

# Assignment No. 2 - Selma Christensen

180021112

---

[https://github.com/selmachristensen/PY4SA\\_Assignment](https://github.com/selmachristensen/PY4SA_Assignment)

## Python Basics

---

**Task 1** Create an If...Else statement that will test whether a number is divisible by three. "YOUR VALUE is divisible by 3" should be printed if the value is divisible by three. "YOUR VALUE is not divisible by three" should be printed if it is not divisible by three. Test the statement on a numeric variable. Upper case text in the print statement should be replaced with the tested number.

```
In [758... x = 9

if x%3 == (0):
    print (str(x) + " is dividble by 3")

else:
    print (str(x) + " is not divisible by 3")
```

9 is dividble by 3

---

**Task 2** Create an If...Else statement that will test whether a type of fruit, represented as a text string, is in a list of acceptable fruits (apple, orange, pear, kiwi, or strawberry). If the fruit is on the list, the following should be printed: "YOUR FRUIT is acceptable." If not, then the following should be printed: "YOUR FRUIT is not acceptable." Upper case text in the print statement should be replaced with the tested fruit.

```
In [184]: fruits = ["apple", "orange", "pear", "kiwi", "strawberry"]
x = "strawberry"

#using an if in loop I identified the variables on the list
if x in fruits:
    print(str(x) + " is acceptable.")
else:
    print(str(x) + " is not acceptable.")
```

strawberry is acceptable.

---

**Task 3** Create a function to calculate the distance between two coordinates using the haversine formula. Write the following formula where the input parameters are a pair of coordinates as two lists.

```
In [28]: #I started out by importing the appropriate functions needed to solve the equation
from math import radians, sin, cos, sqrt, atan2

#then I created a function in which i defined the distance using the harvestine formula as being
#a result of arguments coordinateA and B
def distance_harvestine(coordinateA, coordinateB):

#as part of the harvestine formula we needed a value for the (earths) radius
    Earth_radius_in_km = 6371

#I used the map function to convert the latitude and longitude of the coordinates into
#radians so I could use them in the formula
    latitude1, longitude1 = map(radians, coordinateA)
    latitude2, longitude2 = map(radians, coordinateB)

#then I made an equation for the distance between the two latitude points and the two longitude points
    distance_latitude = latitude2 - latitude1
    distance_longitude = longitude2 - longitude1

#using the previously imported functions, the converted radians, and the distance equation I plugged them
#into the provided formula
    a = sin(distance_latitude/2)**2 + cos(latitude1) * cos(latitude2) * sin(distance_longitude/2)**2
    c = 2 * atan2(sqrt(a), sqrt(1-a))

#finally I plugged all of the values into the final formula (d=R*c) and returned the distance to
```

```
#indicate that this was the full equation
distance = Earth_radius_in_km * c
return distance
```

```
In [29]: #now that the formula had been defined I choose to coordinates to calculate the distance in between
coordinateA = (55.9533, -3.1883) # Edinburgh
coordinateB = (56.3398, -2.7967) # St. Andrews

#writing out the equation i was able to print the results
distance = distance_harvestine(coordinateA, coordinateB)
print(f"The distance between Edinburgh and St Andrews is {distance:.2f} km.")
```

The distance between Edinburgh and St Andrews is 49.35 km.

---

## Pandas and NymPy

The portland\_park\_trees.csv file contains information about individual trees in city parks in Portland, Oregon. These data were obtained from the City of Portland Office of Parks & Recreation (<https://www.portlandoregon.gov/parks/article/433143>).

---

```
In [30]: import pandas as pd
data = pd.read_csv('portland_park_trees.csv')
data.head()
```

Out[30]:

|   | fid | OBJECTID | Inventory_ | Species | DBH  | Condition | TreeHeight | CrownWidth | CrownWid_1 | CollectedB | ... | Genus       | Com |
|---|-----|----------|------------|---------|------|-----------|------------|------------|------------|------------|-----|-------------|-----|
| 0 | 1   | 426      | 2017/05/09 | PSME    | 37.4 | Fair      | 105.0      | 44.0       | 57.0       | staff      | ... | Pseudotsuga | I   |
| 1 | 2   | 427      | 2017/05/09 | PSME    | 32.5 | Fair      | 94.0       | 49.0       | 45.0       | staff      | ... | Pseudotsuga | I   |
| 2 | 3   | 428      | 2017/05/09 | CRLA    | 9.7  | Fair      | 23.0       | 28.0       | 27.0       | staff      | ... | Crataegus   |     |
| 3 | 4   | 429      | 2017/05/09 | QURU    | 10.3 | Poor      | 28.0       | 38.0       | 31.0       | staff      | ... | Quercus     | no  |
| 4 | 5   | 430      | 2017/05/09 | PSME    | 33.2 | Fair      | 102.0      | 43.0       | 44.0       | staff      | ... | Pseudotsuga | I   |

5 rows × 40 columns

**Question 1** How many trees are of the Quercus or Acer genus?

```
In [31]: data = pd.read_csv('portland_park_trees.csv')

#inorder to get a better overview I selected the specific Genus column
data.Genus
#creating a Boolean mask using the isin function I calculated the length(number of)
#Quercus and Acer species within the Genus category
amount_of_quercus_or_acer = len(data[data['Genus'].isin(['Quercus', 'Acer'])])

print("There are " + str(amount_of_quercus_or_acer) + " trees of the Quercus or Acer genus")
```

There are 5675 trees of the Quercus or Acer genus

---

**Question 2** How many trees are of the Quercus or Acer genus and have a DBH larger than 50 inches?

```
In [32]: data = pd.read_csv('portland_park_trees.csv')

#first I created a function for the Quercus genus with a DBH of more than 50 inches
Quercus_DBH_large_than_50_inches = data[(data["Genus"] == "Quercus") & (data["DBH"]>50)][["Genus", "DBH"]]

#secondly I created a function for the Acer genus with a DBH of more than 50 inches
Acer_DBH_large_than_50_inches = data[(data["Genus"] == "Acer") & (data["DBH"]>50)][["Genus", "DBH"]]

#thirdly I calculated the length of the previous function
#(aka the amount of acer genus with a DBH larger than 50 inches)
amount_of_Acer_with_DBH_large_than_50_inches = len(Acer_DBH_large_than_50_inches)

#fourthly I calculated the length of the other previous function
#(aka the amount of Quercus genus with a DBH larger than 50 inches)
amount_of_Quercus_with_DBH_large_than_50_inches = len(Quercus_DBH_large_than_50_inches)

#finally I printed the result
print("There are " + str(amount_of_Acer_with_DBH_large_than_50_inches +
                        amount_of_Quercus_with_DBH_large_than_50_inches) +
      " trees of the Quercus or Acer genus that have a DBH larger than 50 inches")
```

There are 124 trees of the Quercus or Acer genus that have a DBH larger than 50 inches

---

**Question 3** Which genus has the highest mean DBH of the following genera: Quercus, Acer, or Fraxinus?

```
In [33]: data = pd.read_csv('portland_park_trees.csv')

#first I identifid the data sample, showing that I only worked with the 3 specific
#genras within the Genus column

Acer = data[data["Genus"] == "Acer"]

Quercus = data[data["Genus"] == "Quercus"]

Fraxinus = data[data["Genus"] == "Fraxinus"]

#next I calculated the mean DBH within the different genras

DBH_Acer_mean = Acer["DBH"].mean()

DBH_Quercus_mean = Quercus["DBH"].mean()

DBH_Fraxinus_mean = Fraxinus["DBH"].mean()

#finally I printed them out inorder to compare their values

print(DBH_Acer_mean)
print(DBH_Quercus_mean)
print(DBH_Fraxinus_mean)

18.419085331846066
23.56823839157492
11.033609693877551
```

```
In [34]: #having compared their values it was clear to see that Quercus had the highest mean

print("Quercus has the highest mean DBH")

Quercus has the highest mean DBH
```

