DAYMET PRECIPITATION ANALYSIS

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INTRO & PURPOSE

Exploratory data analysis of NASA's DAYMET precipitation data

Long-term goal: determine areas with the highest number of instances where precipitation exceeds 50.8 mm/day or 2 in/day during summer months from 2000-2020

Primary Goal: programmatically subset DAYMET data according to selected variable of interest, study area, and time range.

Study Area: State of Georgia – Census Tract level resolution.

ENVIRONMENT

Geopandas – data manipulation and cleanup

Utilized a combination of Jupyter notebooks and ArcGIS to perform analysis and visualization

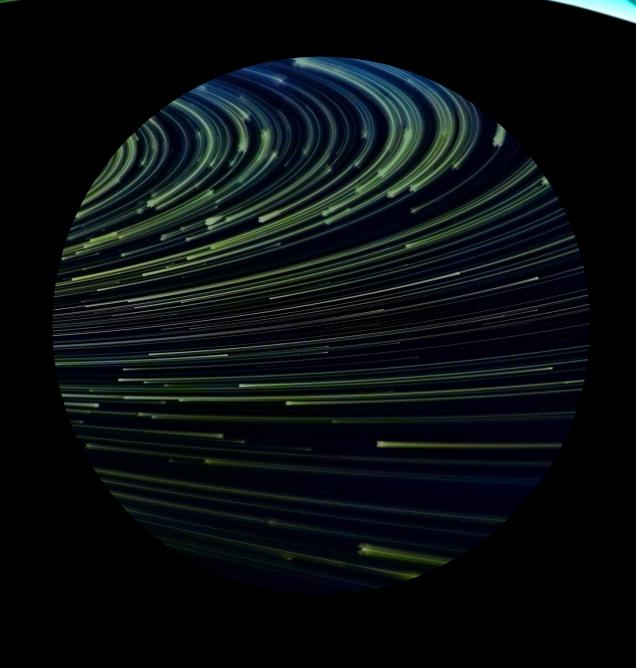
Rasterio – used for processing raster data

Main packages used:

Xarray – data manipulation

Shapely – for manipulation and analysis of planar geometric objects

Pydap – used to inspect data from any scientific dataset available on the internet from OPeNDAP servers.



PROCEDURE

- 1. Import required modules into Jupyter notebook environment
- 2. Set study area and subset parameters
 - a. Read in Georgia state boundary shapefile

```
ga_shape = gpd.read_file(r'C:\Users\samsh\OneDrive\Documents\Productivity\GIS\Research Projects\Marshep\Data\GA_stateBour
          2 # ga shape = gpd.read file(r'C:\Users\samsh\OneDrive\Documents\Productivity\GIS\Research Projects\Marshep\Data\tl 2019 1
          3 ax = ga_shape.plot()
            ga shape.crs
Out[3]: <Geographic 2D CRS: EPSG:4326>
        Name: WGS 84
        Axis Info [ellipsoidal]:
        - Lat[north]: Geodetic latitude (degree)
        - Lon[east]: Geodetic longitude (degree)
        Area of Use:
        - name: World
        - bounds: (-180.0, -90.0, 180.0, 90.0)
        Datum: World Geodetic System 1984
        - Ellipsoid: WGS 84
        - Prime Meridian: Greenwich
         35
         34
         33
         32
         31
```

- Set geographic bounding box used for searching NASA's metadata repository
- Create bounding box that aligns with DAYMET's CRS

```
1 # ga shape 4269 = ga shape.to crs(epsg=4326) # Dont need to reproject because already in WGS84 and no projection applied
In [4]: ▶
             2 xy1 = ga shape.bounds
             3 print(xy1) # Look at creating complex polygon so that the geometry does not change subsetting upon DAYMET overlay
                    minx
                               miny
                                          maxx
                                                    maxy
            0 -85.606675 30.356734 -80.885553 35.00118
In [5]: ▶
             1 xy1 = ga shape.bounds.values.tolist()[0] # We'll need the bounding box as a Python list
                                                              # to server as a subsetting parameter
             3 # xy1 = [-85.605165, 30.357851, -80.839729, 35.000659]
              4 print(xy1)
            [-85.606675, 30.356734, -80.885553, 35.00118]
In [6]: ▶
             1 #defining Daymet proj - we'll use this in a later step
             2 daymet proj = "+proj=lcc +ellps=WGS84 +a=6378137 +b=6356752.314245 +lat 1=25 +lat 2=60 +lon 0=-100 +lat 0=42.5 +x 0=0 +y
                ga_shape_lcc = ga_shape.to_crs(daymet_proj) # to_crs re-projects from NAD 1983 to LCC
             4 | lccbounds = ga shape lcc.bounds # Bounds in LCC projection
               lccbounds.round(2)
   Out[6]:
                    minx
                              minv
                                       maxx
                                                 maxy
             0 1259685.37 -1129705.22 1738456.68 -646896.28
```

