SF & Terra

Announcements

- Final Project Pre-Proposal Due today!
 - Again, no late penalties will be applied but this must be submitted before the Final Pre-Proposal due date, and give time for us to meet.
 - Not looking for anything crazy, and if you don't know what analysis you want to do its ok :)
- Final Project Meeting Calendar fixed
- Last class was only a few days ago, so we will have more time to do
 Labs 2 and 3, with next week's lab being very short to account for this.
- Don't wait to get started on these as they can be long depending on your experience with R.

What are SF & Terra?

- SF & Terra are R integrations of reading, writing, and manipulating vector and raster data respectively.
- SF (Simple Features): Vector
- Terra : Raster

Package History Background

- Packages and standards are always changing, which is why we need to stay up to date on what is new and in use!
- SF recently overtook the SP package, which is now depreciated
- Terra recently overtook the Raster package, which is now depreciated
- Some packages still depend on SP and Raster, and have not been updated to SF and Terra.
- Keep this in mind for your stack exchange answer interpretations.

1. Vectors in R w/ SF

Simple Features (SF) - Cheat Sheet

- st_read(): read simple features from a file.
- st_write(): write simple features to a file.
- st_as_sf(): converts objects to simple features objects.
- st_transform(): transform / reproject CRS.
- st_crs(): retrieves the CRS of an sf object.
- Geometric Binaries
 - st_intersection(): clip
 - o st_union()
 - st_difference(): erase
 - o st_buffer()
 - st_area()
 - st_length()

Anatomy of a Shapefile

- .shp The main file that stores the geometry of all features (points, lines, polygons).
- .shx The index file links the attribute data in the .dbf file to the spatial data in the .shp file.
- .dbf The dBASE table that stores attribute data for each shape.
- .prj The projection file that stores the coordinate system and projection information.
- .sbn and .sbx These are spatial index files that improve the performance of spatial queries. Not all GIS software will create or use these files.
- .xml An optional metadata file.

Reading Data - Shapefiles

- st_read()
- Note: You only need the .shp which will point sf to the other dependency files.
- Example:
 - o polygon = st_read(here::here('data', 'polygon.shp'))

Reading Data - Shapefile Example

Reading Data - Converting Data Frames to SF

- st_as_sf(dataframe, coords = c("Longitude","Latitude"), crs = ___)
- USE EPSG CODES FOR CRS ALWAYS (epsg.io)
- Example:

```
# Example dataframe

df <- data.frame(
    id = 1:3,
    lat = c(40.7128, 34.0522, 41.8781),
    long = c(-74.0060, -118.2437, -87.6298)
)

head(df)
```

```
> head(df)
id lat long
1 1 40.7128 -74.0060
2 2 34.0522 -118.2437
3 3 41.8781 -87.6298
```

Reading Data - Geodatabases

- 1. Inspect Geodatabase
 - > gdb_path = here::here('data', 'default.gdb')
 - > layers = st_layers(gdb_path)
 - > print(layers)
- Specify Layer in st_read()
 - > layer = layers[[1]][#] # subset for your layer of interest
 - > data = st_read(gdb_path, layer = layer)

Use GDB as input Specify Layer w/ layer parameter

Reading Data - Limited Geodatabases Example

 If you simply use the gdb location as the st_read() input, it will default to using the first layer in the gdb and ignore the rest.

```
> gdb = st_read(soils_gdb_path)
Multiple layers are present in data source C:\Users\James\Documents\Academic\SDAR\data
\Soils_MassGIS_GDB\Soils_MassGIS.gdb, reading layer `SOILS_MUPOLYGON_TOP20'.
Use `st_layers' to list all layer names and their type in a data source.
Set the `layer' argument in `st_read' to read a particular layer.
Reading layer `SOILS_MUPOLYGON_TOP20' from data source
   `C:\Users\James\Documents\Academic\SDAR\data\Soils_MassGIS_GDB\Soils_MassGIS.gdb'
   using driver `OpenFileGDB'
Simple feature collection with 249771 features and 44 fields
```

Reading Data - Geodatabases Example

1. Retrieve layers from gdb

```
> soils_gdb_path = here::here('data',
                              'Soils_MassGIS_GDB'.
                              'Soils_MassGIS.qdb')
> soils_gdb_layers = st_layers(soils_gdb_path)
> soils_qdb_layers
Driver: OpenFileGDB
Available layers:
                              geometry_type features fields
                layer_name
                              Multi Polygon 249771
     SOILS_MUPOLYGON_TOP20
                                                         44
                              Multi Polygon
2 SOILS_POLY_PRIMEFARMLAND
                                               94535
        SOILS_SPECFEAT_ARC Multi Line String
                                                4516
        SOILS_SPECFEAT_PT
                                      Point
                                               35542
                              Multi Polygon
  SOILS_SURVEY_AREAS_POLY
                                                  19
                                                         10
```

