```
# Import packages
import os
import glob

import matplotlib.pyplot as plt
from mpl_toolkits.axes_grid1 import make_axes_locatable

import numpy as np
import rasterio
from rasterio import transform
from rasterio.plot import show_hist
from rasterio.plot import show
import xarray
```

### **Question 1**

# a) What is the air temperature (in F) and cloud cover (in %) in Florence, OR (in 2020) on January 31, 2020?

```
# Read data
xds = xarray.open_dataset(filepath + 'era/usa_t2m_tcc_2020.nc',
decode_coords='all')
xds_daily = xds.resample(time='1D').mean()
# Next, find the index of the grid point nearest a specific lat/lon -- in
this case, the city of Florence
florence_weather = xds_daily.sel(latitude=43.974659, longitude=-124.103142,
method='nearest')

# Convert temperature values to Fahrenheit
fahrenheit = (florence_weather['t2m'][31].values - 273.15) * 9/5 + 32

# Note: Jan 31 is DOY 31
print('On January 31, 2020 air temperature in Florence, OR was %.2f F and the
cloud cover was %.2f %%' % (fahrenheit, florence_weather['tcc'][31].values *
100))

On January 31, 2020 air temperature in Florence, OR was 54.04 F and the cloud
cover was 99.88 %
```

# b) What is the air temperature (in F) and cloud cover (in %) in Eugene, OR (in 2020) on February 15, 2020?

```
# Next, find the index of the grid point nearest a specific lat/lon -- in
this case, the city of Eugene
eugene_weather = xds_daily.sel(latitude = 44.052071, longitude = -123.086754,
method='nearest')

# Note: February 15th is DOY 46
# Convert temperature values to Fahrenheit
eug_fahrenheit = (eugene_weather['t2m'][46].values - 273.15) * 9/5 + 32
```

```
print('On February 15, 2020 air temperature in Eugene, OR was %.2f F and the
cloud cover was %.2f %%' % (eug_fahrenheit, eugene_weather['tcc'][46].values
* 100))
```

On February 15, 2020 air temperature in Eugene, OR was 43.79 F and the cloud cover was 97.51 %

### **Question 2**

# a) Find grid cell and rough location of the highest average air temperature (i.e. hottest place)

```
hottest_place = np.mean(xds_daily['t2m'], axis=0).argmax()
index = np.unravel_index(hottest_place, np.mean(xds_daily['t2m'],
axis=0).shape)
xds_daily['t2m'].shape
(366, 99, 233)
np.mean(xds_daily['t2m'], axis = 0).shape
(99, 233)
hottest_long = np.mean(xds_daily['t2m'], axis = 0)[81,55].longitude.values
hottest_lat = np.mean(xds_daily['t2m'], axis = 0)[81,55].latitude.values
print('The coordinates of the hottest place are %.2f, %.2f.' % (hottest_lat,
hottest_long))
The coordinates of the hottest place are 28.99, -111.25.
```

Reverse geocoding the above coordinates shows that the hottest place in this dataset is located at: Plan de Ayala, 83325, Hermosillo, Sonora, Plan de Ayala Hermosillo, México.

# b) Find grid cell and rough location of the lowest average air temperature (i.e. coldest place)

```
coldest_place = np.mean(xds_daily['t2m'], axis=0).argmin()
index = np.unravel_index(coldest_place, np.mean(xds_daily['t2m'],
axis=0).shape)
index
(21, 61)
coldest_long = np.mean(xds_daily['t2m'], axis = 0)[21,61].longitude.values
coldest_lat = np.mean(xds_daily['t2m'], axis = 0)[21,61].latitude.values

print('The coordinates of the coldest place are %.2f, %.2f.' % (coldest_lat,
coldest_long))
The coordinates of the coldest place are 43.99, -109.75.
```

Reverse geocoding the above coordinates shows that the hottest place in this dataset is located in Cody, WY, United States.

# c) Find grid cell and rough location of the highest average cloudiness (i.e. cloudiest place)

```
cloudiest_place = np.mean(xds_daily['tcc'], axis=0).argmax()
index = np.unravel_index(cloudiest_place, np.mean(xds_daily['tcc'],
axis=0).shape)
index
(0, 0)
cloudiest_long = np.mean(xds_daily['tcc'], axis = 0)[0,0].longitude.values
cloudiest_lat = np.mean(xds_daily['tcc'], axis = 0)[0,0].latitude.values

print('The coordinates of the cloudiest place are %.2f, %.2f.' %
(cloudiest_lat, cloudiest_long))
The coordinates of the cloudiest place are 49.24, -125.00.
```

# Reverse geocoding the above coordinates shows that the cloudiest place in this dataset is located in Port Alberni V9Y, British Columbia, Canada