Picture shapefile in original and transformed CRS

```
In [7]: ▶
              1 fig, (ax1, ax2) = plt.subplots(nrows=1, ncols=2, figsize=(12, 8))
              2 #ga_shape.plot(ax=ax1, facecolor='blue');
              3 ga shape.plot(ax=ax1, facecolor='green');
              4 ax1.set_title("GA Geographic Coordinate System");
              5 ga_shape_lcc.plot(ax=ax2, facecolor='gray');
              6 ax2.set_title("GA Daymet LCC Projection");
              7 plt.tight_layout()
                             GA Geographic Coordinate System
                                                                                          GA Daymet LCC Projection
              35
                                                                       -0.7
              34
                                                                       -0.8
              33
                                                                       -0.9
              32
                                                                       -1.0
              31
                                                                       -1.1 -
                                                                                1.3
                                                                                                              1.6
                                                                                                                        17
                                                                                                                             1e6
                                                                -81
```

Specify time range of interest:

```
In [8]: N

1     start_date = dt.datetime(2015, 1, 1) # specify your own start date
2     end_date = dt.datetime(2018, 12, 31) # specify your end start date
3     dt_format = '%Y-%m-%dT%H:%M:%SZ' # format requirement for datetime search
5     temporal_str = start_date.strftime(dt_format) + ',' + end_date.strftime(dt_format)
6     var = 'prcp' # select a Daymet variable of interest
8     print(temporal_str)
9     print(var)

2015-01-01T00:00:00Z,2018-12-31T00:00:00Z
prcp
```

Create URL used to enact data access request:

```
In [9]: ▶
              1 daymet doi = '10.3334/ORNLDAAC/1840' # define the Daymet V4 Daily Data DOI as the variable `daymet doi`
              2 cmrurl='https://cmr.earthdata.nasa.gov/search/' # define the base url of NASA's CMR API as the variable `cmrurl`
              3 doisearch = cmrurl + 'collections.json?doi=' + daymet doi # Create the Earthdata Collections URL
              4 print('Earthdata Collections URL: Daymet V4 Daily -->', doisearch)
             Earthdata Collections URL: Daymet V4 Daily --> https://cmr.earthdata.nasa.gov/search/collections.json?doi=10.3334/ORNLDAAC/1
             840
              1 # From the doisearch, we can obtain the ConceptID for the Daymet V4 Daily data
In [10]:
              2 # We'll search the json response of the Daymet metadata for "id" within the 'entry' dictionary key
              3 response = requests.get(doisearch)
              4 | collection = response.json()['feed']['entry'][0]
              5 #print(collection)
              6 concept id = collection['id']
              7 print('NASA Earthdata Concept ID --> ' , concept id)
            NASA Earthdata Concept ID --> C2031536952-ORNL CLOUD
```

C2031536952-ORNL_CLOUD is the unique NASA-given Concept ID for the Daymet V4 Daily data Collection. We'll use this to search for Daymet V4 Daily files (granules) that match our search criteria.

Create URL used to enact data access request:

2.2.a. We'll build a Request URL granulesearch to create a listing of all the granules (files) in NASA's Earthdata holdings that fit the search criteria we defined.

https://cmr.earthdata.nasa.gov/search/granules.json?collection_concept_id=C2031536952-ORNL_CLOUD&page_size=1000&temporal=2015-01-01T00:00:00Z,2018-12-31T00:00:00Z&bounding box[]=-85.606675,30.356734,-80.885553,35.00118

2.2.b. Again using Python's requests library, we can provide the URL granulesearch to create a listing of all the granules (files) in NASA's Earthdata holdings that fit the search criteria we defined.