---

**Question 4** How many different species of trees are recorded in the Acer genus?

```
In [35]: data = pd.read_csv('portland_park_trees.csv')
#the following line identifies the rows in which the Genus coloumn is equal to acer
#creating a coresponding name
rec_acer= data.loc[data["Genus"] == "Acer"]
#using the unique funtction I found the amount of different values within the species
#categories which genus column was equal to acer (hence the use of the previous function lable)
amount_different_acer_species = len(rec_acer["Species"].unique())

print ("There are " + str(amount_different_acer_species) +
      " different species of trees recorded in Acer genus")
```

There are 20 different species of trees recorded in Acer genus

---

**Using new data set:** The world\_cities.csv is a file that contains cities, countries, population, coordinates (geographic) and a Boolean attribute that defines if the city is the capital city or not. Read this file as a Pandas dataframe and create the required scripts

```
In [36]: #first I made sure to import the appropriate libraries

import numpy as np
import pandas as pd

#then I read the data file, taking a look at the first 5 collums to make sure that it was correct

data = pd.read_csv('world_cities.csv', header=0)
data.head()
```

```
Out[36]:
```

|   | city               | country   | pop   | lat   | lon   | capital |
|---|--------------------|-----------|-------|-------|-------|---------|
| 0 | 'Abasan al-Jadidah | Palestine | 5629  | 31.31 | 34.34 | 0       |
| 1 | 'Abasan al-Kabirah | Palestine | 18999 | 31.32 | 34.35 | 0       |
| 2 | 'Abdul Hakim       | Pakistan  | 47788 | 30.55 | 72.11 | 0       |
| 3 | 'Abdullah-as-Salam | Kuwait    | 21817 | 29.36 | 47.98 | 0       |
| 4 | 'Abud              | Palestine | 2456  | 32.03 | 35.07 | 0       |

---

## Question 5

Calculate a new column named "pop\_M" (population in millions), by transforming the "pop" (population) column

```
In [37]: data = pd.read_csv('world_cities.csv', header=0)

#I started out by creating a new column in which the old pop column was being divided by a million to
#represent population in millions

data['pop_M'] = data['pop']/1000000

#Then I printed out the first 10 rows of the table to make sure it was correct
print(data.head(10))
```

|   | city               | country      | pop   | lat   | lon   | capital | pop_M    |
|---|--------------------|--------------|-------|-------|-------|---------|----------|
| 0 | 'Abasan al-Jadidah | Palestine    | 5629  | 31.31 | 34.34 | 0       | 0.005629 |
| 1 | 'Abasan al-Kabirah | Palestine    | 18999 | 31.32 | 34.35 | 0       | 0.018999 |
| 2 | 'Abdul Hakim       | Pakistan     | 47788 | 30.55 | 72.11 | 0       | 0.047788 |
| 3 | 'Abdullah-as-Salam | Kuwait       | 21817 | 29.36 | 47.98 | 0       | 0.021817 |
| 4 | 'Abud              | Palestine    | 2456  | 32.03 | 35.07 | 0       | 0.002456 |
| 5 | 'Abwein            | Palestine    | 3434  | 32.03 | 35.20 | 0       | 0.003434 |
| 6 | 'Adadlay           | Somalia      | 9198  | 9.77  | 44.65 | 0       | 0.009198 |
| 7 | 'Adale             | Somalia      | 5492  | 2.75  | 46.30 | 0       | 0.005492 |
| 8 | 'Afak              | Iraq         | 22706 | 32.07 | 45.26 | 0       | 0.022706 |
| 9 | 'Afif              | Saudi Arabia | 41731 | 23.92 | 42.93 | 0       | 0.041731 |

Remove the original "pop" column

```
In [38]: data = pd.read_csv('world_cities.csv', header=0)
data['pop_M'] = data['pop']/1000000

#next I deleted/dropped the "pop" column, by using the .drop function, identifying the column,
#and stating that inplace=true to make it a permanent change
data.drop(columns=['pop'], inplace=True)

#again I printed out the first 10 rows to ensure it was correctly executed
print(data.head(10))
```



|   | city               | country      | lat   | lon   | capital | pop_M    |
|---|--------------------|--------------|-------|-------|---------|----------|
| 0 | 'Abasan al-Jadidah | Palestine    | 31.31 | 34.34 | 0       | 0.005629 |
| 1 | 'Abasan al-Kabirah | Palestine    | 31.32 | 34.35 | 0       | 0.018999 |
| 2 | 'Abdul Hakim       | Pakistan     | 30.55 | 72.11 | 0       | 0.047788 |
| 3 | 'Abdullah-as-Salam | Kuwait       | 29.36 | 47.98 | 0       | 0.021817 |
| 4 | 'Abud              | Palestine    | 32.03 | 35.07 | 0       | 0.002456 |
| 5 | 'Abwein            | Palestine    | 32.03 | 35.20 | 0       | 0.003434 |
| 6 | 'Adadlay           | Somalia      | 9.77  | 44.65 | 0       | 0.009198 |
| 7 | 'Adale             | Somalia      | 2.75  | 46.30 | 0       | 0.005492 |
| 8 | 'Afak              | Iraq         | 32.07 | 45.26 | 0       | 0.022706 |
| 9 | 'Afif              | Saudi Arabia | 23.92 | 42.93 | 0       | 0.041731 |

**Choose/subset a city that starts with the same letter as your first name (for example, "Mexico City" if your first name is Michael)**

```
In [39]: # As my name starts with an S, I chose the city Sololo in Gambia
# Below is the query I ran, to show the different characteristics of Sololo "city",
#"country", "lat", "lon", "capital", "pop_M"
sololo_query= data.query('city=="Sololo"')[["city", "country", "lat", "lon", "capital", "pop_M"]]
print(sololo_query)
```

|       | city   | country | lat   | lon    | capital | pop_M    |
|-------|--------|---------|-------|--------|---------|----------|
| 35731 | Sololo | Gambia  | 13.45 | -14.68 | 0       | 0.000799 |

**Subset the five biggest (i.e., largest population sizes) cities from the country where your selected city is**

```
In [40]: # I made a subset query in which I first identified the data I wanted to work with (Gambia)
# Then I sorted the values within the query according to their population size (pop_m)
# from highest to lowest (ascending=False)
# Finally I added the three categories I wanted to be shown, country, pop_M, and city
subset_query = data.query('country=="Gambia").sort_values(["pop_M"], ascending=False)
[["city", "country", "pop_M"]]
print(subset_query.head(5))
```

|       | city        | country | lat   | lon    | capital | pop_M    |
|-------|-------------|---------|-------|--------|---------|----------|
| 34573 | Serre Kunda | Gambia  | 13.45 | -16.68 | 0       | 0.335733 |
| 5382  | Brikama     | Gambia  | 13.28 | -16.66 | 0       | 0.080726 |
| 2925  | Bakau       | Gambia  | 13.49 | -16.69 | 0       | 0.045529 |
| 3264  | Banjul      | Gambia  | 13.46 | -16.60 | 1       | 0.034388 |
| 11205 | Farafenni   | Gambia  | 13.57 | -15.61 | 0       | 0.030418 |

## Python Data Visualization

Using the same dataset portland\_park\_trees.csv, create using seaborn, pandas or matplotlib libraries the following charts:

