### Metadata

Course: DS 5100

Module: 04 Functions HW

Title: Fighting Forest Fires with Functions

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 URL of this file in GitHub: https://github.com/rachel-holman/DS5100dnw9qk/blob/main/lessons/M04/hw04.ipynb

### Instructions

In your **private course repo on Rivanna**, write a Jupyter notebook running Python that performs the numbered tasks below.

For each task, create one or more code cells to perform the task.

Save your notebook in the M04 directory as hw04.ipynb.

Add and commit these files to your repo.

Then push your commits to your repo on GitHib.

Be sure to fill out the **Student Info** block above.

To submit your homework, save the notebook as a PDF and upload it to GradeScope, following the instructions.

**TOTAL POINTS: 14** 

### Overview

In this homework, you will work with the Forest Fires Data Set from UCI.

There is a local copy of these data as a CSV file in the HW directory for this module in the course repo.

You will create a group of related functions to process these data.

This notebook will set the table for you by importing and structuring the data first.

# **Setting Up**

First, we read in our local copy of the dataset and save it as a list of lines.

```
In [1]: data_file = open('uci_mldb_forestfires.csv', 'r').readlines()
```

Then, we inspect first ten lines, replacing commas with tabs for readability.

```
In [2]:
        for row in data file[:10]:
             row = row.replace(',', '\t')
             print(row, end='')
        X
                 Y
                         month
                                  day
                                          FFMC
                                                   DMC
                                                           DC
                                                                    ISI
                                                                                    RH
                                                                            temp
        wind
                 rain
                         area
        7
                 5
                                  fri
                                          86.2
                                                   26.2
                                                           94.3
                                                                    5.1
                                                                            8.2
                                                                                     51
                         mar
        6.7
                 0.0
                         0.0
        7
                         oct
                                          90.6
                                                   35.4
                                                           669.1
                                                                   6.7
                                                                            18.0
                                                                                     33
                 4
                                  tue
        0.9
                 0.0
                         0.0
        7
                                          90.6
                                                                                     33
                 4
                         oct
                                  sat
                                                   43.7
                                                           686.9
                                                                    6.7
                                                                            14.6
        1.3
                 0.0
                         0.0
                 6
                         mar
                                  fri
                                          91.7
                                                   33.3
                                                           77.5
                                                                    9.0
                                                                            8.3
                                                                                     97
        4.0
                 0.2
                         0.0
                                          89.3
                                                                                     99
        8
                 6
                         mar
                                  sun
                                                  51.3
                                                           102.2
                                                                   9.6
                                                                            11.4
        1.8
                 0.0
                         0.0
        8
                 6
                                          92.3
                                                   85.3
                                                           488.0
                                                                   14.7
                                                                            22.2
                                                                                    29
                         aug
                                  sun
                 0.0
        5.4
                         0.0
        8
                 6
                         aug
                                 mon
                                          92.3
                                                   88.9
                                                           495.6
                                                                   8.5
                                                                            24.1
                                                                                     27
        3.1
                 0.0
                         0.0
                                          91.5
                                                  145.4
                                                           608.2
                                                                   10.7
                                                                            8.0
                                                                                     86
                 6
                         aug
                                 mon
        2.2
                 0.0
                         0.0
                                          91.0
                                                   129.5
                                                           692.6
                                                                   7.0
                                                                            13.1
                                                                                     63
        8
                 6
                         sep
                                  tue
        5.4
                 0.0
                         0.0
```

#### Convert CSV into Datafame-like Data Structure

We use a helper function to convert the data into the form of a dataframe-like dictionary.

That is, we convert a list of rows into a dictionary of columns, each cast to the appropriate data type.

Later, we will use Pandas and R dataframes to do this work.

First, we define the data types by inspecting the data and creating a dictionary of lambda functions to do our casting.

```
's': lambda x: str(x),
'f': lambda x: float(x)
}
```

Next, we grab the column names from the first row or list.

Note that <code>.strip()</code> is a string function that removes extra whitespace from before and after a string.

```
In [4]: cols = data_file[0].strip().split(',')
```

Finally, we iterate through the list of rows and flip them into a dictionary of columns.

The key of each dictionary element is the columns name, and the value is a list of values with a common data type.

```
In [5]: # Get the rows, but not the first, and convert them into lists
  rows = [line.strip().split(',') for line in data_file[1:]]

# Initialize the dataframe by defining a dictionary of lists, with each column
  firedata = {col:[] for col in cols}

# Iterate through the rows and convert them to columns
  for row in rows:
    for j, col in enumerate(row):
        firedata[cols[j]].append(caster[dtypes[j]](col))
```

Test to see if it worked ...

```
In [6]: firedata['Y'][:5]
Out[6]: [5, 4, 4, 6, 6]
```

# Working with spatial coordinates X, Y

For the first tasks, we grab the first two columns of our table, which define the spatial coordinates within the Monteshino park map.

```
In [7]: X, Y = firedata['X'], firedata['Y']
In [8]: X[:10], Y[:10]
Out[8]: ([7, 7, 7, 8, 8, 8, 8, 8, 8, 7], [5, 4, 4, 6, 6, 6, 6, 6, 6, 5])
```

#### Task 1

(2 points)

Write a function called coord\_builder() with these requirements:

• Takes two lists, X and Y, as inputs. X and Y must be of equal length.

- Returns a list of tuples [(x1,y1), (x2,y2), ..., (xn,yn)] where (xi,yi) are the ordered pairs from X and Y.
- Uses the zip() function to create the returned list.
- Use a list comprehension to actually build the returned list.
- Contains a docstring with short description of the function.

#### Task 2

(1 PT)

Call your coord\_builder() function, passing in X and Y.

Then print the first ten tuples.