Reading Data - Geodatabases Example

2. Subset layer output list to layer of interest

Reading Data - Geodatabases Example

3. Utilize st_layers() to choose layer of interest from gdb

```
> layer = soils_gdb_layers[[1]][2]
> soils_primefarm = st_read(soils_gdb_path, layer = layer)
Reading layer `SOILS_POLY_PRIMEFARMLAND' from data source
    `C:\Users\James\Documents\Academic\SDAR\data\Soils_MassGIS_GDB\Soils_MassGIS.gdb'
    using driver `OpenFileGDB'
Simple feature collection with 94535 features and 7 fields
Geometry type: MULTIPOLYGON
Dimension: XY
Bounding box: xmin: 35513.3 ymin: 777671.1 xmax: 330291.3 ymax: 959747.1
Projected CRS: NAD83 / Massachusetts Mainland
```

> layer = layers[[1]][#] # subset for your layer of interest

Simple Feature Objects

- sf dataframes store spatial metadata like the CRS and the geometry
 - head(sf1, 3) will show you the metadata

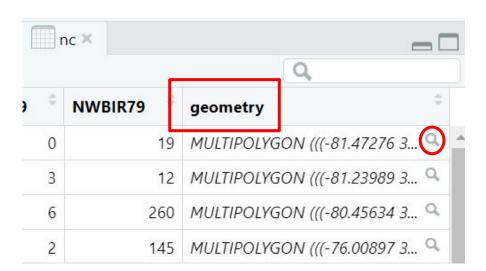
Simple Feature Objects - Inspecting Data Example

```
> head(nc, 3)
Simple feature collection with 3 features and 14 fields
Geometry type: MULTIPOLYGON
Dimension:
             XY
Bounding box: xmin: -81.74107 ymin: 36.23388 xmax: -80.43531 ymax: 36.58965
Geodetic CRS: NAD27
  AREA PERIMETER CNTY_ CNTY_ID
                                  NAME FIPS FIPSNO CRESS_ID BIR74 SID74
                                   Ashe 37009 37009
1 0.114
           1.442 1825
                         1825
                                                              1091
                                                                       1
2 0.061
           1.231 1827 1827 Alleghany 37005 37005
                                                           3 487
                                                                      0
3 0.143
       1.630 1828 1828
                                  Surry 37171 37171
                                                          86
                                                              3188
 NWBIR74 BIR79 SID79 NWBIR79
                                                 geometry
                         19 MULTIPOLYGON (((-81.47276 3...
      10 1364
      10
           542
                         12 MULTIPOLYGON (((-81.23989 3...
     208 3616
                        260 MULTIPOLYGON (((-80.45634 3...
```

Simple Feature Objects

- sf dataframes store spatial metadata like the CRS and the geometry
 - head(sf1, 3) will show you the metadata
- Geometry is stored in a list column!
- The geometry list column is a SFC class = Simple Features Column
 - st_geometry(sf1): Allows you to extract geometry to its own vector

Simple Feature Objects - SF Geometry Example



- The geometry column contains a list format of coordinates for each element.
- Remember that lines and polygons are made of points, so these geometry lists contain point info.

| Name | Type | Value |
|----------------|-----------------------------------|-------------------------------------|
| onc[[15]][[1]] | list [1] (S3: XY, MULTIPOLYGON, s | List of length 1 |
| ○ [[1]] | list [1] | List of length 1 |
| [[1]] | double [27 x 2] | -81.5 -81.5 -81.6 -81.6 -81.7 -81.7 |

Simple Feature Objects - SF Geometry Tips

- The geometry column will be retained through subsets
 - Keep this in mind if your input requires no geometry column
 - You can utilize the st_drop_geometry() function to remove geometry

CRS & Projections

- R cannot project 'on the fly' like ArcPro
 - Visually two layers may look properly projected, but to the ArcPro tools they are not!
 - ArcPro tools cannot project on the fly, so CRS/Projections must be manually consistent
- st_crs(): get CRS/Projection
- st transform(): project & reproject sf objects
- Example: We can ensure matching projectings with these two functions
 - o > crs_sf1 = st_crs(sf1)
 - > sf2_transformed = st_transform(sf2, crs = crs_sf1)
 - > st_crs(sf1) == st_crs(sf2)
 - > TRUE

