Assignment 9: Spatial Analysis in R

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OVERVIEW

This exercise accompanies the lessons in Environmental Data Analytics (ENV872L) on spatial analysis.

Directions

- 1. Change "Student Name" on line 3 (above) with your name.
- 2. Use the lesson as a guide. It contains code that can be modified to complete the assignment.
- 3. Work through the steps, creating code and output that fulfill each instruction.
- 4. Be sure to **answer the questions** in this assignment document. Space for your answers is provided in this document and is indicated by the ">" character. If you need a second paragraph be sure to start the first line with ">". You should notice that the answer is highlighted in green by RStudio.
- 5. When you have completed the assignment, **Knit** the text and code into a single HTML file.
- 6. After Knitting, please submit the completed exercise (PDF file) in Sakai. Please add your last name into the file name (e.g., "Fay_A10_SpatialAnalysis.pdf") prior to submission.

DATA WRANGLING

Set up your session

- 1. Check your working directory
- 2. Import libraries: tidyverse, sf, leaflet, and mapview

```
library(knitr)
opts_chunk$set(tidy.opts=list(width.cutoff=50),tidy=TRUE, echo = T)
knitr::opts_chunk$set(tidy.opts=list(width.cutoff=60), tidy=TRUE, echo=TRUE)
#1.
getwd()
```

[1] "/Users/britneypepper/Desktop/ENVIRON 872 and L/GitHubRepositories/Environmental_Data_Analytics_

```
#2.
#install.packages("lubridate")
#install.packages("tidyverse")
#install.packages("sf")
#install.packages("leaflet")
#install.packages("mapview")
#install.packages("ggplot2")
#install.packages("RColorBrewer")
library(lubridate)
```

```
##
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
##
      date, intersect, setdiff, union
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.1 --
## v ggplot2 3.3.5
                               0.3.4
                     v purrr
## v tibble 3.1.6
                               1.0.8
                     v dplyr
          1.1.4
## v tidyr
                     v stringr 1.4.0
## v readr
          2.1.1
                     v forcats 0.5.1
## -- Conflicts ----- tidyverse_conflicts() --
## x lubridate::as.difftime() masks base::as.difftime()
## x lubridate::date() masks base::date()
## x dplyr::filter()
                           masks stats::filter()
## x lubridate::intersect() masks base::intersect()
## x dplyr::lag()
                            masks stats::lag()
## x dplyf::lag() masks stats::lag()
## x lubridate::setdiff() masks base::setdiff()
## x lubridate::union()
                            masks base::union()
library(sf)
## Linking to GEOS 3.9.1, GDAL 3.4.0, PROJ 8.1.1; sf_use_s2() is TRUE
library(leaflet)
library(mapview)
library(ggplot2)
library(RColorBrewer)
```

Read (and filter) county features into an sf dataframe and plot

In this exercise, we will be exploring stream gage height data in Nebraska corresponding to floods occurring there in 2019. First, we will import from the US Counties shapefile we've used in lab lessons, filtering it this time for just Nebraska counties. Nebraska's state FIPS code is 31 (as North Carolina's was 37).

- 3. Read the cb_2018_us_county_20m.shp shapefile into an sf dataframe, filtering records for Nebraska counties (State FIPS = 31)
- 4. Reveal the dataset's coordinate reference system
- 5. Plot the records as a map (using mapview or ggplot)

```
# 3. Read in Counties shapefile into an sf dataframe,
# filtering for just NE counties
Nebraska_counties <- st_read("/Users/britneypepper/Desktop/ENVIRON 872 and L/GitHubRepositories/Environ
filter(STATEFP == 31)</pre>
```

```
## Reading layer 'cb_2018_us_county_20m' from data source
## '/Users/britneypepper/Desktop/ENVIRON 872 and L/GitHubRepositories/Environmental_Data_Analytics_20
## using driver 'ESRI Shapefile'
## Simple feature collection with 3220 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension: XY
## Bounding box: xmin: -179.1743 ymin: 17.91377 xmax: 179.7739 ymax: 71.35256
## Geodetic CRS: NAD83

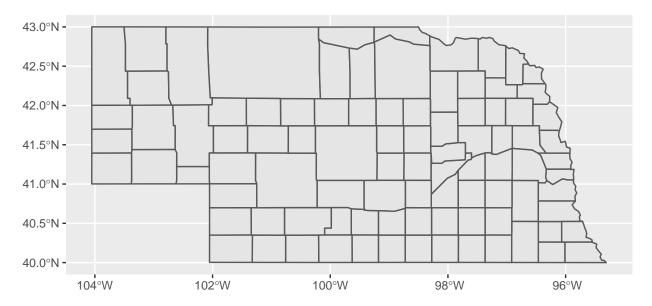
## 4. Reveal the CRS of the counties features
Nebraska_counties_CRS <- st_crs(Nebraska_counties)$epsg</pre>
```

[1] 4269

print(Nebraska_counties_CRS)

```
# this returns the EPSG code 4269 for the associated CRS

# 5. Plot the data
Nebraska_counties_plot <- ggplot(data = Nebraska_counties) +
    geom_sf()
plot(Nebraska_counties_plot)</pre>
```



6. What is the EPSG code of the Counties dataset? Is this a geographic or a projected coordinate reference system? (Or, does this CRS use angular or planar coordinate units?) To what datum is this CRS associated? (Tip: look the EPSG code on https://spatialreference.org)

ANSWER: The EPSG code of the Counties dataset is 4269. This is a geographic coordinate reference system because the coordinates are measured in degrees of latitude and longitude. The datum that is associated with the CRS is NAD83.