---

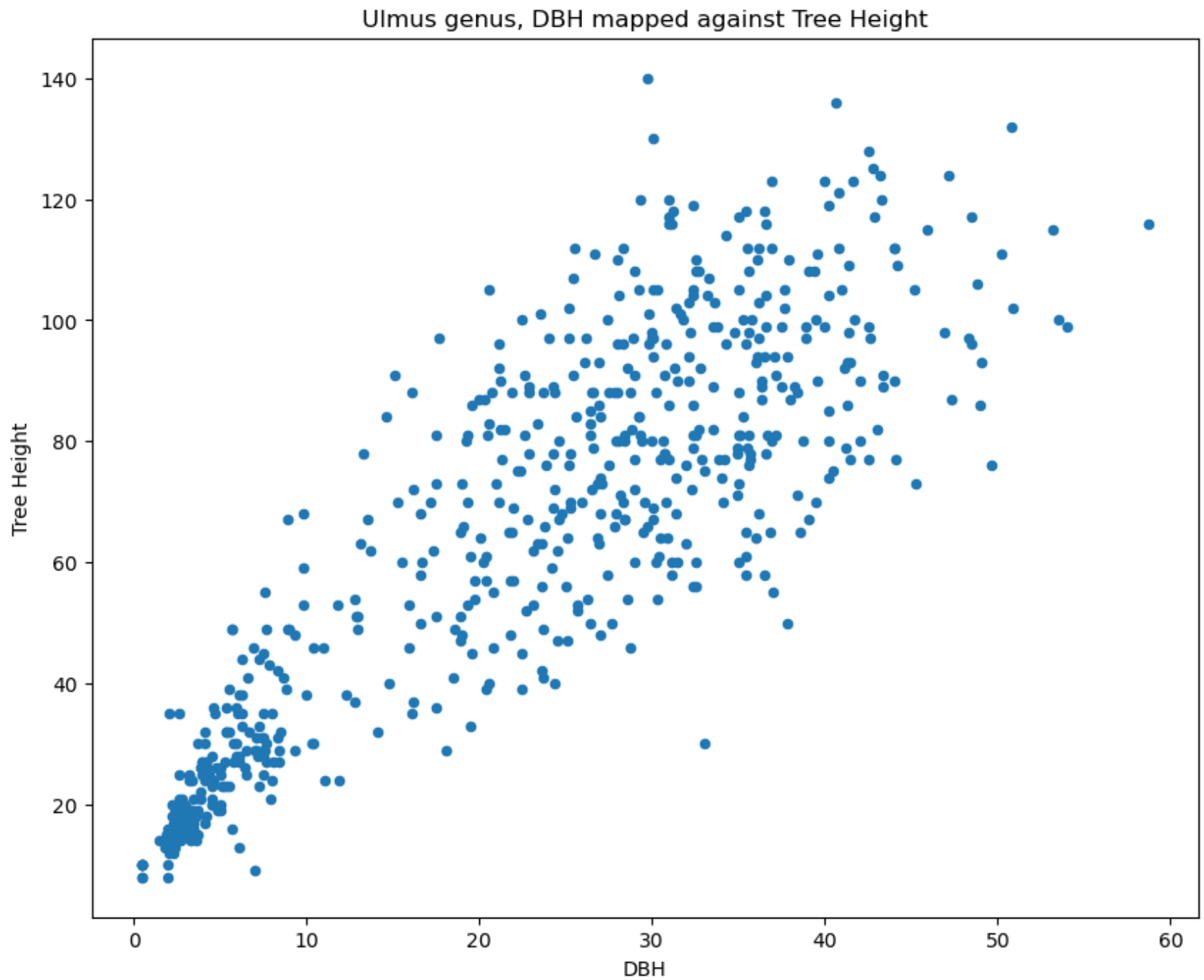
```
In [41]: import numpy as np
import pandas as pd
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
import matplotlib.pyplot as plt
plt.rcParams['figure.figsize'] = [10, 8]
%matplotlib inline
```

---

**Graph 1** Create a scatterplot for just trees in the Ulmus genus with DBH mapped to the x-axis and tree height mapped to the y-axis (Hint: You will need to use the "Genus", "DBH", and "TreeHeight" attributes.).

```
In [42]: data= pd.read_csv('portland_park_trees.csv')
Ulmus_sample = data[(data["Genus"]=="Ulmus")]
Ulmus_sample.plot.scatter(x = "DBH", y = "TreeHeight")
plt.title ("Ulmus genus, DBH mapped against Tree Height") #title of the graph
plt.xlabel("DBH") # x axis label
plt.ylabel("Tree Height") #y axis label
```

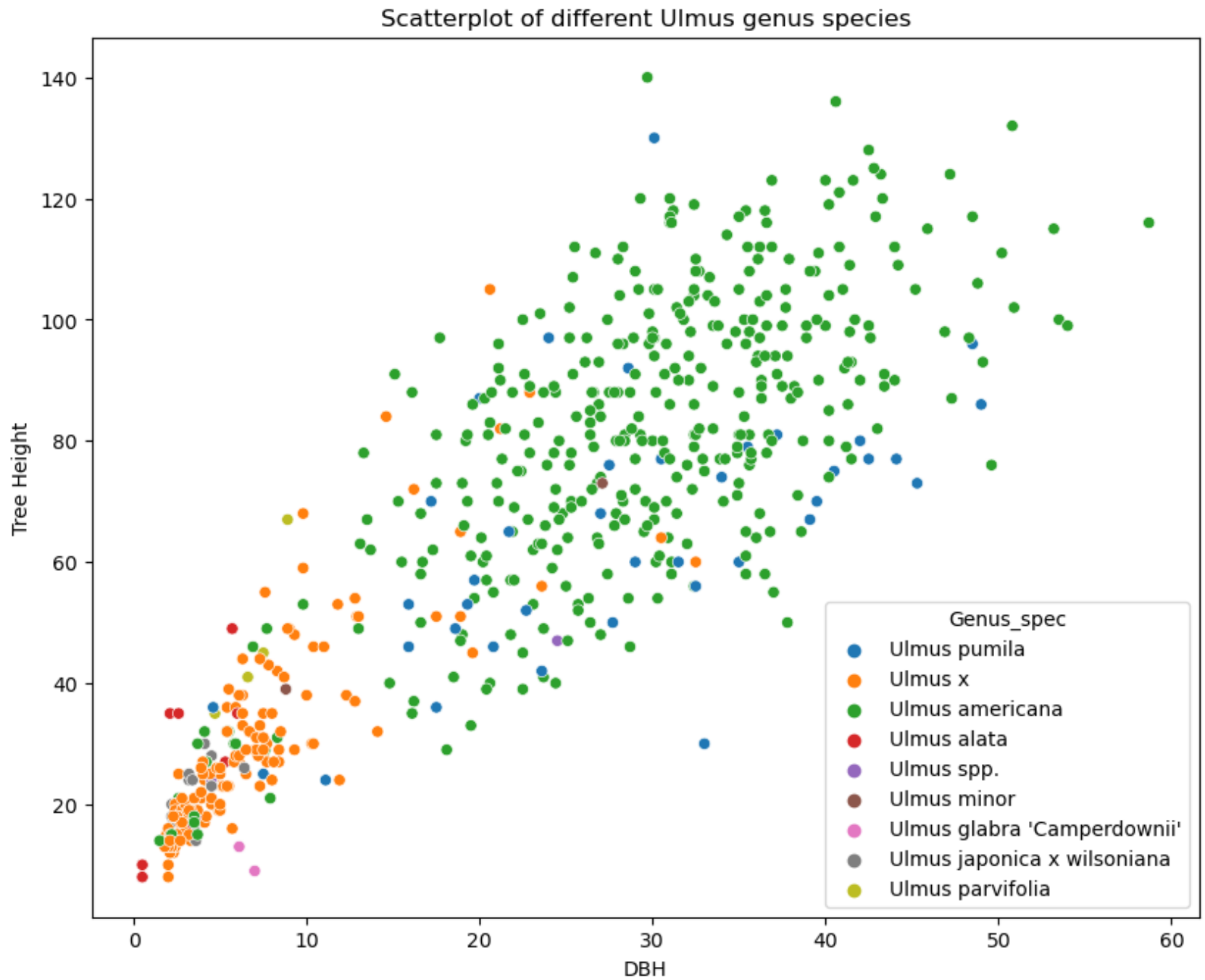
```
Out[42]: Text(0, 0.5, 'Tree Height')
```



**Graph 2** Create a scatterplot for just trees in the Ulmus genus with DBH mapped to the x-axis, tree height mapped to the y-axis, and tree species mapped to hue (Hint: You will need to use the "Genus", "Genus\_spec", "DBH", and "TreeHeight" attributes.)