Projections - Inspecting CRS Example

```
> st_crs(nc)
Coordinate Reference System:
  User input: NAD27
 wkt:
GEOGCRS ["NAD27",
    DATUM["North American Datum 1927",
        ELLIPSOID["clarke 1866",6378206.4,294.978698213898,
            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", 4267]]
```

Projections - Transforming with EPSG Codes

```
> st_transform(nc, crs = "EPSG:4326")
Simple feature collection with 100 features and 14 fields
Geometry type: MULTIPOLYGON
Dimension:
              XY
Bounding box: xmin: -84.32377 ymin: 33.88212 xmax: -75.45662 ymax: 36.58973
Geodetic CRS: WGS 84
First 10 features:
                                          FIPS FIPSNO CRESS ID BIR74 SID74
   AREA PERIMETER CNTY_ CNTY_ID
                                      NAME
            1.442
1 0.114
                  1825
                          1825
                                      Ashe 37009
                                                 37009
                                                                1091
  0.061
            1.231
                  1827
                          1827
                                 Alleghany 37005
                                                 37005
                                                                 487
                                                                         0
  0.143 1.630 1828
                          1828
                                     Surry 37171
                                                 37171
                                                             86 3188
                          1831
                                 Currituck 37053
                                                 37053
                                                             27
4 0.070
        2.968 1831
                                                                  508
  0.153
        2.206 1832
                          1832 Northampton 37131
                                                 37131
                                                             66 1421
  0.097
            1.670 1833
                          1833
                                  Hertford 37091
                                                 37091
                                                             46 1452
                          1834
                                                                  286
  0.062
            1.547 1834
                                    Camden 37029
                                                 37029
                                                             15
            1.284 1835
                                                                         0
8 0.091
                          1835
                                    Gates 37073
                                                 37073
                                                             37
                                                                 420
  0.118
            1.421 1836
                          1836
                                    Warren 37185
                                                 37185
                                                             93
                                                                  968
                                                             85
                                                                1612
10 0.124
            1.428 1837
                          1837
                                    Stokes 37169
                                                 37169
```

Geometric Binaries - Cheat Sheet

```
st_intersection(sf1, sf2) : clip

st_difference(sf1, sf2) : erase

st_union(sf1) or st_union(sf1, sf2)

st_buffer(sf1, buf_dist)
```

Distance unit is based on CRS

Geometric Operations - Cheat Sheet

```
st_join(sf1, sf2) or st_merge(sf1, sf2)
st_centroid(sf1): will find centroids for your
st_area(sf1)
st_length(sf1)
st_distance(sf[1,], sf1[2,])
```

Simple Feature Manipulation - Geometric Integrity

- Features may not always be valid by the standards of SF
 - Self-Intersecting Polygons
 - Non-Closed Polygons
 - Duplicate Points
 - Incorrect Ring Ordering
 - Holes Outside of Polygons
 - Overlapping or Duplicate Rings
 - Zero-Area Geometries
 - Invalid Geometric Constructions
- st_is_valid(sf) # Determines validity of features in SF
 - Will return TRUE if feature is valid
 - Will return FALSE if feature is invalid
- st_make_valid(sf) # Fixes invalid features in SF

Simple Feature Manipulation - st_cast()

- st_cast()
- This function is particularly useful when you want to change the type of geometry of an sf object, for example, converting a multipolygon to individual polygons, or a multiline to individual lines.
 - Some functions require a specific sf geometry type, so this is helpful to wrangle data!
- Example:
 - > sf_polygon <- st_cast(sf_multipolygon, "POLYGON")

