ApplyingFilters

August 8, 2023

1 Applying Filters

First, run the cell below to import the packages that we will use.

```
[1]: import pandas as pd import numpy as np import os
```

1.1 Step 1: Inspect the Data

We will be working with the "adult" data set which contains Census information from 1994.

The code cell below loads the data set by using the Pandas pd.read_csv() function to read in the CSV file that contains the data. The file is located in a folder named "data" and has the name "adult.data.partial." The pd.read_csv() function returns a Pandas DataFrame. We will assign the data to a DataFrame object called df. Run the cell below to load the data.

```
[2]: filename = os.path.join(os.getcwd(), "data", "adult.data.partial")
    df = pd.read_csv(filename, header=0)
```

In the code cell below, use the Pandas DataFrame head() method to display the first few rows of the DataFrame df.

```
[3]: # YOUR CODE HERE - this cell will not be graded df.head()
```

[3]:	age	workclass	fnlwgt	education	education-num	marital-status	\
0	36	State-gov	112074	Doctorate	16	Never-married	
1	35	Private	32528	HS-grad	9	Married-civ-spouse	
2	21	Private	270043	Some-college	10	Never-married	
3	45	Private	168837	Some-college	10	Married-civ-spouse	
4	39	Private	297449	Bachelors	13	Married-civ-spouse	
			-		7 CTD	\	

	occupation	relationship	race	sex_selilD	capital-gain	\
0	Prof-specialty	Not-in-family	White	Non-Female	0	
1	Handlers-cleaners	Husband	White	Non-Female	0	
2	Other-service	Own-child	White	Female	0	
3	Adm-clerical	Wife	White	Female	0	
4	Prof-specialty	Husband	White	Non-Female	0	

capital-loss hours-per-week native-country label

```
0
              0
                              45 United-States <=50K
1
              0
                                 United-States <=50K
2
              0
                                  United-States
                                                  <=50K
3
              0
                              24
                                         Canada
                                                   >50K
4
              0
                              40
                                                   >50K
                                 United-States
```

Use the Pandas shape property to display the number of rows and columns in the data. If you forgot the syntax, call df.shape? in the cell below to read the documentation.

How many examples (rows) do we have? How many features (columns)?

```
[4]: # YOUR CODE HERE - this cell will not be graded df.shape
```

[4]: (7000, 15)

1.2 Step 2: Random Sampling of the Data

Random sampling from the data using np.random.choice and loc

We will start by sampling some of the data. You will learn more about sampling in a future exercise.

For now, imagine that you need to randomly select 30% of the data examples.

First, we will do this the 'NumPy' way.

In the cell below, some code is already pre-written to randomly select 30% of rows and save their indices to variable indices.

The variable indices only contains the indices of rows, not the actual data in the rows.

Complete the code below to obtain only the rows in df with the indices specified in variable indices.

You will recall that you can use loc[] to index into a DataFrame to acces rows. Use loc[] accomplish this task.

Save this result to a new DataFrame named df_subset.

1.2.1 Graded Cell

The cell below will be graded. Remove the line "raise NotImplementedError()" before writing your code.

1.2.2 Self-Check

Run the cell below to test the correctness of your code above before submitting for grading. Do not add code or delete code in the cell.

```
[6]: # Run this self-test cell to check your code;
# do not add code or delete code in this cell
from jn import testSubset

try:
    p, err = testSubset(df, df_subset)
    print(err)
except Exception as e:
    print("Error!\n" + str(e))
```

Correct!

Note that you could write some of the code in the cell above in a single line, without creating a new array indices, which you likely won't use again. Note how the cell below accomplished that.

This compressed style may seem a little bit bulky and intimidating at first, but will become easier to comprehend as you get more experience.