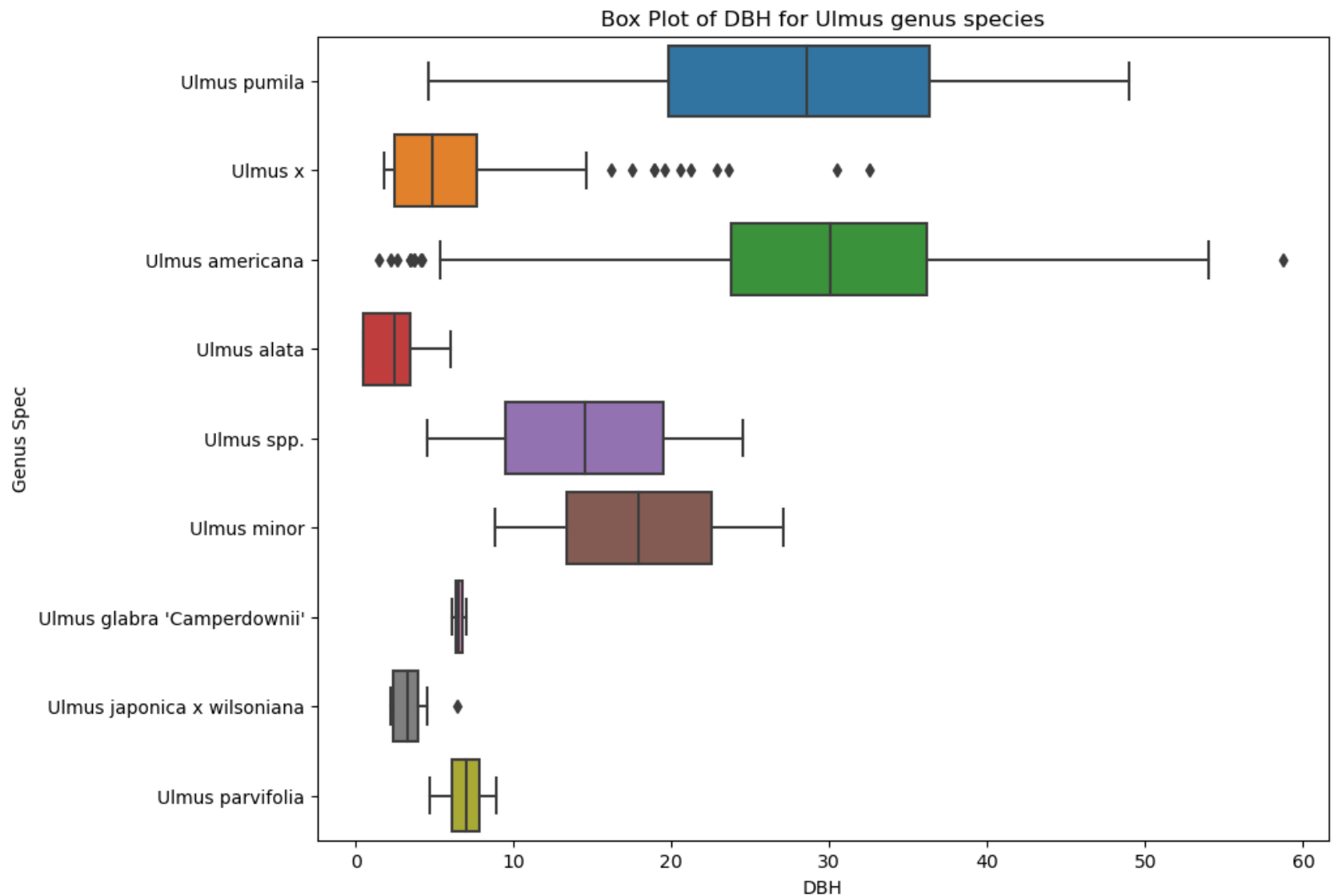
```
In [43]: Ulmus_sample = data[(data["Genus"]=="Ulmus")]#defining what data to use
sns.scatterplot(data=Ulmus_sample, x="DBH", y="TreeHeight", hue="Genus_spec")
#the scatterplot mapped as a Seaborn
plt.title ("Scatterplot of different Ulmus genus species") #title of the graph
plt.xlabel("DBH") # x axis label
plt.ylabel("Tree Height") #y axis label
```

```
Out[43]: Text(0, 0.5, 'Tree Height')
```



**Graph 3** Create a boxplot of DBH for just the Ulmus genus differentiated by species (or, each species should have its own boxplot).

```
In [44]: data= pd.read_csv('portland_park_trees.csv')
Ulmus_sample = data[(data["Genus"]=="Ulmus")] #defining the specific data
sns.boxplot(data=Ulmus_sample, x="DBH", y="Genus_spec")#boxplot mapped as a Seaborn
plt.xlabel("DBH") # x axis label
plt.ylabel("Genus Spec") #y axis label
plt.title("Box Plot of DBH for Ulmus genus species")#title of the graph
plt.show()
```