# d) Find grid cell and rough location of the lowest average cloudiest (i.e. least cloudy place)

```
least_cloudy_place = np.mean(xds_daily['tcc'], axis=0).argmin()
index = np.unravel_index(least_cloudy_place, np.mean(xds_daily['tcc'],
axis=0).shape)
index
(71, 41)
least_cloudy_long = np.mean(xds_daily['tcc'], axis = 0)[71,
41].longitude.values
least_cloudy_lat = np.mean(xds_daily['tcc'], axis = 0)[71,
41].latitude.values

print('The coordinates of the cloudiest place are %.2f, %.2f.' %
(least_cloudy_lat, least_cloudy_long))
The coordinates of the cloudiest place are 31.49, -114.75.
```

Reverse geocoding the above coordinates shows that the least cloudy place in this dataset is located off the coast of Baja California, Mexico in the Gulf of California.

# e) Find the grid cell and rough location of the place with the highest range in daily air temperature

```
# Resample the original dataset to reduce the time data dimension
xds_daily_min = np.min(xds_daily['t2m'], axis = 0)
xds_daily_min
xds_daily_max = np.max(xds_daily['t2m'], axis = 0)
xds_daily_max
xds_daily_max
xds_daily_range = xds_daily_max - xds_daily_min
xds_daily_range
max_daily_range = xds_daily_range.argmin()
index = np.unravel index(max daily range, xds daily range.shape)
```

```
index
(46, 7)
max_range_long = xds_daily_range[46, 7].longitude.values
max_range_lat = xds_daily_range[46, 7].latitude.values

print('The coordinates of the place with the highest daily range in temperature are %.2f, %.2f.' % (max_range_lat, max_range_long))
The coordinates of the place with the highest daily range in temperature are 37.74, -123.25.
```

The coordinates of the location with the highest daily range (37.74, -123.25) are located in the Pacific Ocean just off the coast of San Francisco, CA.

#### f) Find the grid cell and rough location of the place with the absolute coldest temperature on a single day

```
# Calculate the absolute min of any day from the minimums data array created
in the previous question

xds_abs_min = xds_daily_min.argmin()
index = np.unravel_index(xds_abs_min, xds_daily_min.shape)
index
(2, 55)
abs_min_long = xds_daily_min[2, 55].longitude.values
abs_min_lat = xds_daily_min[2, 55].latitude.values

print('The coordinates of the place with the coldest absolute temperature on a single day are %.2f, %.2f.' % (abs_min_lat, abs_min_long))
The coordinates of the place with the coldest absolute temperature on a single day are 48.74, -111.25.
```

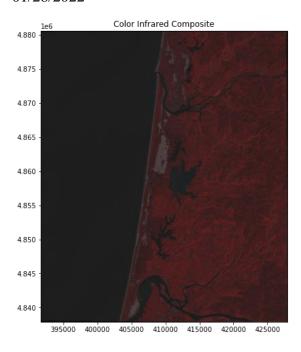
The coordinates of the location with the absolute coldest temperature on a single day (48.74, -111.25) are located near Chester, MT, USA just outside of Mt. Brown.

### **Question 3**

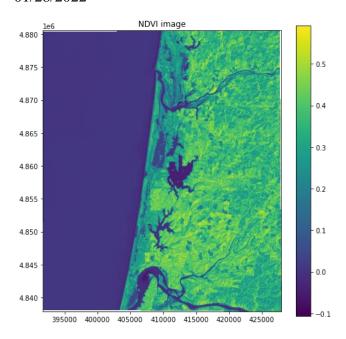
#### a) Creating a color infrared composite of the Landsat image over Florence, OR

```
# Define file path
filepath = '/Users/lily/Documents/GitHub/geospatial-data-science/labs/lab4/'
# Define list of Landsat bands
files = sorted(glob.glob(filepath + 'landsat/*.tif'))
print(files)
# Open a single band
src = rasterio.open(files[0])
band_1 = src.read(1)
# Find metadata (e.g. driver, data type, coordinate reference system,
transform etc.)
print(src.profile)
```

```
# Get corners of dataset
full extent = [src.bounds.left, src.bounds.right, src.bounds.bottom,
src.bounds.top]
print(full extent)
# Plot dataset
fig, ax = plt.subplots(figsize=(8,8))
im = ax.imshow(band 1, cmap='gray', extent=full extent)
ax.set title ("Band \overline{1}")
fig.colorbar(im, orientation='vertical')
plt.show()
# Open all bands in a loop
list bands = []
for file in files:
    # Read band
    src = rasterio.open(file)
   band = src.read(1)
    # Append to list
    list bands.append(band)
# Convert from list of arrays to n-dimensional array
all bands = np.dstack(list bands)
all bands.shape
# Convert 16-bit values to an 8-bit range of 0-255
all bands image = np.uint8((all bands / 65536) * 255)
# Produce a new array by stacking the RGB bands
color ir comp =
np.dstack((all bands image[:,:,4],all bands image[:,:,2],all bands image[:,:,
21))
# Write an array as a raster band to a new 8-bit file. For the new file's
profile,
# we start with the profile of the source
profile = src.profile
# And then change the band count to 3, set the dtype to uint8, and specify
LZW compression.
profile.update(dtype=rasterio.uint8, count=3, compress='lzw')
with rasterio.open(filepath + 'landsat/color IR.tif', 'w', **profile) as dst:
    # Write array
    dst.write(np.rollaxis(rgb, axis=2))
# Plot as Color Infrared Composite image
fig, ax = plt.subplots(figsize=(8,8))
im = ax.imshow(color ir comp, extent=full extent)
ax.set title("Color Infrared Composite")
plt.show()
```