Read in gage locations csv as a dataframe, then display the column names it contains

Next we'll read in some USGS/NWIS gage location data added to the Data/Raw folder. These are in the NWIS_SiteInfo_NE_RAW.csv file.(See NWIS_SiteInfo_NE_RAW.README.txt for more info on this dataset.)

- 7. Read the NWIS SiteInfo NE RAW.csv file into a standard dataframe.
- 8. Display the column names of this dataset.

```
# 7. Read in gage locations csv as a dataframe
gage_locations <- read.csv("/Users/britneypepper/Desktop/ENVIRON 872 and L/GitHubRepositories/Environment
stringsAsFactors = TRUE)

# 8. Reveal the names of the columns
colnames(gage_locations)</pre>
```

9. What columns in the dataset contain the x and y coordinate values, respectively?

> ANSWER: > The column "dec_lat_va" contains the y coordinate values and the "dec_long_va" column containes the x coordinate values.

Convert the dataframe to a spatial features ("sf") dataframe

10. Convert the dataframe to an sf dataframe.

[4] "coord_acy_cd"

- Note: These data use the same coordinate reference system as the counties dataset
- 11. Display the column names of the resulting sf dataframe

```
# 10. Convert to an sf object
gage_locations_sf <- read_csv("/Users/britneypepper/Desktop/ENVIRON 872 and L/GitHubRepositories/Environ</pre>
   group_by(site_no, station_nm, site_tp_cd, dec_lat_va, dec_long_va,
       coord_acy_cd, dec_coord_datum_cd) %>%
   st_as_sf(coords = c("dec_long_va", "dec_lat_va"), crs = 4269)
## Rows: 175 Columns: 7
## -- Column specification -----
## Delimiter: ","
## chr (5): site_no, station_nm, site_tp_cd, coord_acy_cd, dec_coord_datum_cd
## dbl (2): dec_lat_va, dec_long_va
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
# Sorry that you can't see the end of the read_csv line.
# There is a %>% at the end of it so that the pipe works.
# 11. Re-examine the column names
colnames(gage_locations_sf)
## [1] "site_no"
                           "station_nm"
                                                "site_tp_cd"
```

"dec_coord_datum_cd" "geometry"

12. What new field(s) appear in the sf dataframe created? What field(s), if any, disappeared?

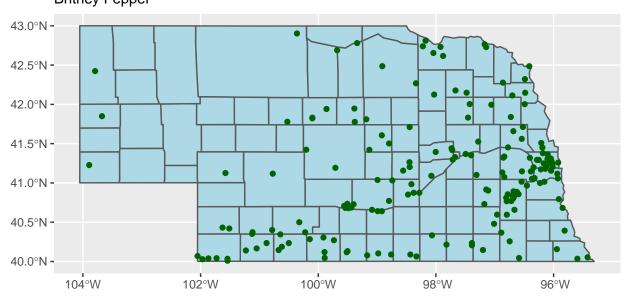
ANSWER: There is a new field called "geometry" appeared in the sf dataframe we created and the "dec_lat_va" and "dec_long_va" fields disappeared.

Plot the gage locations on top of the counties

- 13. Use ggplot to plot the county and gage location datasets.
 - Be sure the datasets are displayed in different colors
 - Title your plot "NWIS Gage Locations in Nebraska"
 - Subtitle your plot with your name

```
# 13. Plot the gage locations atop the county features
gage_counties_ggplot1 <- ggplot() + geom_sf(data = Nebraska_counties,
    fill = "light blue") + geom_sf(data = gage_locations_sf,
    color = "dark green")
print(gage_counties_ggplot1 + ggtitle("NWIS Gage Locations in Nebraska") +
    labs(subtitle = "Britney Pepper"))</pre>
```

NWIS Gage Locations in Nebraska Britney Pepper



Read in the gage height data and join the site location data to it.

Lastly, we want to attach some gage height data to our site locations. I've constructed a csv file listing many of the Nebraska gage sites, by station name and site number along with stream gage heights (in meters) recorded during the recent flood event. This file is titled NWIS_SiteFlowData_NE_RAW.csv and is found in the Data/Raw folder.