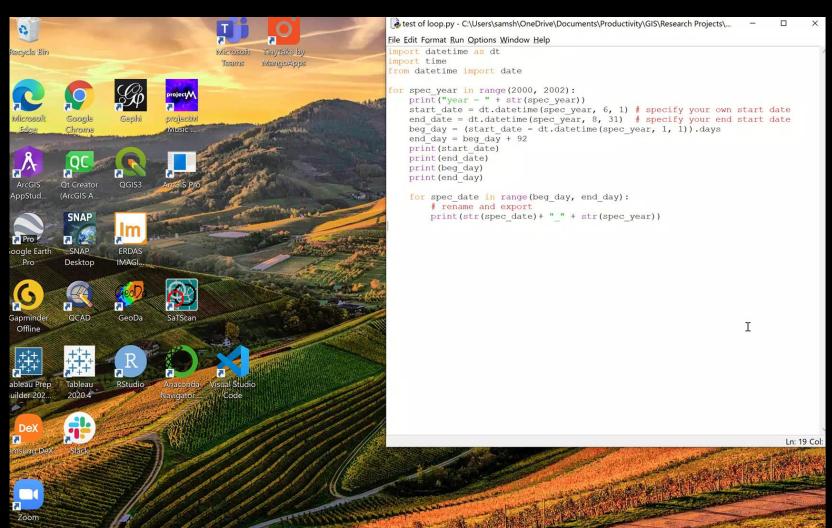
```
Daymet_Daily_V4.daymet_v4_daily_na_prcp_2015.nc
Daymet_Daily_V4.daymet_v4_daily_na_prcp_2016.nc
Daymet_Daily_V4.daymet_v4_daily_na_prcp_2017.nc
Daymet_Daily_V4.daymet_v4_daily_na_prcp_2018.nc
```

Enact request and retrieve data:

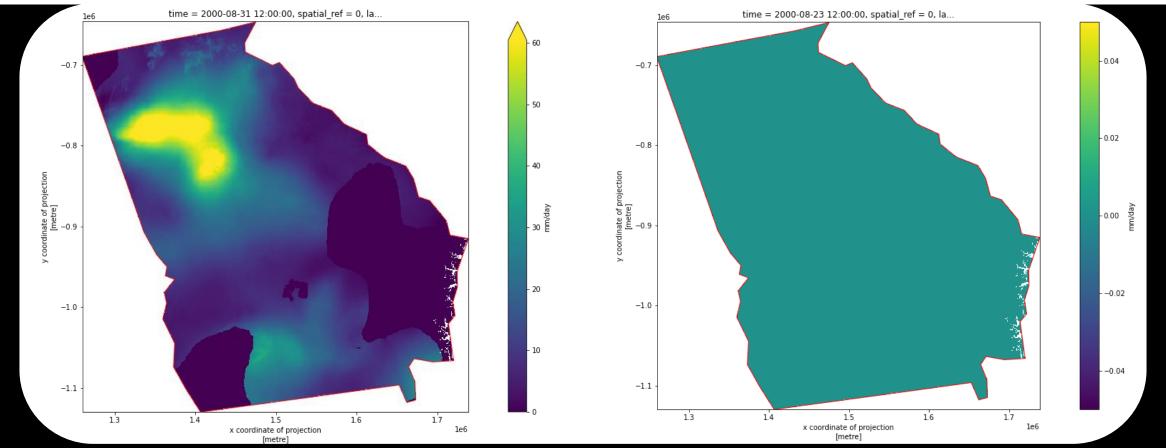
```
In [14]: H
               1 #from pydap.client import open url
               2 #import xarray as xr
               3 #import time
                 thredds url = 'https://thredds.daac.ornl.gov/thredds/dodsC/ornldaac/1840/' # ORNL DAAC TDS OPeNDAP URL
                                                                                             # for Daymet V4 Daily Files
               8 before = time.time()
               9 cnt = 0
              10 for g name in granule names:
                     print(' ***GRANULE NAME*** ---->', g name)
              11
              12
                     granule_dap = thredds_url + g_name.replace('Daymet_Daily_V4.','')
              13
                     print(granule dap)
              14
              15
                     # Using pydap's open url
                     thredds ds = open url(granule dap)
              16
              17
              18
                     # Xarray DataSet - opening dataset via remote OPeNDAP
                     ds = xr.open_dataset(xr.backends.PydapDataStore(thredds_ds), decode_coords="all")
              19
              20
              21
                     temp=ds['prcp'].sel(x=slice(lccbounds.minx[0],lccbounds.maxx[0]), y=slice(lccbounds.maxy[0],lccbounds.miny[0]))
              22
              23
                     if cnt==0:
              24
                          prcp = temp
              25
                     else:
              26
                         prcp = xr.concat([prcp, temp], dim="time")
              27
              28
                     cnt += 1
              29
              30
              31 # save to netcdf
              32 prcp.to_netcdf(var + '_tdssubset.nc')
              33 print("Processing Time: ", time.time() - before, 'seconds')
              34 # Processing Time: 50.4509379863739 seconds
```

