Assignment 9: Spatial Analysis in R

Yue Zhang

OVERVIEW

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

Directions

- 1. Rename this file <FirstLast>_A09_SpatialAnalysis.Rmd (replacing <FirstLast> with your first and last name).
- 2. Change "Student Name" on line 3 (above) with your name.
- 3. Use the lesson as a guide. It contains code that can be modified to complete the assignment.
- 4. Work through the steps, creating code and output that fulfill each instruction.
- 5. 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.
- 6. When you have completed the assignment, **Knit** the text and code into a single HTML file.

DATA WRANGLING

Set up your session

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

```
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.2 --
## v ggplot2 3.4.0
                   v purrr
                            1.0.1
## v tibble 3.1.8
                   v dplyr
                            1.1.0
## v tidyr
           1.3.0
                   v stringr 1.5.0
## v readr
           2.1.3
                   v forcats 1.0.0
## -- Conflicts ----- tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                  masks stats::lag()
library(sf)
## Warning: package 'sf' was built under R version 4.2.3
```

Linking to GEOS 3.9.3, GDAL 3.5.2, PROJ 8.2.1; sf_use_s2() is TRUE

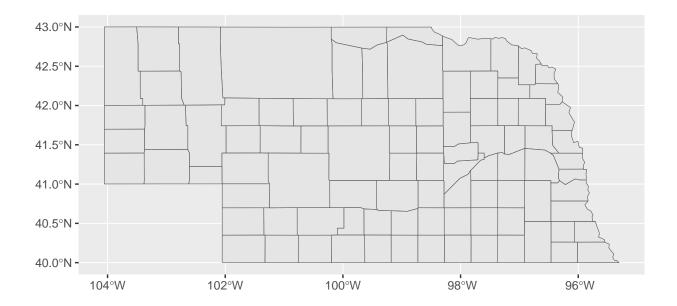
```
library(leaflet)
## Warning: package 'leaflet' was built under R version 4.2.3
library(mapview)
## Warning: package 'mapview' was built under R version 4.2.3
#2.
getwd()
## [1] "E:/MCRP/Spring2023/EDA_Spring2023"
Read (and filter) county features into an sf dataframe and plot
In this exercise, we 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_sf = st_read("./Data/Spatial/cb_2018_us_county_20m.shp") %%
    filter(STATEFP == "31")
## Reading layer 'cb_2018_us_county_20m' from data source
     \verb|`E:\MCRP\Spring2023\EDA_Spring2023\Data\Spatial\cb_2018_us\_county_20m.shp'| \\
##
     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
st_crs(Nebraska_sf)
## Coordinate Reference System:
##
     User input: NAD83
     wkt:
##
## GEOGCRS["NAD83",
       DATUM["North American Datum 1983",
##
##
           ELLIPSOID["GRS 1980",6378137,298.257222101,
##
               LENGTHUNIT["metre",1]]],
```

PRIMEM["Greenwich",0,

##

```
ANGLEUNIT["degree", 0.0174532925199433]],
##
##
       CS[ellipsoidal,2],
           AXIS["latitude", north,
##
                ORDER[1],
##
                ANGLEUNIT["degree", 0.0174532925199433]],
##
           AXIS["longitude", east,
##
##
               ORDER[2],
                ANGLEUNIT["degree", 0.0174532925199433]],
##
##
       ID["EPSG",4269]]
```

```
# 5. Plot the data
ggplot() + geom_sf(data = Nebraska_sf)
```



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: EPSG: 4269, this is a geographic coordinate reference system. It is associated with WGS84 datum.

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.

9. What columns in the dataset contain the x and y coordinate values, respectively? > ANSWER: x:dec_long_va; y:dec_lat_va >

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

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

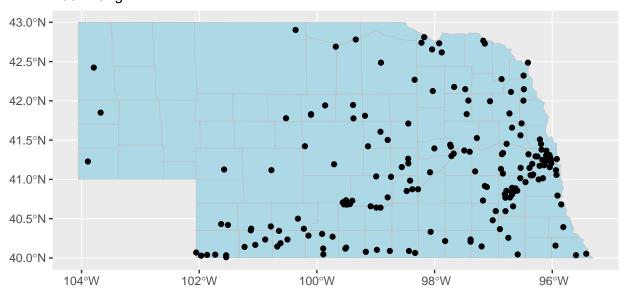
12. What new field(s) appear in the sf dataframe created? What field(s), if any, disappeared?

