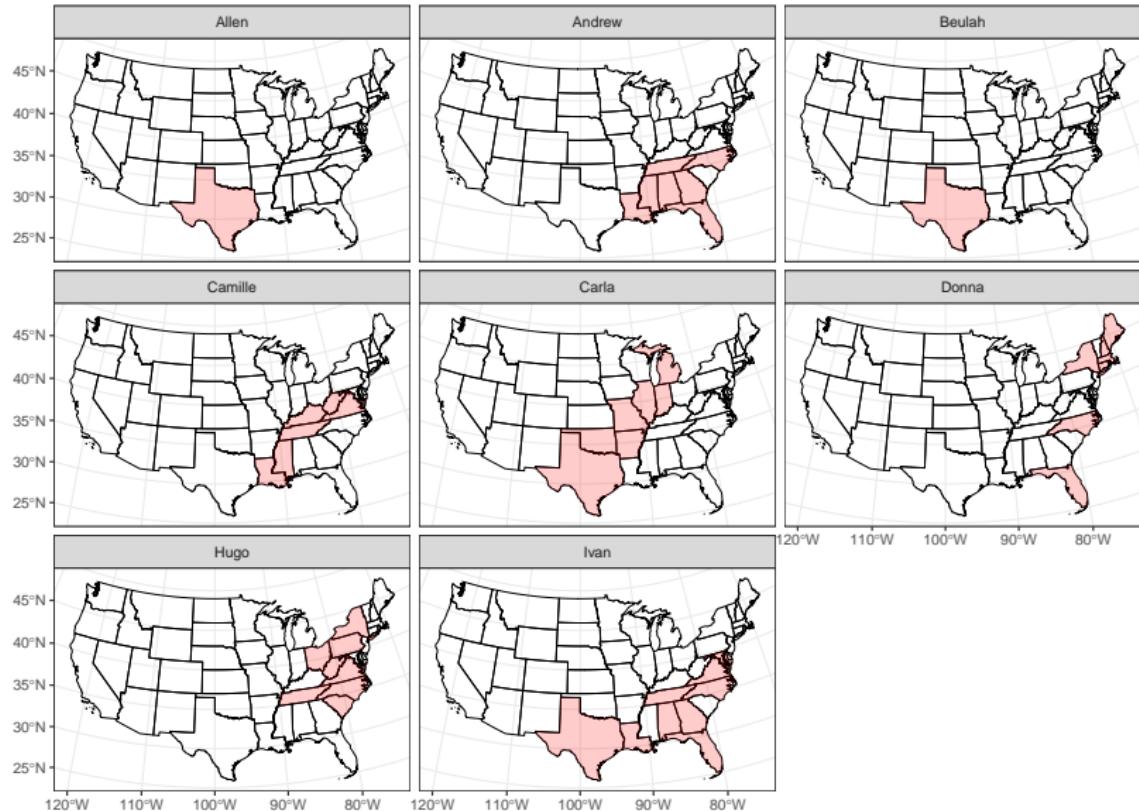


Figure 24: Using ggplot2 facets

Mapping with ggplot2

```
ggplot() +
  geom_sf(
    data = states,
    fill = NA, colour = "darkgrey", size = 0.1
  ) +
  geom_sf(
    data = states_tracks_h5,
    fill = "red", colour = "black", size = 0.3, alpha = 0.2
  ) +
  geom_sf(
    data = tracks_h5,
    aes(size = wind_mph, colour = wind_mph)
  ) +
  geom_sf(
    data = tracks_h5,
    colour = "black", size = 0.2
  ) +
```