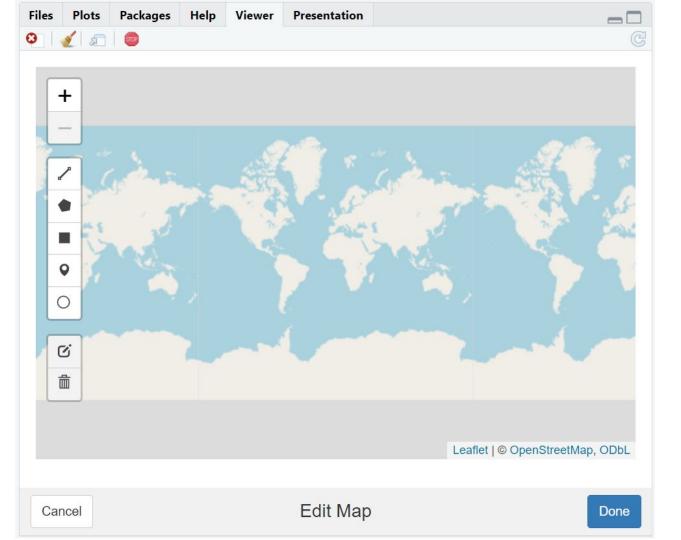
Writing SF Data

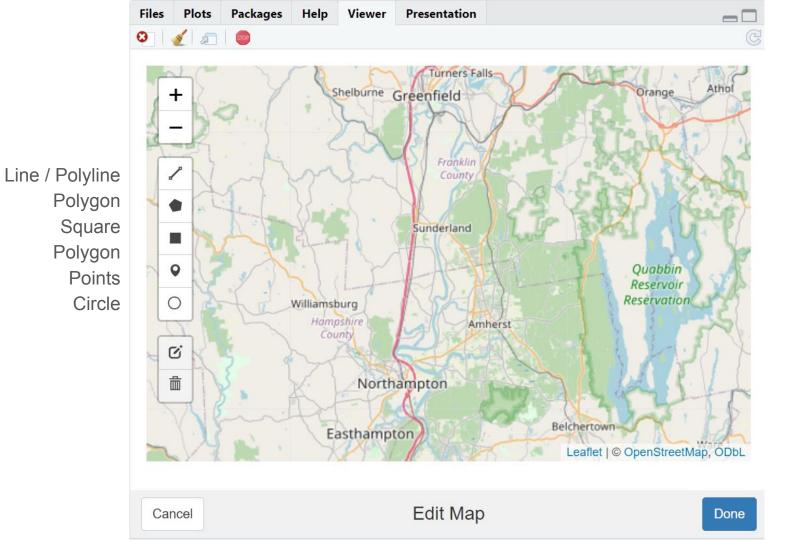
- st_write()
- Example:
 - > st_write(points, "points.shp")
- Write points to .csv w/ coordinates
 - > st_write(points, "points.csv",
 - o layer_options = "GEOMETRY=AS_XY")