PROCEDURE

Enact request and retrieve data:



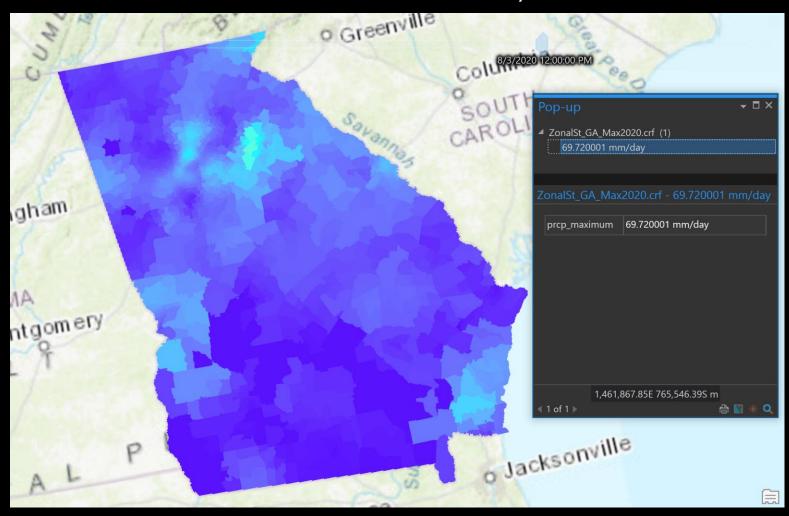
Visualize and Clip Daymet Precipitation Subset:



- Create function that automatically subsets and exports data for all days of interest
 - Interested in summer months from 2000-2020

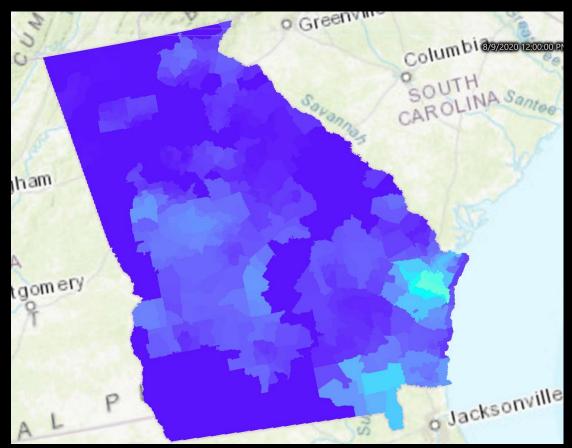
```
In [15]: ▶
              1 for spec year in range(2016, 2018):
                         print("year = " + str(spec year))
                          start date = dt.datetime(spec year, 6, 1) # specify your own start date
                          end date = dt.datetime(spec_year, 7, 1) # specify your end start date
                          beg day = (start date - dt.datetime(spec year, 1, 1)).days
                          end day = beg day + 92
                          print(start date)
                          print(end date)
                         print(beg day)
                         print(end day)
              10
              11
                         for spec day in range(beg day, end day):
              12
                             # rename and export
              13
                               print(str(spec day)) #spec day will be the value to use to specify what time equals a particular day
              14 #
              15
                             # The line below is the likely result of the inaccurate specification of year when automating the subset.
                               prcp 1day export = prcp clip.isel(time=spec day)
              16 #
                             prcp_1day_export = prcp_clip.sel(time=slice(start_date, end_date))
              17
                             print(str(prcp 1day export))
              18
                             prcp_1day_export.rio.to_raster("prcp_1day_export" + "_" + str(spec_day) + "_" + str(spec_year) + ".tif")
              19
                             fig, ax = plt.subplots(figsize = (20,10))
              20
                             prcp_1day_export.plot(ax=ax, robust=True, cbar_kwargs={'label': 'mm/day'})
              21
                             ga_shape_lcc.plot(ax = ax, color = 'none', edgecolor = 'red')
              22
                             fig.savefig("prcp_1day_export" + "_" + str(spec_day) + "_" + str(spec_year) + ".png")
              23
              24
              26 | print("Finished")
```

- Bring data into ArcGIS Pro
- Conduct zonal statistics for each year at Census tract level



Pictured: 8/3/2020

- Final step is to aggregate daily data to provide summary statistics for summer months of each year
- <u>Currently underway</u>: ArcGIS has ability to process multidimensional data but working with data that has a time dimension must be done by hand presently



QUESTIONS?