Mapping with ggplot2

```
coord_sf(  
  xlim = st_bbox(states)[c(1,3)],  
  ylim = st_bbox(states)[c(2,4)]  
) +  
scale_size_continuous(  
  range = c(0.1, 5), guide = FALSE  
) +  
scale_colour_distiller(  
  "Wind speed\n(miles / hour)", palette = "Spectral"  
) +  
facet_wrap(~ name, ncol = 3) +  
theme_bw() +  
theme(  
  panel.grid.major = element_line(colour = "transparent"),  
  axis.text = element_blank(),  
  axis.ticks = element_blank(),  
  legend.position = c(0.8, 0.17)  
)
```

Mapping with ggplot2

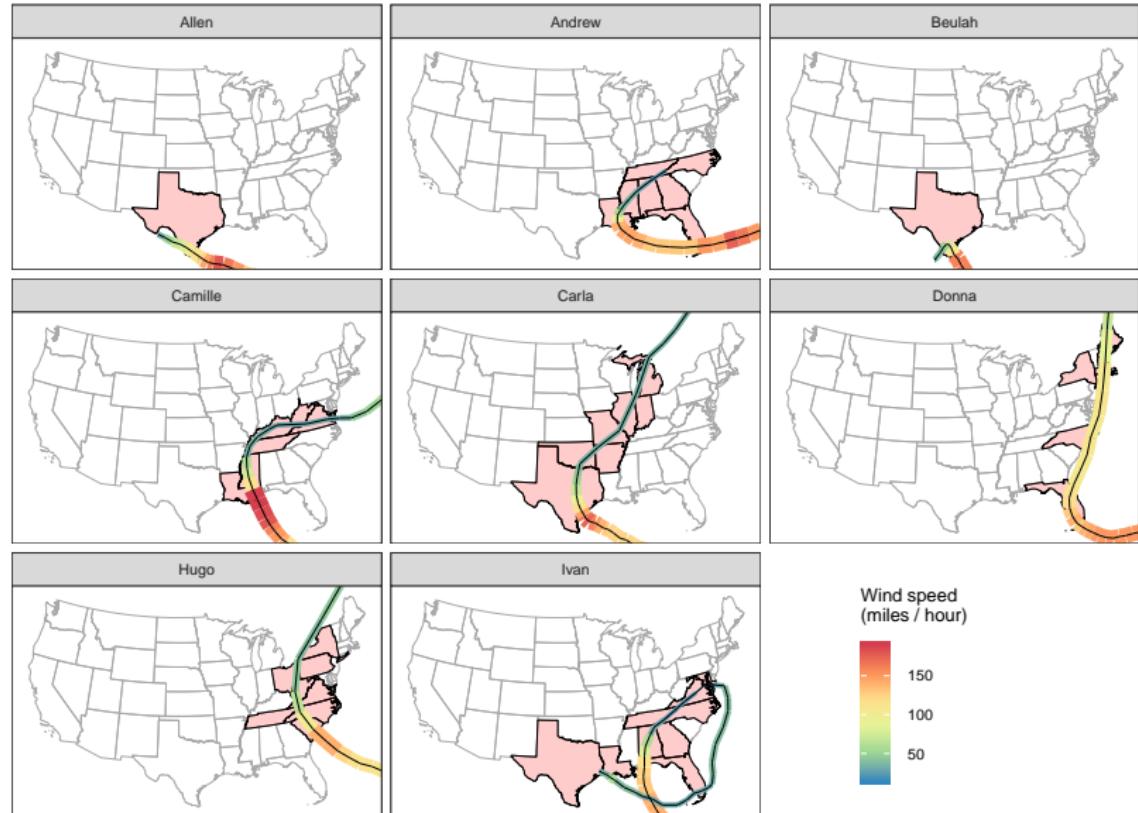


Figure 25: Level 5 hurricanes, wind speed and affected states

More information

- ▶ Official `sf` [tutorials](#)
- ▶ `sf` tutorial from [useR!2017](#) conference
- ▶ `sf` tutorial from [rstudio::conf 2018](#) conference
- ▶ The r-spatial [blog](#)
- ▶ Geocomputation with R (2018) [book](#)

Thank you for listening!