# Working with AREA

Next, we work the area column of our data.

```
In [11]: area = firedata['area']
```

```
In [12]: area[-10:]
Out[12]: [0.0, 0.0, 2.17, 0.43, 0.0, 6.44, 54.29, 11.16, 0.0, 0.0]
```

#### Task 3

(1 PT)

Write code to print the minimum area and maximum area in a tuple (min\_value, max\_value).

Save min\_value and max\_value as floats.

```
In [13]: min_value = float(min(area))
    max_value = float(max(area))
    area_tuple = (min_value, max_value)
    print(area_tuple)
    (0.0, 1090.84)
```

#### Task 4

(2 PTS)

Write a lambda function that applies the following function to x:

```
log_{10}(1+x)
```

Return the rounded value to 2 decimals.

Assign the function to the variable mylog10.

Then call the lambda function on area and print the last 10 values.

Hints:

- Use the log10 function from Python's math module. You'll need to import it.
- Use a list comprehension to make the lambda function a one-liner.
- To get the last members of a list, used negative offset slicing. See the Python documentation on lists for a refresher on slicing.

```
In [28]: import math
    mylog10 = lambda x: [round(math.log10(1+a),2) for a in x]
    mylog10(area)[-10:]

Out[28]: [0.0, 0.0, 0.5, 0.16, 0.0, 0.87, 1.74, 1.08, 0.0, 0.0]
```

## Working with MONTH

The month column contains months of the year in abbreviated form - jan to dec.

```
In [15]: month = firedata['month']
In [16]: month[:10]
Out[16]: ['mar', 'oct', 'oct', 'mar', 'mar', 'aug', 'aug', 'aug', 'sep', 'sep']
```

#### Task 5

(1 PT)

Create a function called get\_uniques() that extracts the unique values from a list.

- Do not use set() but instead use a **dictionary comprehension** to capture the unique names.
- Hint: They keys in a dictionary are unique.
- Hint: You do not need to count how many times a name appears in the source list.

Then function should optionally return the list as sorted in ascending order.

Then apply it to the month column of our data with sorting turned on.

Then print the unique months.

['mar', 'oct', 'aug', 'sep', 'apr', 'jun', 'jul', 'feb', 'jan', 'dec', 'may',

'nov']

#### Task 6

(1 PT)

Write a lambda function called <code>get\_month\_for\_letter</code> that uses a list comprehension to select all months starting with a given letter from the list of unique month names you just created.

The function should assume that the list of unique month names exists in the global context.

The returned list should contain uppercase strings.

Run and print the result with a as the paramter.

```
In [37]: months = get_uniques(month, True)
    get_month_for_letter = lambda letter: [val.upper() for val in months if val.sta
    get_month_for_letter("a")
Out[37]: ['APR', 'AUG']
```

# Working with DMC

DMC - DMC index from the FWI system: 1.1 to 291.3

```
In [20]: dmc = firedata['DMC']
In [21]: dmc[:10]
Out[21]: [26.2, 35.4, 43.7, 33.3, 51.3, 85.3, 88.9, 145.4, 129.5, 88.0]
```

#### Task 7

(2 PTS)

Write a function called bandpass\_filter() with these requirements:

- Takes three inputs:
  - A list of numbers num\_list.
  - An integer serving as a lower bound lower\_bound.
  - An integer serving as an upper bound upper\_bound.
- Returns a new array containing only the values from the original array which are greater than lower\_bound and less than upper\_bound.

```
In [38]: def bandpass_filter(num_list, lower_bound, upper_bound):
    return [num for num in num_list if (num > lower_bound) and (num < upper_bound)</pre>
```

#### Task 8

(1 PT)

Call bandpass\_filter() passing dmc as the list, with lower\_bound=25 and upper\_bound=35.

Then print the result.

```
In [39]: print(bandpass_filter(dmc, 25, 35))
[26.2, 33.3, 32.8, 27.9, 27.4, 25.7, 33.3, 33.3, 30.7, 33.3, 25.7, 25.7, 25.7, 32.8, 27.2, 27.8, 26.4, 25.4, 25.4, 25.4, 25.4, 26.7, 25.4, 27.5, 28.0, 25.4]
```

# Working with FFMC

FFMC - FFMC index from the FWI system: 18.7 to 96.20

```
In [24]: ffmc = firedata['FFMC']
In [25]: ffmc[:10]
Out[25]: [86.2, 90.6, 90.6, 91.7, 89.3, 92.3, 92.3, 91.5, 91.0, 92.5]
```

### Task 9

(2 PTS)

Write a lambda function get\_mean that computes the mean  $\mu$  of a list of numbers.

The mean is jus the sum of a list of numeric values divided by the length of that list.

Write another lambda function get\_ssd that computes the squared deviation of a number.

- The function takes two arguments, a number from a given list and the mean of the numbers in that list.
- The function is meant to be used in a for-loop that iterates through a list.
- The squared deviation of a list element  $x_i$  is  $(x_i \mu)^2$ .

Then write get\_sum\_sq\_err() with these requirements:

- Takes a numeric list as input.
- Computes the mean  $\mu$  of the list using get\_mean .

• Computes the sum of squared deviations for the list using a list comprehension that applies get\_ssd.

• Returns the sum of squared deviations.

```
In [26]: get_mean = lambda x: sum(x)/len(x)

get_ssd = lambda x, mu: [(nums-mu)**2 for nums in x]

def get_sum_sq_error(nums):
    mu = get_mean(nums)
    ssd = get_ssd(nums, mu)
    return sum(ssd)
```

### Task 10

(1 PT)

Call sum\_sq\_err() passing ffmc as the list and print the result.

```
In [27]: get_sum_sq_error(ffmc)
Out[27]: 15723.357872340412
```