Let's check that our sampling worked. You should expect to see that the shape of the new object df_subset reflects that it has 30% of the original row number:

```
[8]: print(df.shape) #original number of rows
print(df_subset.shape) #30% of the number of rows

(7000, 15)
(2100, 15)
```

But did you actually select the rows randomly? Look at the indices in the new DataFrame:

```
[9]: df_subset.head()
[9]:
          age workclass fnlwgt
                                    education education-num
                                                                   marital-status
   2003
          28
                Private 106951
                                                                    Never-married
                                      HS-grad
   6369
          42
                          78765 Some-college
                                                           10
                                                               Married-civ-spouse
                Private
           25
                                                               Married-civ-spouse
   1337
                Private 209428 Some-college
                                                           10
   1027
                                      HS-grad
                                                            9
           30
                    NaN
                         164940
                                                                        Separated
   275
           36
                Private
                         115700
                                      HS-grad
                                                               Married-civ-spouse
                 occupation
                              relationship
                                                   sex_selfID
                                                                capital-gain
                                              race
   2003
         Handlers-cleaners
                             Not-in-family
                                            White
                                                    Non-Female
                                                                           0
   6369
                                                   Non-Female
                                                                         3103
               Craft-repair
                                   Husband White
   1337
                      Sales
                                   Husband White
                                                   Non-Female
                                                                           0
                        NaN
   1027
                                 Unmarried Black
                                                        Female
                                                                           0
   275
                      Sales
                                   Husband White Non-Female
                                                                           0
```

	capital-loss	hours-per-week	native-country	label
2003	0	42	United-States	<=50K
6369	0	45	United-States	>50K
1337	0	25	El-Salvador	<=50K
1027	0	25	United-States	<=50K
275	0	50	United-States	<=50K

It seems random. To convince yourself that it is, try running the sampling code above again, and then re-run the head() method to above and inspect the results. You should see a different random sample each time you re-run the sampling code cell.

We will now see how to perform sampling using the Pandas way:

```
[10]: percentage = 0.3
     num_rows = df.shape[0]
     df_subset = df.sample(int(percentage*num_rows))
     df_subset.head()
[10]:
           age workclass
                           fnlwgt
                                       education
                                                   education-num
                                                                       marital-status
                           181655
     6702
            26
                  Private
                                       Assoc-voc
                                                               11
                                                                   Married-civ-spouse
     3748
            18
                           210828
                                    Some-college
                                                               10
                                                                        Never-married
                  Private
     4430
            41
                                    Some-college
                  Private
                           187802
                                                               10
                                                                              Divorced
     3923
            46
                           133616
                                    Some-college
                  Private
                                                               10
                                                                              Divorced
     5022
            27
                  Private
                           142075
                                    Some-college
                                                               10
                                                                        Never-married
                   occupation
                                 relationship
                                                       sex selfID
                                                                    capital-gain
                                                 race
     6702
                 Adm-clerical
                                      Husband White
                                                       Non-Female
                                                                                0
                                                       Non-Female
     3748
           Handlers-cleaners
                                    Own-child Other
                                                                                0
     4430
                 Tech-support
                               Not-in-family
                                               White
                                                       Non-Female
                                                                                0
     3923
                 Adm-clerical
                                    Unmarried White
                                                           Female
                                                                                0
     5022
                Other-service
                                    Own-child White
                                                       Non-Female
                                                                                0
           capital-loss
                          hours-per-week native-country
                                                           label
     6702
                    2377
                                           United-States
                                                           <=50K
                                       45
     3748
                       0
                                                           <=50K
                                       30
                                           United-States
     4430
                       0
                                       50
                                           United-States
                                                           <=50K
                                           United-States
     3923
                       0
                                       40
                                                           <=50K
     5022
                       0
                                           United-States
                                                           <=50K
```

1.3 Step 3: Filter a DataFrame by Column Values

Imagine that you want to examine only the private sector employees that we have in DataFrame df. The cell below contains a conditional statement df['workclass'] =='Private'

This will evaluate to a collection of True/False values per row. A value of True indicates that the corresponding row fulfills the condition. This collection of True/False values is of data type Pandas Series (a one-dimensional array). The array is assigned to variable condition.