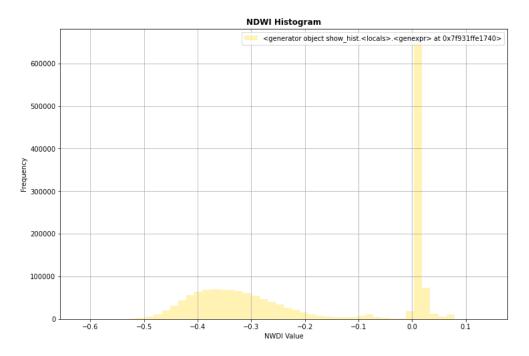
#### b) Creating an NDVI image



### **Question 4**

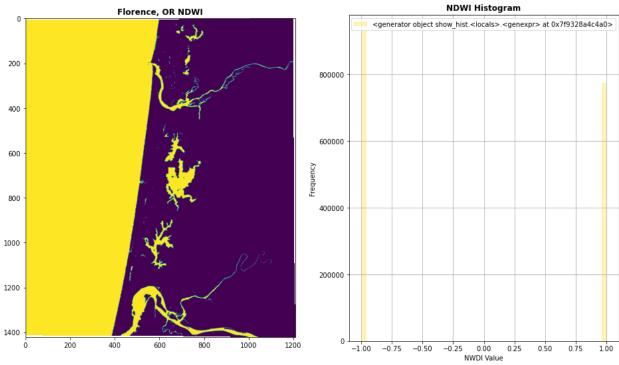
#### a) Produce an NDWI histogram for the Landsat image of Florence

```
# Compute NDWI
np.seterr(divide='ignore', invalid='ignore')
ndwi = np.divide((all bands[:,:,2].astype(float) -
all bands[:,:,4].astype(float)), \
                 (all_bands[:,:,2].astype(float) +
all bands[:,:,4].astype(float)))
ndwi
array([[
               nan,
                            nan,
                                          nan, ...,
                                                            nan,
               nan,
                           nan],
                           nan,
                                          nan, ...,
                                                            nan,
       [
               nan,
               nan,
                           nan],
       [
               nan,
                           nan,
                                          nan, ...,
                                                            nan,
                nan,
                            nan],
       . . . ,
                nan,
                             nan,
                                          nan, ..., -0.29082536,
       -0.31481248, -0.30104966],
                                          nan, ..., -0.32622883,
           nan,
       -0.30386218, -0.27118251],
                                          nan, ..., -0.34158455,
              nan,
                            nan,
       -0.27173558, -0.30858688]])
# Plot NDVI Histogram
fig, axes = plt.subplots(figsize=(12,8))
show hist (
    ndwi, bins=50, lw=0.0, stacked=False, alpha=0.3,
    histtype='stepfilled', ax = axes, title="NDWI Histogram")
axes.set xlabel("NWDI Value")
Text(0.5, 0, 'NWDI Value')
```



#### b) Choose a threshold and produce a binary water mask

```
# Plot NDVI image
fig, ax = plt.subplots(figsize=(8,8))
im = ax.imshow(ndwi, extent=full extent)
ax.set title("NDWI image")
fig.colorbar(im, orientation='vertical')
plt.show()
                  NDWI image
4.880
                                            0.1
                                            0.0
4.870
                                            -0.1
4.865
                                            -0.2
4.855
4.850
4.845
                                             -0.5
4.840
     395000 400000 405000 410000 415000 420000 425000
# Note: The water in this image shows up in the 0.05-0.1 range
# Create a binary mask
ndwi_mask = np.copy(ndwi)
not_water = ndwi_mask[ndwi_mask < -0.05] = -1</pre>
```



#### c) Compute the area of water in the image (including ocean)

```
water_px = np.sum(ndwi_mask == 1)
water_px
774639
# Note: Landsat spatial resolution is 30x30 m
# Calculate area of water px
water_area = water_px * 30
# Convert to square km
water_area_km = water_area / 1000000
water_area_km
23.23917
```

In this Landsat image of Florence, OR, there is 23.23917  $km^2$  of water (including the ocean).