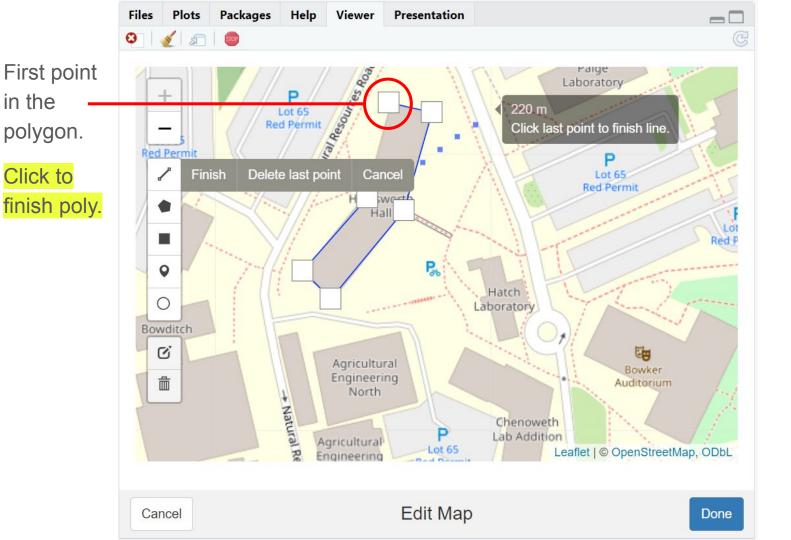
2. Leaflet & mapedit

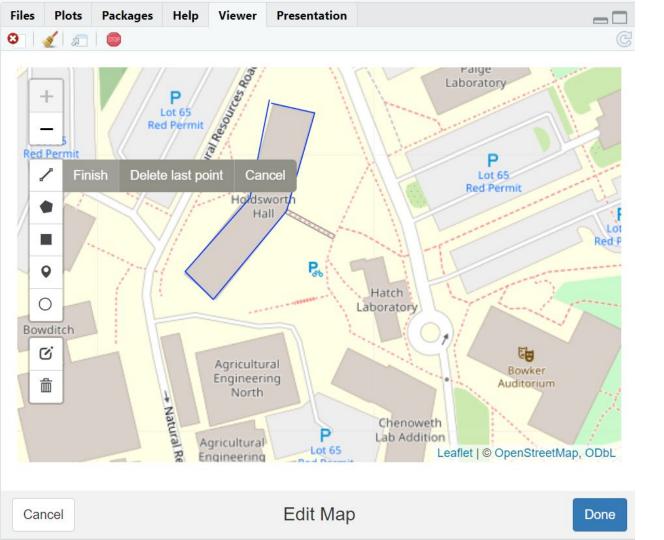
Creating Vectors in R

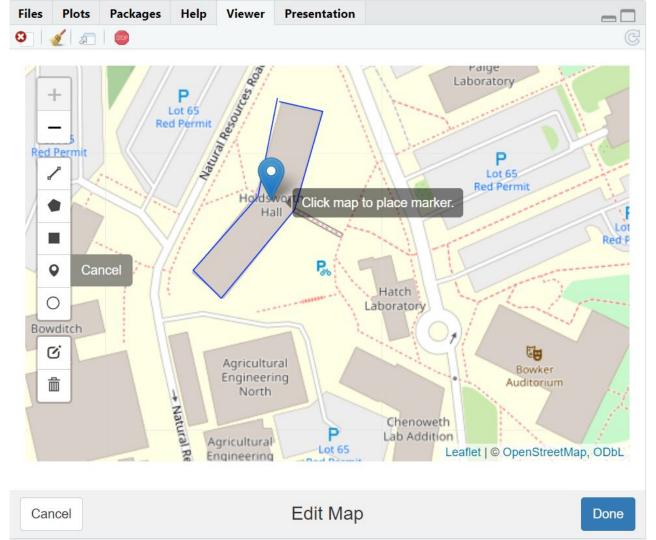
- Leaflet is a free and open-source mapping project to make interactive maps
- mapedit is a package which leverages leaflet to create a mapping UI within R to create points, lines, and polygons.
- These are useful if you don't want to open a new QGIS/ArcPro session just to create a vector.

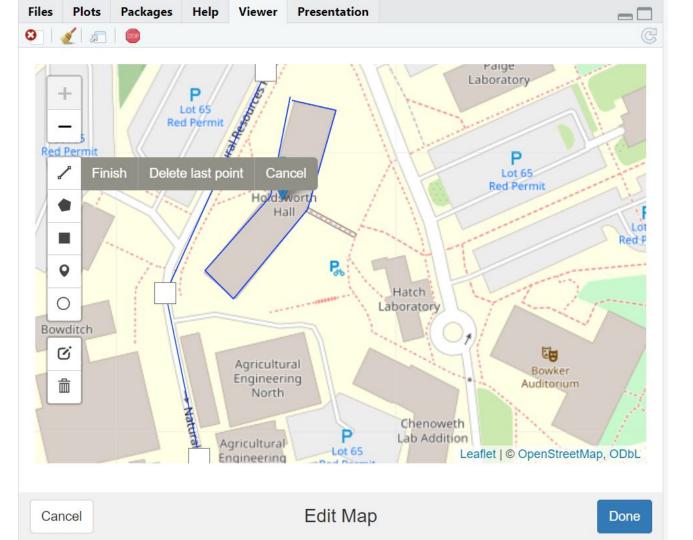












How to setup mapedit!

```
# Prepare a leaflet map for mapedit
my_map = leaflet() %>% addTiles()

# Start a mapedit session
drawn_features = editMap(my_map)
```

Working with mapedit output

| Name | Туре | Value | | |
|----------------|-----------------------------------|--|--|--|
| drawn_features | list [5] | List of length 5 | | |
| drawn | list [3 x 3] (S3: sf, data.frame) | A data.frame with 3 rows and 3 columns | | |
| edited | NULL | Pairlist of length 0 | | |
| deleted | NULL | Pairlist of length 0 | | |
| finished | list [3 x 3] (S3: sf, data.frame) | A data.frame with 3 rows and 3 columns | | |
| all | list [3 x 3] (S3: sf, data.frame) | A data.frame with 3 rows and 3 columns | | |

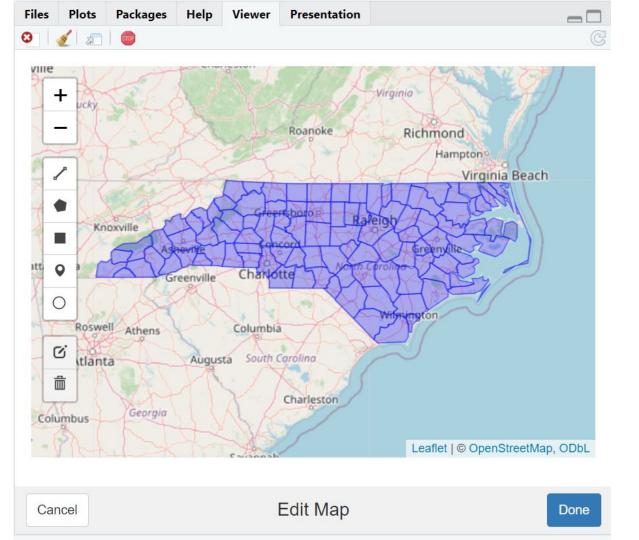
- The default mapedit output is a list, which is harder to work with in R
- But we can convert it to an sf dataframe!

