Matching points to polygons

This document demonstrates approaches for determining the (multi)polygon each geographic point falls into. The process relies heavily on the sf package (Pebesma, 2018).

Specifically, we will determine which zip code tabulation area (ZCTA) 960 points spread throughout the state of Colorado fall into.

The materials for this demonstration are available at [https://github.com/jfrench/code demonstration/tree/ main/match_points_to_polygons]

Interested parties will need to:

- Install and load the sf package.
- Download usgs_data_series_520_clean.csv into their current working directory.
- Unzip co_zcta.zip into their current working directory. I assumed the zip file is unzipped into the co_zcta folder.

```
library(sf) # needed for spatial analysis
# download csv file to working directly
download.file("https://raw.githubusercontent.com/jfrench/code_demonstration/main/match_points_to_polygon
              destfile = "usgs_data_series_520_clean.csv")
# download zip file to working directly
download.file("https://github.com/jfrench/code_demonstration/raw/main/match_points_to_polygons/co_zcta..
              "co_zcta.zip")
# unzip co_zcta.zip into working directory
unzip("co_zcta.zip")
```

First, we use sf::st_read to import the shapefile describing the ZCTAs. The sf::st_read function will import the shapefile as an sf object with MULTIPOLYGON geometry type. We assign the name co_zcta to this object.

```
co_zcta <- st_read("./co_zcta/Colorado_ZIP_Code_Tabulation_Areas_ZCTA.shp")</pre>
## Reading layer `Colorado_ZIP_Code_Tabulation_Areas_ZCTA' from data source
##
     `C:\Users\frencjos\Documents\OneDrive - The University of Colorado Denver\GitHub\code_demonstration
     using driver `ESRI Shapefile'
## Simple feature collection with 526 features and 3 fields
```

Geometry type: MULTIPOLYGON

Dimension:

read multipolygon object

Bounding box: xmin: -109.2712 ymin: 36.84912 xmax: -101.9768 ymax: 41.12812

We first visualize co_zcta by using the sf::st_geometry function to extract its geometry object and then plot it using the plot function.

```
plot(st_geometry(co_zcta))
```

Next, we read in a csv file containing heavy metal data from the U.S. Geological Survey [https://pubs .usgs.gov/ds/520/] (Smith et al., 2010) using the read.csv function and assign it the name co_points. co_points is a data frame contains the measurements for 960 locations in Colorado; the data frame includes longitude/latitude coordinates for each measurement.

```
# read data frame
co_points <- read.csv("usgs_data_series_520_clean.csv")</pre>
class(co points)
## [1] "data.frame"
# column names
names(co points)
    [1] "laboratory_number" "field_number"
                                                    "latitude"
##
    [4] "longitude"
                              "date_collected"
                                                    "land_use"
##
   [7] "hg_mg_kg"
                              "al_pct"
                                                    "ca_pct"
## [10] "fe_pct"
                              "k_pct"
                                                    "mg_pct"
## [13] "na_pct"
                              "s_pct"
                                                    "ti_pct"
  [16] "ag_mg_kg"
##
                              "as_mg_kg"
                                                    "ba_mg_kg"
## [19] "be_mg_kg"
                              "bi_mg_kg"
                                                    "cd_mg_kg"
## [22] "ce_mg_kg"
                              "co_mg_kg"
                                                    "cr_mg_kg"
## [25] "cs_mg_kg"
                              "cu_mg_kg"
                                                    "ga_mg_kg"
## [28] "in_mg_kg"
                              "la_mg_kg"
                                                    "li_mg_kg"
## [31] "mn_mg_kg"
                              "mo_mg_kg"
                                                    "nb_mg_kg"
## [34] "ni_mg_kg"
                              "p_mg_kg"
                                                    "pb_mg_kg"
## [37] "rb_mg_kg"
                              "sb_mg_kg"
                                                    "sc_mg_kg"
## [40] "sn_mg_kg"
                              "sr_mg_kg"
                                                    "te_mg_kg"
## [43] "th_mg_kg"
                              "tl_mg_kg"
                                                    "u_mg_kg"
  [46] "v_mg_kg"
                              "w_mg_kg"
                                                    "y_mg_kg"
## [49] "zn_mg_kg"
                              "se_mg_kg"
We need to convert the co_points data frame to an sf object, which can be easily done using sf::st_as_sf.
The first argument is the object to be converted and the coords argument is the column names describing
the spatial locations associated with each row of the data frame.
# convert to sf object
co_points <- sf::st_as_sf(co_points, coords = c("longitude", "latitude"))</pre>
class(co_points)
## [1] "sf"
                      "data.frame"
Unfortunately, the coordinate reference system (CRS) of the co_zcta and co_points do not match. We can
see this using the sf::st_crs function.
st_crs(co_zcta)
## 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]]
st_crs(co_points)
```