Run the cell below and inspect the results.

```
[11]: condition = df['workclass'] == 'Private'
     condition
[11]: 0
              False
               True
     1
     2
               True
     3
               True
     4
               True
     6995
               True
     6996
               True
     6997
              False
     6998
               True
     6999
               True
     Name: workclass, Length: 7000, dtype: bool
```

In the code cell below, use the condition variable to extract the private employee sector data from data DataFrame df. Hint: Index into df using bracket notation and supply it the variable condition. Save the results to variable df_private. Use the head() method to inspect the new DataFrame df_private.

1.3.1 Graded Cell

The cell below will be graded. Remove the line "raise NotImplementedError()" before writing your code.

```
[12]: # YOUR CODE HERE
     df_private=df[condition]
     df_private.head()
                                   education education-num
                                                                   marital-status
[12]:
        age workclass
                        fnlwgt
         35
                         32528
                                     HS-grad
                                                               Married-civ-spouse
     1
              Private
                                                            9
     2
         21
              Private
                        270043
                                Some-college
                                                           10
                                                                    Never-married
     3
         45
                                Some-college
                                                               Married-civ-spouse
              Private
                        168837
                                                           10
     4
         39
                        297449
                                   Bachelors
                                                               Married-civ-spouse
              Private
                                                           13
     5
         27
              Private
                        233421
                                Some-college
                                                           10
                                                                    Never-married
               occupation relationship
                                                 sex_selfID
                                                              capital-gain
                                           race
        Handlers-cleaners
                                                 Non-Female
     1
                                Husband
                                          White
     2
            Other-service
                              Own-child White
                                                     Female
                                                                          0
     3
             Adm-clerical
                                   Wife
                                         White
                                                     Female
                                                                          0
     4
           Prof-specialty
                                                                          0
                                Husband White Non-Female
             Adm-clerical
     5
                              Own-child White
                                                 Non-Female
                                                                          0
        capital-loss
                       hours-per-week native-country
                                                       label
                                        United-States
                                                        <=50K
     1
                    0
                                   45
     2
                    0
                                   16
                                        United-States
                                                       <=50K
     3
                    0
                                   24
                                               Canada
                                                        >50K
     4
                    0
                                       United-States
                                                        >50K
                                   40
     5
                    0
                                   20
                                       United-States <=50K
```

1.3.2 Self-Check

Run the cell below to test the correctness of your code above before submitting for grading. Do not add code or delete code in the cell.

```
[13]: # Run this self-test cell to check your code;
# do not add code or delete code in this cell
from jn import testPrivate

try:
    p, err = testPrivate(df, df_private, condition)
    print(err)
except Exception as e:
    print("Error!\n" + str(e))
```

Correct!

How many of the rows are in the new DataFrame df_private?

In the cell below, display the number of rows in DataFrame df_private using the shape property. Save the results to variable num_rows and print num_rows. Hint: Recall that the shape property returns a tuple, with the first value corresponding to the number of rows and the second value corresponding to the number of columns.

1.3.3 Graded Cell

The cell below will be graded. Remove the line "raise NotImplementedError()" before writing your code.

```
[14]: # YOUR CODE HERE
num_rows=df_private.shape[0]
print(num_rows)
```

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1.3.4 Self-Check

Run the cell below to test the correctness of your code above before submitting for grading. Do not add code or delete code in the cell.

```
[15]: # Run this self-test cell to check your code;
# do not add code or delete code in this cell
from jn import testRows

try:
    p, err = testRows(num_rows)
    print(err)
except Exception as e:
    print("Error!\n" + str(e))
```

Correct!

1.4 Step 4. Data Analysis using Filtering

The code cell below finds the average age of people who self-reported as female in DataFrame df.

```
[16]: condition = df['sex_selfID'] == 'Female'
df[condition]['age'].mean()
```

[16]: 36.764213309828115

Notice that here we do not create a new DataFrame for the filtered data. Instead, we perform the computation and display the result. If you do not anticipate working further with a subset of your DataFrame

(e.g., querying it or finding more summary statistics about Females), then you don't need to save your results to a new DataFrame object.