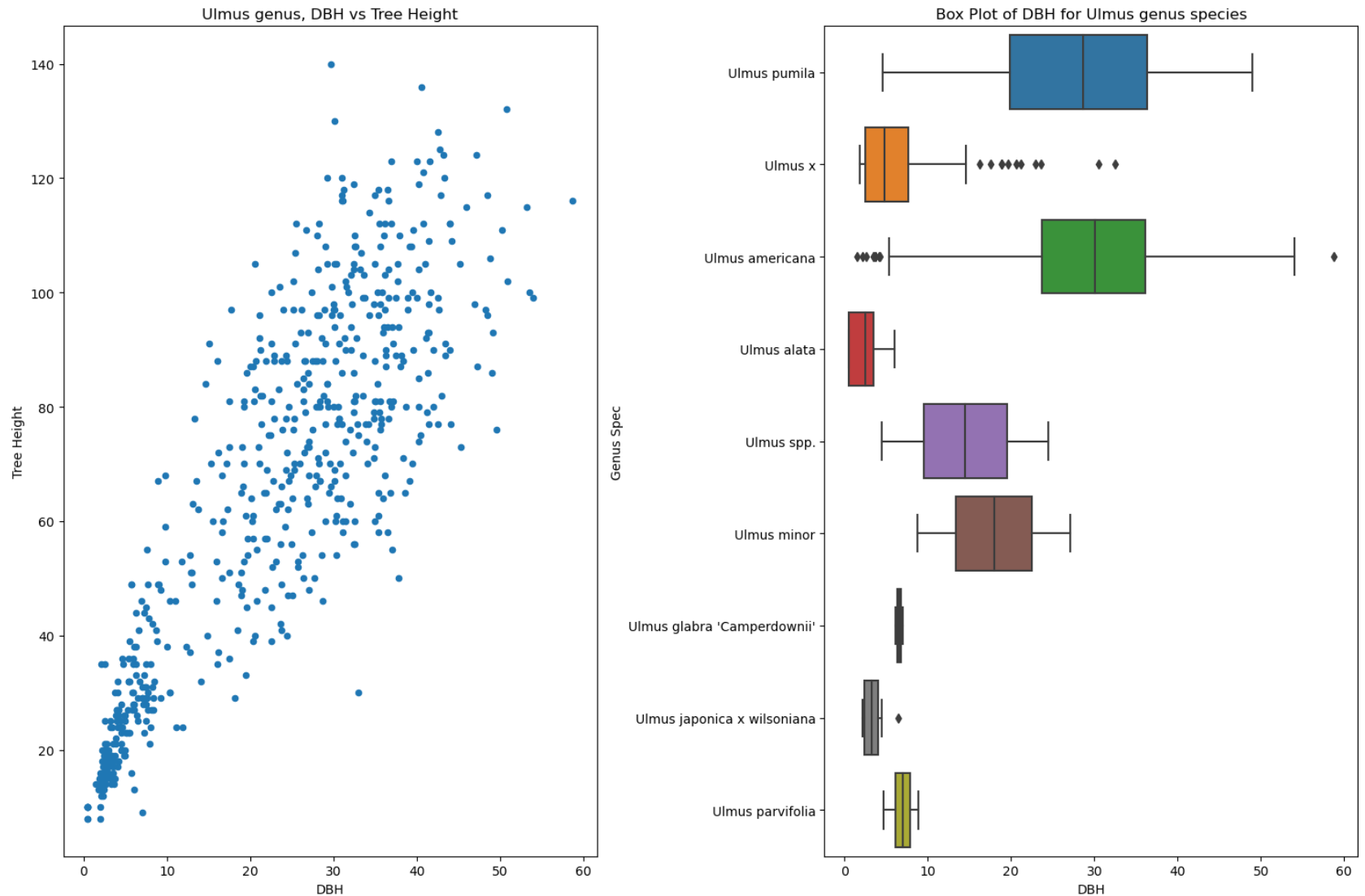


**Graph 4** Combine Graphs 1 and 3 into a single figure. Do not plot a legend for any of the graphs.

```
In [45]: data = pd.read_csv('portland_park_trees.csv')
Ulmus_sample = data[(data["Genus"]=="Ulmus")]
```

```
fig1,(ax1, ax2) = plt.subplots(ncols=2, figsize=(15, 10))  
#creating a subplot inorder to include both graphs  
  
Ulmus_sample.plot.scatter(x = "DBH", y = "TreeHeight", ax=ax1) #showing that scatterplot is on axis 1  
ax1.set_title("Ulmus genus, DBH vs Tree Height") #titel label  
ax1.set_xlabel("DBH") #x-axis label  
ax1.set_ylabel("Tree Height") #y-axis label  
  
sns.boxplot(data=Ulmus_sample, x= "DBH", y = "Genus_spec", ax=ax2) #showing that boxplot is on axis 2  
ax2.set_title("Box Plot of DBH for Ulmus genus species") #titel label  
ax2.set_xlabel("DBH") #x-axis label  
ax2.set_ylabel("Genus Spec") #y-axis label  
  
plt.tight_layout() #making sure the graphs do not overlap  
plt.show()
```





## Python GeoPandas

Go to the Spatial Data Portal of Scotland and find any spatial data that you find interesting in a shapefile format. Download this data and Produce code to complete the requested tasks.

---

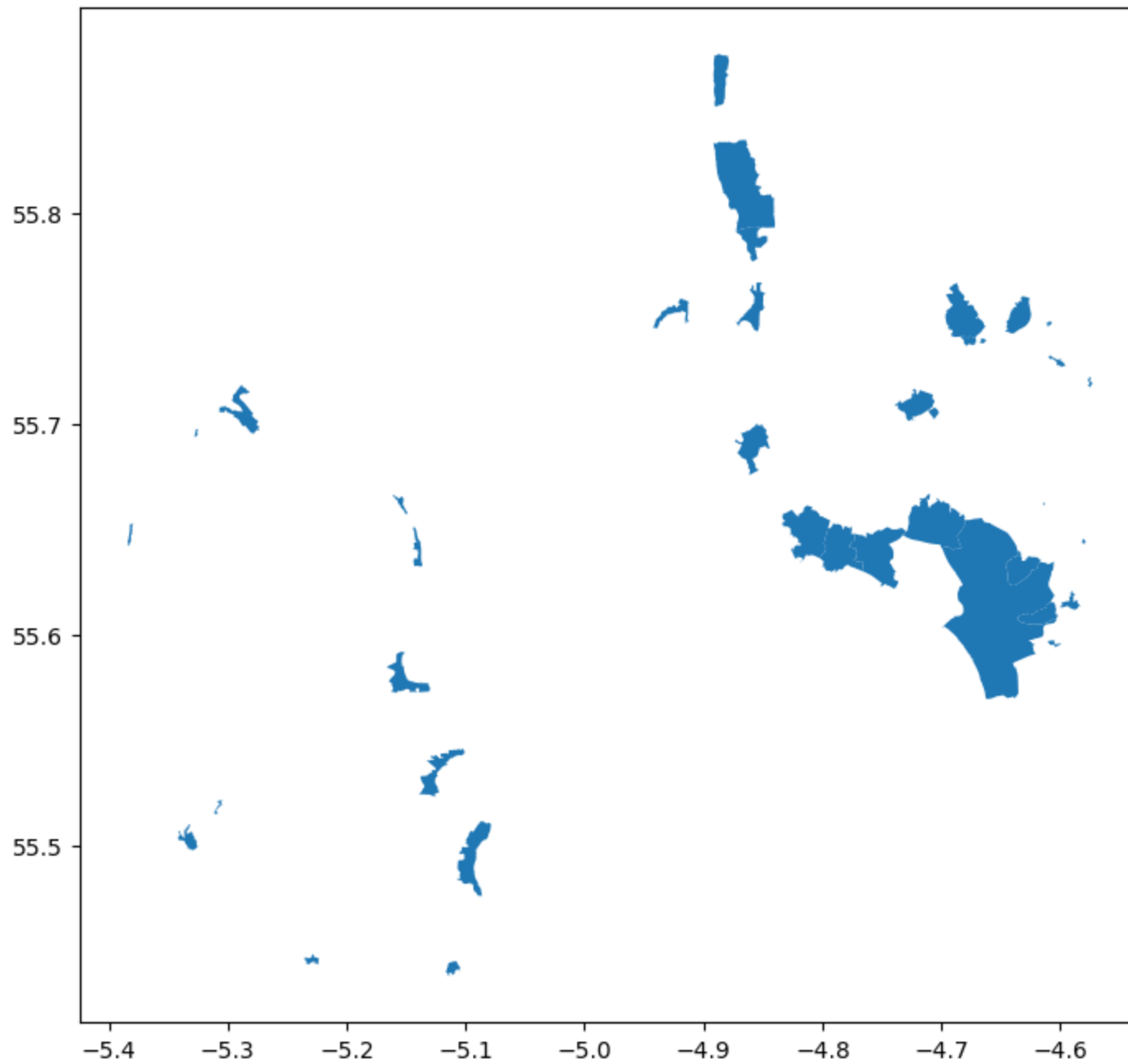
**Task 1** Read the selected dataset as GeoPandas DataFrame

```
In [46]: import geopandas as gpd
import numpy as np
import pandas as pd

APA = gpd.read_file('Alcohol_Prohibition_Areas.shp') #reading file through GeoPanda gpd

APA.plot()
```

```
Out[46]: <AxesSubplot: >
```