ANSWER: The fields: "dec_long_va" and "dec_lat_va" disappeared, and the field "geometry" generated.

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

NWIS Gage Locations in Nebraska Yue Zhang



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 in and wrangle the NWIS_SiteFlowData_NE_RAW.csv dataset:
 - Be sure the site_no column is read in as a factor
- Group by site_no and station_nm, computing the mean gage_ht across groups values
- 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

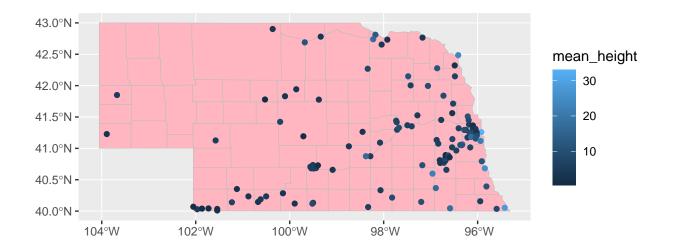
```
# 14. Read the site flow data into a data frame
NWIS height = read.csv("./Data/Raw/NWIS SiteFlowData NE RAW.csv",
    colClasses = c(site_no = "factor")) %>%
    group_by(site_no, station_nm) %>%
    summarize(mean_height = mean(gage_ht))
## 'summarise()' has grouped output by 'site_no'. You can override using the
## '.groups' argument.
# 15. Show the column names
colnames(NWIS_height)
## [1] "site_no"
                     "station_nm" "mean_height"
# 16. Join the flow data to our NWIS gage
# location spatial dataframe
NWIS_join = merge(x = NWIS_sf, y = NWIS_height, by = c("site_no",
    "station_nm"))
# 17. Show the column names of the joined dataset
colnames(NWIS_join)
## [1] "site_no"
                            "station nm"
                                                 "site_tp_cd"
## [4] "coord_acy_cd"
                            "dec_coord_datum_cd" "mean_height"
## [7] "geometry"
# 18. Show the dimensions of this joined dataset
dim(NWIS_join)
```

Map the pattern of gage height data

[1] 118

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
NWIS_join_sf = st_as_sf(NWIS_join, coords = c("dec_long_va",
        "dec_lat_va"), crs = 4269)
ggplot() + geom_sf(data = Nebraska_sf, fill = "lightpink",
        color = "grey") + geom_sf(data = NWIS_join_sf,
        aes(color = mean_height))
```

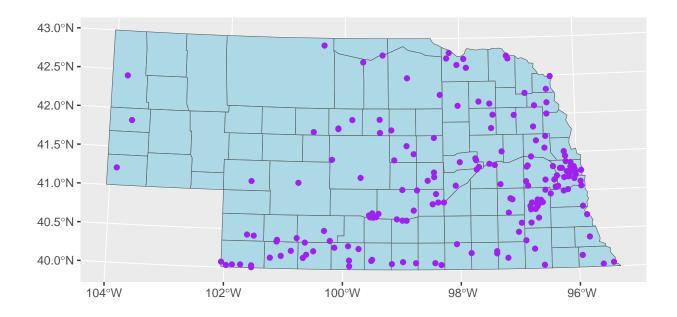


SPATIAL ANALYSIS

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



Select the gages falling within a given county

Now let's zoom into a particular county and examine the gages located there. 22. Select Douglas county from your projected 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

```
# 22 Select the county
Douglas_sf = filter(Nebraska_utm_sf, NAME == "Douglas")

# 23 Select gages within the selected county
gage_intersect = NWIS_UTM_sf %>%
    st_filter(Douglas_sf, .pred = st_intersects)

# 24 Plot
ggplot() + geom_sf(data = Nebraska_utm_sf, fill = "lightblue") +
    geom_sf(data = Douglas_sf, fill = "blue") + geom_sf(data = gage_intersect,
    color = "orange")
```