- 14. Read the NWIS_SiteFlowData_NE_RAW.csv dataset in as a dataframe.
- 15. Show the column names .
- 16. Join our site information (already imported above) to these gage height data.

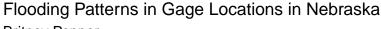
- The site_no and station_nm can both/either serve as joining attributes.
- Construct this join so that the result only includes spatial features where both tables have data.
- 17. Show the column names in this resulting spatial features object
- 18. Show the dimensions of the resulting joined dataframe

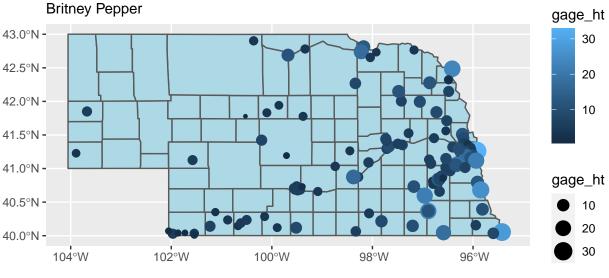
```
# 14. Read the site flow data into a data frame
nwis_gage_sites <- read.csv("/Users/BritneyPepper/Desktop/ENVIRON 872 and L/GitHubRepositories/Environm
    colClasses = c(site_no = "factor"), stringsAsFactors = TRUE)
# 15. Show the column names
colnames(nwis_gage_sites)
## [1] "site_no"
                    "station_nm" "date"
                                               "gage_ht"
# 16. Join location data to it
counties_sf_join <- merge(gage_locations_sf, nwis_gage_sites,</pre>
    by = c("site_no", "station_nm"))
# 17. Show the column names of the joined dataset
colnames(counties_sf_join)
## [1] "site no"
                             "station nm"
                                                  "site_tp_cd"
## [4] "coord_acy_cd"
                             "dec_coord_datum_cd" "date"
## [7] "gage_ht"
                             "geometry"
# 18. Show the dimensions of this joined dataset
dim(counties sf join)
## [1] 136
nrow(counties_sf_join)
## [1] 136
ncol(counties_sf_join)
## [1] 8
```

Map the pattern of gage height data

Now we can examine where the flooding appears most acute by visualizing gage heights spatially. 19. Plot the gage sites on top of counties (using mapview, ggplot, or leaflet) * Show the magnitude of gage height by color, shape, other visualization technique.

```
# Map the points, sized by gage height
gage_counties_ggplot2 <- ggplot() + geom_sf(data = Nebraska_counties,
    fill = "light blue", size = 0.5) + geom_sf(data = counties_sf_join,
    aes(size = gage_ht, color = gage_ht))
print(gage_counties_ggplot2 + ggtitle("Flooding Patterns in Gage Locations in Nebraska") +
    labs(subtitle = "Britney Pepper"))</pre>
```





SPATIAL ANALYSIS

Up next we will do some spatial analysis with our data. To prepare for this, we should transform our data into a projected coordinate system. We'll choose UTM Zone 14N (EPGS = 32614).

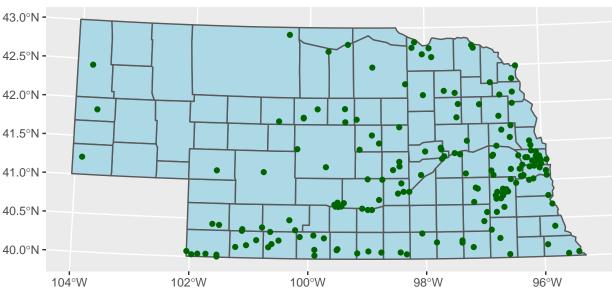
Transform the counties and gage site datasets to UTM Zone 14N

- 20. Transform the counties and gage sf datasets to UTM Zone 14N (EPGS = 32614).
- 21. Using mapview or ggplot, plot the data so that each can be seen as different colors

```
# 20 Transform the counties and gage location datasets to
# UTM Zone 14
Nebraska_counties_utm <- st_transform(Nebraska_counties, crs = 32614)
gage_locations_sf_utm <- st_transform(gage_locations_sf, crs = 32614)

# 21 Plot the data
gage_counties_ggplot2_utm <- ggplot() + geom_sf(data = Nebraska_counties_utm,
    fill = "light blue") + geom_sf(data = gage_locations_sf_utm,
    color = "dark green")
print(gage_counties_ggplot2_utm + ggtitle("Gage Locations in Nebraska (UTM Zone 14N)") +
    labs(subtitle = "Britney Pepper"))</pre>
```

Gage Locations in Nebraska (UTM Zone 14N) Britney Pepper



Select the gages falling within a given county

Now let's zoom into a particular county and examine the gages located there. 22. Select Lancaster county from your county sf dataframe 23. Select the gage sites falling within that county * Use either matrix subsetting or tidy filtering 24. Create a plot showing: * all Nebraska counties, * the selected county, * and the gage sites in that county

Gage Locations in Lancaster County, Nebraska Britney Pepper