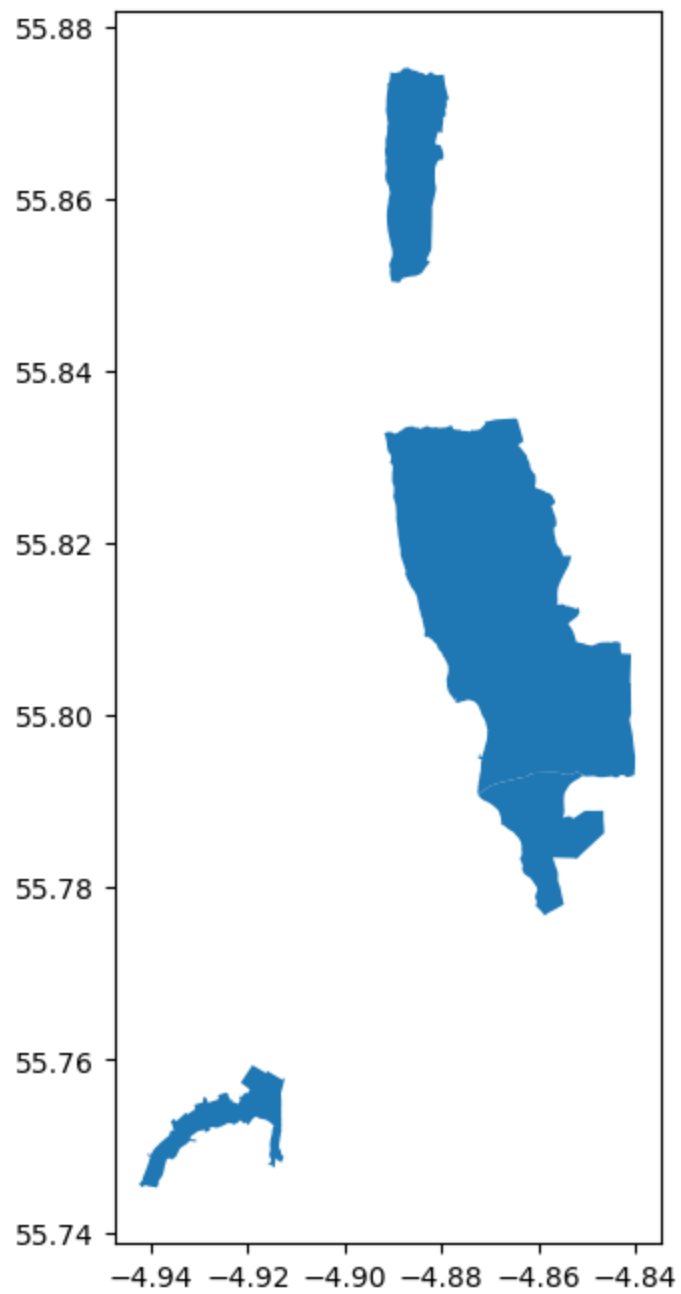
**Task 2** Use the correct code to plot the first 5 and the last 5 sets of records in your selected dataset.

```
In [47]: APA.head(5)
```

```
Out[47]:
```

|   | OBJECTID | NAME          | geometry  |
|---|----------|---------------|---|
| 0 | 1        | Skelmorlie    | POLYGON ((-4.88868 55.87468, -4.88868 55.87468... |
| 1 | 2        | Largs A       | POLYGON ((-4.87840 55.83358, -4.87839 55.83358... |
| 2 | 3        | Largs B       | POLYGON ((-4.84823 55.78890, -4.84817 55.78890... |
| 3 | 4        | Millport      | POLYGON ((-4.93146 55.75405, -4.93092 55.75427... |
| 4 | 5        | West Kilbride | POLYGON ((-4.85966 55.69754, -4.85935 55.69769... |

```
In [48]: APA = gpd.read_file('Alcohol_Prohibition_Areas.shp')
plotted_APAhead = APA[0:4] #defining what part to plot
plotted_APAhead.plot()
plt.show() #the first five sets of the record
```

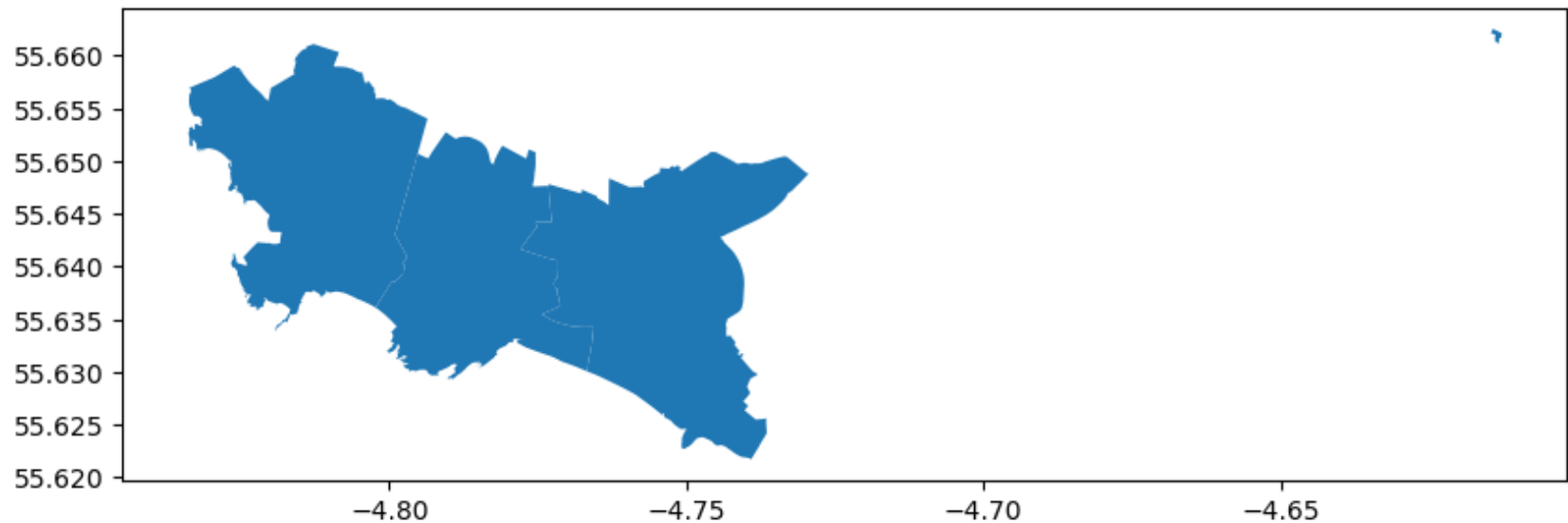


```
In [49]: APA.tail(5)
```

Out [49]:

|    | OBJECTID | NAME       | geometry  |
|----|----------|------------|---|
| 32 | 33       | Torranyard | POLYGON ((-4.61340 55.66227, -4.61339 55.66226... |
| 33 | 34       | Stevenston | POLYGON ((-4.73590 55.65008, -4.73579 55.65010... |
| 34 | 35       | Saltcoats  | MULTIPOLYGON (((-4.77448 55.64761, -4.77447 55... |
| 35 | 36       | Ardrossan  | POLYGON ((-4.79495 55.65117, -4.79496 55.65114... |
| 36 | 37       | Fairlie    | MULTIPOLYGON (((-4.85460 55.74369, -4.85515 55... |

```
In [50]: APA = gpd.read_file('Alcohol_Prohibition_Areas.shp')
         plotted_APAtail = APA[32:36] #defining what part to plot
         plotted_APAtail.plot()
         plt.show() #the last five sets of the record
```



**Task 3** Create a map where you can explore the selected dataset. Try to plot the map using some categorical attribute. Include a Tooltip.

```
In [51]: #I chose the categorical attribute "NAME"
         APA.explore(column="NAME", cmap='RdYlBu')
```

Out [51]: Make this Notebook Trusted to load map: File -> Trust Notebook

Drybridge  
Fairlie  
Gateside  
Girdle Toll  
Glengarnuck  
Irvine  
Kilbirnie  
Kildonan  
Kilwinning  
Lagg  
Lamlash  
Largs A  
Largs B  
Lochranza  
Longbar  
Millport  
Pirnmill  
Saltcoats  
Sannox  
Shiskine  
Skelmorlie  
Springside  
Stevenston  
Torranyard  
West Kilbride  
Whiting Bay

---

**Task 4** What is the Coordinate Reference System of the selected dataset?

In [52]: `APA.crs`

```
Out[52]: <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 ensemble
- Ellipsoid: WGS 84
- Prime Meridian: Greenwich
```

---

**Task 5** How many features does the selected dataset contain?

```
In [53]: print("The data set I selected has " + str(APA.shape[0]) + " features.")
         #using .shape to define the amount of features
```

The data set I selected has 37 features.

---

**Task 6** Define a sub-setting criterion to create a new geopandas dataframe where you filter the selected dataset based on a categorical attribute.

```
In [54]: APACat = APA[["NAME"]] #showing which categorical attribute to include
         APACat.head(10)
```



Out [54]:

|   | NAME          |
|---|---------------|
| 0 | Skelmorlie    |
| 1 | Largs A       |
| 2 | Largs B       |
| 3 | Millport      |
| 4 | West Kilbride |
| 5 | Dalry         |
| 6 | Burnhouse     |
| 7 | Barmill       |
| 8 | Gateside      |
| 9 | Longbar       |

---

**Task 7** Define a sub-setting criterion to create a new geopandas dataframe where you filter the selected dataset based on a numerical attribute.