Working with mapedit output cont.

```
# Convert drawn_features list to an sf dataframe
drawn_features_sf = drawn_features$finished
plot(drawn_features_sf[2])
                                   feature_type
                                                           pol
                                                          mar
```

Leaflet Layers

- You can utilize your own layers by adding to the leaflet map
- Note: You will need to transform your layer to WGS84 in order to add it to the leaflet interactive map!



3. Rasters in R w/ Terra

Terra Essentials - Cheat Sheet

- rast(): read in a raster
- writeRaster(): write a raster output
- app(): apply a function on a raster object
- global(): apply global statistics on a raster object
- mask(): mask a raster using another raster or polygon (essentially a clip)
- merge(): merge multiple rasters
- aggregate(): aggregate raster data to a lower resolution.
- disagg(): disaggregate raster data to a higher resolution.
- crs() + project(): get CRS and reproject rasters.

Reading Rasters

- rast()
- Example:
 - > raster = rast(here::here('data', 'raster.tif'))
- Note: R has a weird time retaining SpatRasters after you shutdown R. If you
 are getting empty plots or weird errors for your SpatRaster operations, just try
 reloading the SpatRaster.

Inspecting Rasters - Cheat Sheet

- summary(): returns general stats on the raster
- cellSize(): returns general info like dimensions of cells in raster
- ext(): returns extent of raster layer

Inspecting Rasters - summary(raster) Example

```
> summary(p80_1)
wc2.1_10m_prec_1980.01.tif
Min. : 0.00
1st Qu.: 7.43
Median : 21.07
Mean : 53.18
3rd Qu.: 56.37
Max. :930.36
NA's :75438
```

Inspecting Rasters - cellSize(raster)

- cellSize() will give you all the essential information on your raster
 - Dimensions: The number of rows, columns, and layers (bands) in the raster.
 - Resolution: The size of each cell in units of the coordinate reference system.
 - Spatial Extent
 - Coordinate Reference System (CRS)
 - Source: Information about the source of the raster data.
 - Basic statistics for each layer, such as the min, max, mean, and sd.
- This is useful, but knowing how to retrieve this information individually is better done
 with the specific commands like ext() or crs().

Inspecting Rasters - cellSize()

cellSize() returns the size of each cell in your Raster

```
> cellSize(p80_1)
class : SpatRaster
dimensions : 1080, 2160, 1 (nrow, ncol, nlyr)
resolution: 0.1666667, 0.1666667 (x, y)
extent : -180, 180, -90, 90 (xmin, xmax, ymin, ymax)
coord. ref. : lon/lat WGS 84 (EPSG:4326)
source(s) : memory
           : wc2.1_10m_prec_1980-01
varname
name
                   area
min value : 504025.1
max value : 341918442.3
```

Inspecting Rasters - ext()

```
> ext(p80_1)
SpatExtent : -180, 180, -90, 90 (xmin, xmax, ymin, ymax)
```

This is an alternative to cellSize() if you only wanted the extent

Terra Essentials - Cheat Sheet

- rast(): read in a raster
- writeRaster(): write a raster output
- app(): apply a function on a raster object
- global(): apply global statistics on a raster object
- mask(): mask a raster using another raster or polygon (essentially a clip)
- merge(): merge multiple rasters
- aggregate(): aggregate raster data to a lower resolution.
- disagg(): disaggregate raster data to a higher resolution.
- crs() + project(): get CRS and reproject rasters.

Raster Calculations - global()

- global() can perform a variety of summary operations.
- Example: Let's say you want to calculate the mean of all the raster values.
 - o > mean_raster = global(raster, fun="mean")
- fun specifies the function you want to apply.
 - o "mean"
 - o "sum"
 - "min" / "max"
 - o "range"
 - o "prod"
 - o etc.

Raster Calculations - Global Stats

```
# Example: Get mean of global values in a raster
mean_prec_jan = global(p80_1, fun = mean, na.rm = TRUE)
print(mean_prec_jan)
```

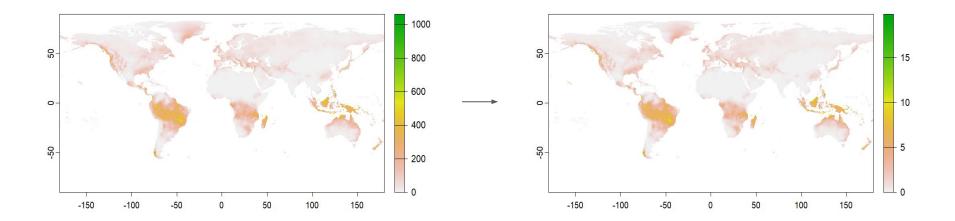
```
mean
wc2.1_10m_prec_1980-01.tif 53.3882
```

Raster Calculations - app()

- app()
- Example:
 - > app_calc = app(raster, fun = FUNCTION, na.rm=TRUE)
- Note: app() can leverage parallel processing!