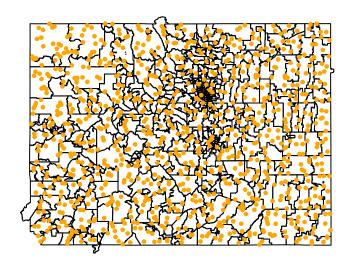
Coordinate Reference System: NA

The geometry objects for both co_zcta and co_points are both defined in terms of longitude/latitude coordinates, but have different CRSs. When sf objects have different CRSs, we can use sf::st_transform to transform one of the objects to the same CRS as the other object. However, since the CRS of co_points is NA, we will simply set the CRS of co_points to the CRS of co_zcta using sf::st_set_crs.

```
# set crs of co_points to same crs as co_zcta
co_points <- sf::st_set_crs(co_points, sf::st_crs(co_zcta))</pre>
```

To check whether there are any obvious problems with our two sf objects, we plot the geometry of both objects in one graphic. Note that we use add = TRUE in the second plot call to draw the geometry of co_points on the plot of co_zcta. The object seems to mesh together nicely

```
# plot geometry
plot(st_geometry(co_zcta))
# plot points on polygons
plot(st_geometry(co_points), add = TRUE, pch = 19, col = "orange", cex = 0.5)
```



To determine the points of co_points that are in the multipolygons contained in co_zcta, we can use sf::st_contains. The first argument is the object whose geometries contain the geometries of the second argument. We store the results in sf::st_contains returns a list; each element of the list indicates the points in co_points that are contained in the associated multipolygon of co_zcta. e.g., After assigning the list returned by sf::st_contains the name region_points_list, we see that the 200th point in co_points

is contained in the 4th multipolygon of co_zcta.

```
# determine which points are in which polygon of co_zcta
region_points_list <- sf::st_contains(co_zcta, co_points)
region_points_list[4]</pre>
```

```
## [[1]]
## [1] 200
```

Once we have the list of indices of the points contained in each multipolygon, we can use sapply and length to count the number of points in each multipolygon.

```
# convert list of points to count
region_count <- sapply(region_points_list, length)
head(region_count)</pre>
```

```
## [1] 0 0 1 1 11 1
```

We check the length of the geometry of co_zcta and the length of the counts to ensure they are the same length.

```
# make sure number of regions matches number of counts
length(st_geometry(co_zcta))
```

```
## [1] 526
```

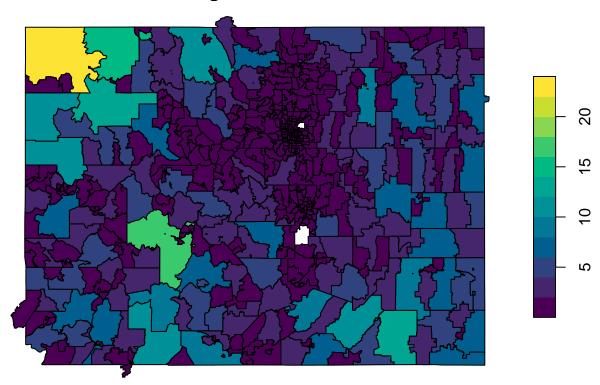
length(region_count)

```
## [1] 526
```

Since the lengths are the same, we add the region_count variable to the co_zcta sf object and plot the result.

```
# add region_count to co_zcta
co_zcta$region_count <- region_count
# plot region_count variable
plot(co_zcta["region_count"], pal = hcl.colors)</pre>
```

region_count



Alternatively, you may wish to identify which multipolygon of co_zcta contains each point in co_points. The sf::st_within function will perform this task. The first argument is the object whose geometries are in the second argument; in this case, the points in co_points that are within the multipolygons of co_zcta. We assign the list of results the name points_region_list.

```
# determine the multipolygon of co_zcta each point is in
points_region_list <- sf::st_within(co_points, co_zcta)</pre>
```

Each point should only be in no more than a single multipolygon. We compute the length of each element of the list using the sapply function and then check the range of the results. Each point is in 0 or 1 regions (there are some points outside the multipolygons in the bottom left of the graphic above.)

```
# double-check that each point is in at most 1 region
range(sapply(points_region_list, length))
```

[1] 0 1

Since each point is only in a single multiplygon, we use as.numeric to convert the list to a vector.

```
# since the point should only intersect a single region, convert to number within <- as.numeric(points_region_list)
```

To double-check that we have performed the desired task correctly, we check that the length of within is the same as the number of points in co_points.

```
# double-check that length(within) matches number of points
# add within column to co_points
length(within)
```

[1] 960

length(st_geometry(co_points))

[1] 960

Since the lengths do match, we add within as a column to co_points.

```
# add within column to co_points data frame
co_points$within <- within</pre>
```

References

Smith, D.B., Ellefsen, K.J., and Kilburn, J.E., 2010. Geochemical data for Colorado soils—Results from the 2006 state-scale geochemical survey: U.S. Geological Survey, Data Series 520, 9 p.

Pebesma, E., 2018. Simple Features for R: Standardized Support for Spatial Vector Data. The R Journal 10 (1), 439-446, [https://doi.org/10.32614/RJ-2018-009]