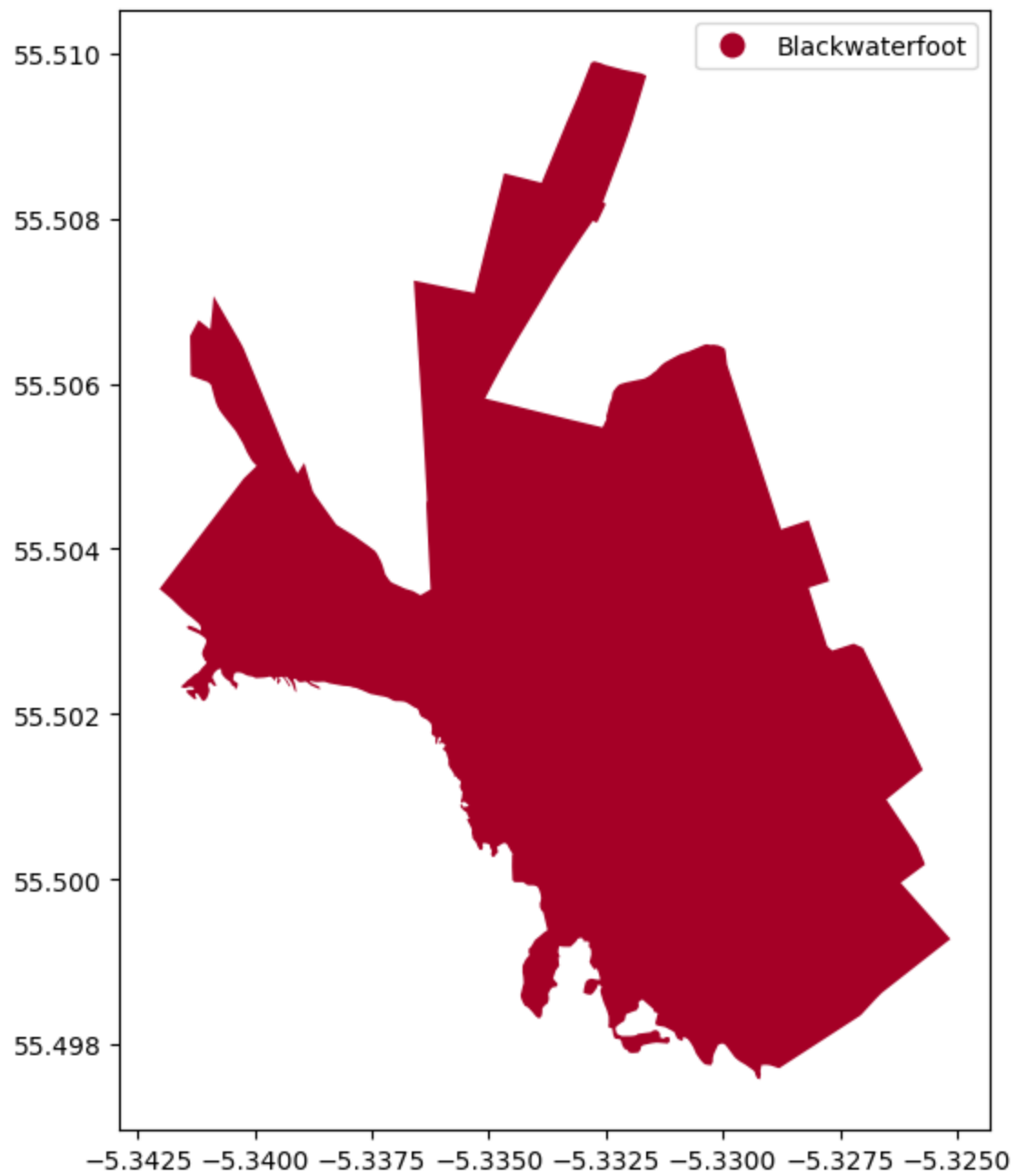
```
In [55]: APAnum = APA["NAME"] == "Blackwaterfoot" #showing which numerical attribute to include
APAnum.tail(10)
```

```
Out [55]: 27    True
28    False
29    False
30    False
31    False
32    False
33    False
34    False
35    False
36    False
Name: NAME, dtype: bool
```

---

**Task 8** Plot the new/filtered geopandas dataframe using one of the attributes to create a choropleth map.

```
In [56]: APAnum = APA[APA["NAME"] == "Blackwaterfoot"]  
          #shows only those with the attribute Blackwaterfoot within the NAME category  
  
          APAnum.plot(column="NAME", cmap='RdYlBu', legend=True) #plots the map, including the legend  
          plt.show()
```



## Python Rasterio

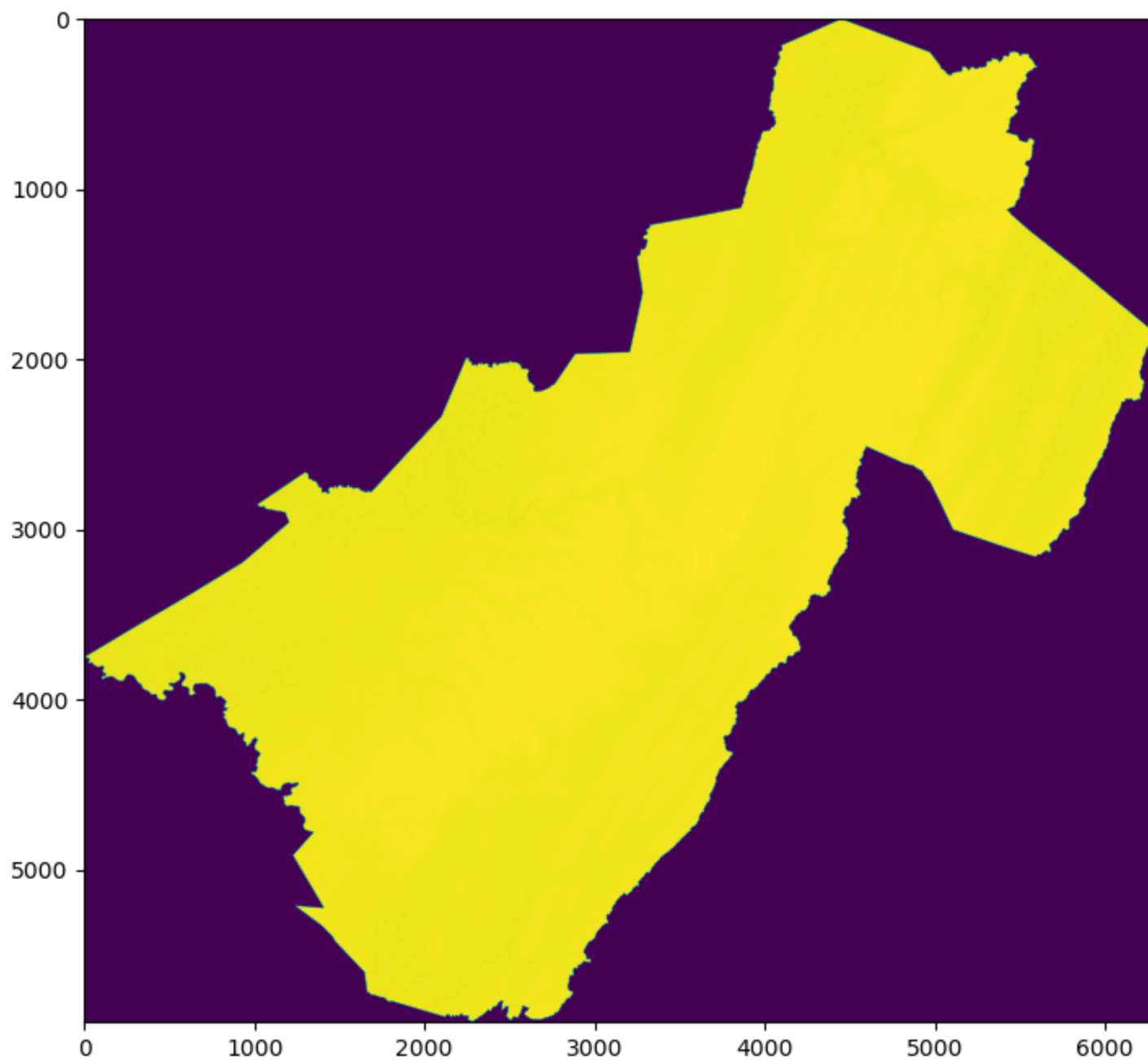
The dataset for this part of the assignment is elev.tif a 30 m spatial resolution digital elevation model (DEM) derived from the National Elevation Dataset (NED) in Canada with elevation in meters.

```
In [59]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
import geopandas as gpd
import contextily as ctx
import rasterio as rio
from rasterio import plot
from rasterio.plot import show
from rasterio.plot import show_hist
```