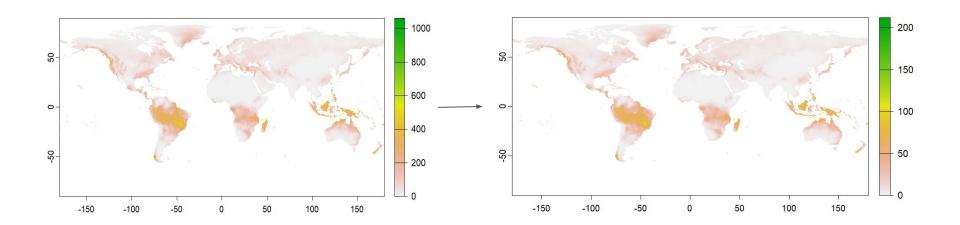
Raster Calculations - Single Raster Example

```
# Example: Simple Raster Arithmetic
p80_1_stand = p80_1 / mean_prec_jan[1,1]
plot(p80_1_stand)
```



Raster Calculations - Single Raster Custom Example

```
# Example: Custom Raster Calculation Functions with app() p80_1_customfun = app(p80_1, fun = function(x) { x / 10 * 2}) plot(p80_1_customfun)
```

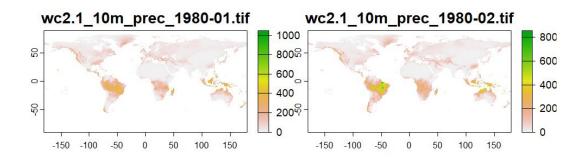


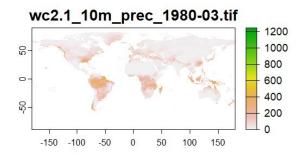
Raster Stacks

- A raster stack is a SpatRaster that contains multiple rasters in it!
- Useful for multilayer spatial analysis and calculations.
- Some packages expect a raster stack as an input.
- Example:
 - o > raster1 = rast(here::here('data', 'X.tif')
 - > raster2 = rast(here::here('data', 'Y.tif')
 - > raster3 = rast(here::here('data', 'Z.tif')
 - > raster_stack = c(raster1, raster2, raster3)

Raster Stacks - Plotting

 Plotting a raster stack will typically plot all the rasters within the stack individually.





Raster Stack Calculations - app()

- app()
 - Purpose: Designed to apply a function to each layer of a raster stack individually, or to each cell of a single layer.
 - Custom and Complex Operations: It's particularly useful for applying custom functions that are not inherently vectorized or when the operation depends on other factors (like conditional operations based on cell values).
 - Good for complex functions that involve conditional logic and looping etc.
 - Output depends on the function being applied.
 - It can be a raster layer, a raster stack, or even non-raster outputs if the function is designed that way.

Raster Stack Calculations - Example

```
    app(raster_stack, fun = function)
    # Define a custom function to convert from Fahrenheit to Celsius
    > convert_to_celsius <- function(fahrenheit) {</li>
    celsius <- (fahrenheit - 32) * 5/9</li>
    return(celsius)
    }
    # Apply the function to each layer
    > adjusted_temp_stack <- app(raster_stack, fun=convert_to_celsius)</li>
    # Output will be a single raster stack
    # but with values of each layer converted to Celsius
```

aggregate()

- Used for summarizing or aggregating data based on certain criteria, such as finding averages, sums, or counts within grouped data.
- Basic Syntax: aggregate(x, by, FUN)
 - x: The data to be aggregated.
 - by: A list of grouping variables.
 - The aggregation is performed separately for each combination of these variables.
 - FUN: The function to be applied for aggregation, such as mean, sum, max, etc.