---

### Task 1 Read the file as a rasterio dataset

```
In [61]: with rio.open('elev.tif') as elev: #imports dataset as elev
    elev = elev.read(1) #reads dataset
    show(elev) #shows dataset
```



Out[61]: <AxesSubplot: >

---

**Task 2** What is the CRS of the dataset?

```
In [62]: with rio.open('elev.tif') as elev: #opens dataset
         crs = elev.crs #defines the CRS
         print(crs) #prints CRS
```

EPSG:32617

---

**Task 3** Describe the raster dataset regarding the raster extent (bounds), the reference system, and how many bands are in this dataset.

```
In [63]: raster_extent = elev.bounds #defines the bounds within dataset
         reference_system = elev.crs #defines the CRS (reference system) within dataset
         amount_of_bands = elev.count #defines the amount of bands within dataset

         print("Results:")
         print(raster_extent)
         print(reference_system)
         print(amount_of_bands)
         print(" ")
         print("Description:")
         print("The raster dataset we are working with is using the " + str(reference_system) +
               " , the Projected coordinate system for the area between 84°W and 78°W, "
               " northern hemisphere between equator and 84°N, " +
               " onshore and offshore. It has " + str(amount_of_bands) +
               " band, meaning that it only has one data layer showing the elevation in meters in Canada. "
               " The raster extent (bounds) of the dataset, is " + str(raster_extent) +
               " showing the area which the dataset encompasses.")
```

**Results:**

BoundingBox(left=479753.39945587853, bottom=4170823.2037591375, right=668843.3994558785, top=4347733.203759138)  
EPSG:32617  
1

**Description:**

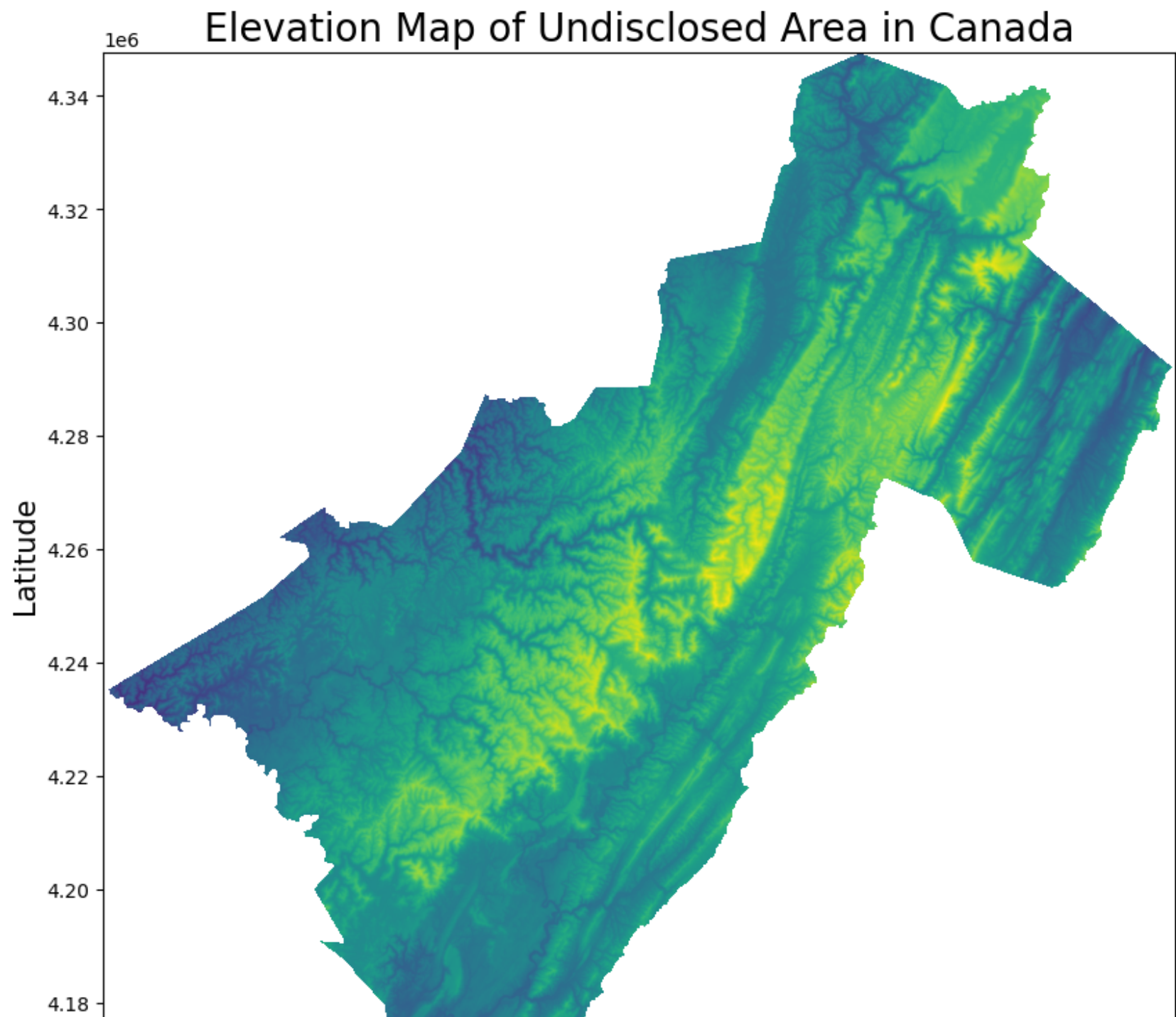
The raster dataset we are working with is using the EPSG:32617 , the Projected coordinate system for the area between 84°W and 78°W, northern hemisphere between equator and 84°N, onshore and offshore. It has 1 band, meaning that it only has one data layer showing the elevation in meters in Canada. The raster extent (bounds) of the dataset, is BoundingBox(left=479753.39945587853, bottom=4170823.2037591375, right=668843.3994558785, top=4347733.203759138) showing the area which the dataset encompasses.

---

**Task 4** Create a plot/map of the raster dataset.

```
In [64]: elev = rio.open('elev.tif') #opens dataset as elev
fig2, ax = plt.subplots(figsize=(10,10)) #defines the figure as a subplot on one ax
show(elev, ax=ax) #shows the data (aka prints it)
elev.close() #closes raster file
ax.set_title('Elevation Map of Undisclosed Area in Canada', fontsize=20) #prints title onto dataset
ax.set_xlabel('Longitude', fontsize=15) #prints x label onto dataset
ax.set_ylabel('Latitude', fontsize=15) #prints y label onto dataset
```

```
Out[64]: Text(0, 0.5, 'Latitude')
```





---

**Task 5** Create Histograms from the raster.

```
In [65]: elev = rio.open('elev.tif') #opens dataset

show_hist(elev, bins=70, lw=0.0, stacked=False, alpha=0.7, histtype='stepfilled',
          title='Histogram of Elevation Raster Data')
# prints histogram while defining the different proximities, data amounts etc.

elev.close()#closes dataset
```