aggregate() Example

```
> head(state.x77)
          Population Income Illiteracy Life Exp Murder HS Grad Frost
                                                                  Area
               3615
                      3624
                                 2.1
                                       69.05
                                              15.1
                                                      41.3
                                                                50708
Alabama
Alaska 
                365
                      6315
                                 1.5
                                       69.31
                                              11.3
                                                      66.7
                                                            152 566432
Arizona
               2212
                      453Ø
                                1.8
                                       70.55
                                             7.8
                                                      58.1
                                                             15 113417
Arkansas
               2110
                      3378
                                1.9
                                       70.66 10.1
                                                      39.9
                                                             65 51945
California
                      5114
                                1.1
                                       71.71
                                              10.3
                                                     62.6
                                                             20 156361
              21198
Colorado
               2541
                      4884
                                 Ø.7
                                       72.06
                                               6.8
                                                      63.9
                                                            166 103766
```

```
# Compute the averages for the variables in 'state.x77', grouped # according to the region (Northeast, South, North Central, West) that # each state belongs to.
```

aggregate(state.x77, list(Region = state.region), mean)

| Region | Population | Income | Illiteracy | Life [‡] Exp | Murder | HS [‡] Grad | Frost | Area ‡ |
|---------------|------------|----------|------------|--------------------------|-----------|-------------------------|----------|-----------|
| Northeast | 5495.111 | 4570.222 | 1.000000 | 71.26444 | 4.722222 | 53.96667 | 132.7778 | 18141.00 |
| South | 4208.125 | 4011.938 | 1.737500 | 69.70625 | 10.581250 | 44.34375 | 64.6250 | 54605.12 |
| North Central | 4803.000 | 4611.083 | 0.700000 | 71.76667 | 5.275000 | 54.51667 | 138.8333 | 62652.00 |
| West | 2915.308 | 4702.615 | 1.023077 | 71.23462 | 7.215385 | 62.00000 | 102.1538 | 134463.00 |

Scaling Rasters - terra::aggregate() & disagg()

- Changing raster resolutions can be the difference between your code running or not!
- Hardware limitations are very real in science, and limits on RAM are especially troublesome when doing raster analytics.
 - RAM = Random Access Memory: Temporary, fast computer memory
- aggregate() downsamples your data to a more coarse resolution
- disagg() upsamples your data to a finer resolution

aggregate() & disagg() Examples

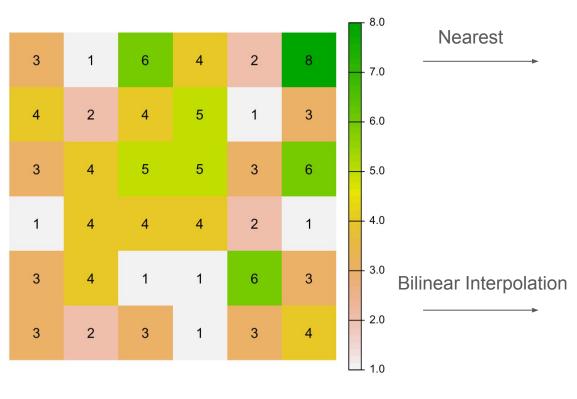
aggregate()

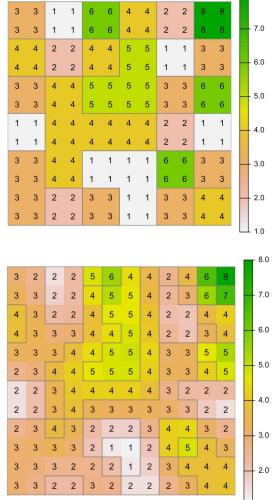
- > # Downsample your raster by a factor of 2
- > terra::aggregate(raster, fact=2)
- > # A 100x100 raster would downsample to 50x50 cells

disagg()

- > # Upsample your raster by a factor of 2
- > disagg(raster, fact=2, method="near")
- > # A 100x100 raster would upsample to 200x200 cells

disagg() - Raster Upsample





write.Raster()

- Used to write your raster to the format of your liking
- Input the SpatRaster and the name of the output you want
 - Note: Must specify the data type you want to output as in the name!
- Example:
 - write.raster(raster, "name_of_raster.tif")
 - write.raster(raster, "name of raster.ascii")

Projecting Rasters - Example

```
p80_1_wgs84 = project(p80_1, "EPSG:4326", method = 'bilinear')
```

methods

- near: nearest neighbor. This method is fast, but not a good choice for continuous values.
- o bilinear: bilinear interpolation
- o cubic: cubic interpolation.
- o cubicspline: cubic spline interpolation.
- lanczos: Lanczos windowed sinc resampling.

Other useful Terra Functions

- contour(): Generate contours for your SpatRaster
- terrain(): Terrain analytics for your SpatRaster
- Terra can handle vectors as well (SpatVectors)
 - interpIDW(): Inverse Distance Weighting Interpolation
 - interpNear(): Nearest Neighbor Interpolation
 - interpolation(): General Interpolation Function
 - o cartogram(): Map where the area of polygons is made proportional to another variable.