As a practice, use the code cell below to play around with the statement that computes the mean: df[condition]['age'].mean():

In particular: - Write df[condition] in the cell. run the cell and inspect the results. - Next, write df[condition]['age'] in the cell. Run the cell and inspect the new DataFrame. - Next, write df[condition]['age'].mean() in the cell. Run the cell and inspect the results. _____

```
45
8
        20
10
        54
17
        41
6980
        57
6985
        44
6986
        27
6991
        17
6996
        19
Name: age, Length: 2269, dtype: int64>
```

Next you want to know how many people work for the local government for more than 40 hours per week. Using the code above as a guide, in the code cell below: 1. Define the conditions that will find the appropriate data from DataFrame df. 2. Apply the condition to DataFrame df.

- 3. Use the shape property to obtain the number of rows and assign the results to variable rows. Follow these steps:
 - 1. Create the first condition and name it condition1. condition1 will look for the number of people who work for the local government. Employment information is found in the column workclass. The value is Local-gov.
 - 2. Create the second condition and name it condition2. condition2 will check whether the number of hours worked per week is more than 40 hours. The number of hours worked can be found in the column hours-per-week.
 - 3. Combine these two conditions using the & operator to create a compound statement. Assign that to variable condition (condition = condition1 & condition2).

- 4. Apply condition to DataFrame df using bracket notation, and save the result to DataFramedf_local.
- 5. Use the shape property to obtain the number of rows in df_local. Assign the result to variable rows.

1.4.1 Graded Cell

The cell below will be graded. Remove the line "raise NotImplementedError()" before writing your code.

```
[21]: # YOUR CODE HERE
    condition1=df['workclass']=='Local-gov'
    condition2=df['hours-per-week']>40
    condition=condition1&condition2
    df_local=df[condition]
    rows=df_local.shape[0]
```

1.4.2 Self-Check

Run the cell below to test the correctness of your code above before submitting for grading. Do not add code or delete code in the cell.

```
[22]: # Run this self-test cell to check your code;
# do not add code or delete code in this cell
from jn import testCondition

try:
    p, err = testCondition(df, condition1, condition2, condition, df_local, 
    →rows)
    print(err)
except Exception as e:
    print("Error!\n" + str(e))
```

Correct!

Sometimes your data may contain missing values. One such column that contains missing values in DataFrame df is native-country. Not everyone's native country has been supplied. Such columns contain the value Nan.

The code cell below randomly samples 50% of rows for which the native country information is available and ignores missing values. It uses pandas notnull() method. You can read more about notnull() in the online documentation.

```
[23]: percentage = 0.5

# obtain all rows in which the column 'native-country' contains a value
df_country_notnull = df[df['native-country'].notnull()]

# obtain the number of rows in df_country_notnull
num_rows = df_country_notnull.shape[0]
```

```
# obtain a 50% random sample of rows from df_country_notnull and save the_
indices of these rows

indices = np.random.choice(df_country_notnull.index,_
size=int(percentage*num_rows), replace=False)

# using the row indices, save these row values to new DataFrame df_filtered

df_filtered = df_country_notnull.loc[indices]
```

In the code cell below, find the mean age of individuals in DataFrame df_filtered and save the value to variable mean_age.

1.4.3 Graded Cell

The cell below will be graded. Remove the line "raise NotImplementedError()" before writing your code.

```
[24]: # YOUR CODE HERE
mean_age=df_filtered['age'].mean()
```

1.4.4 Self-Check

Run the cell below to test the correctness of your code above before submitting for grading. Do not add code or delete code in the cell.

```
[25]: # Run this self-test cell to check your code;
# do not add code or delete code in this cell
from jn import testMean
try:
    p, err = testMean(df_filtered, mean_age)
    print(err)
except Exception as e:
    print("Error!\n" + str(e))
```

Correct!

You have been selecting a single column (e.g., 'age') by using bracket notation df_filtered['age']. You will sometimes also encounter columns being selected using dot notation df_filtered.age. Note that this won't work if the column name includes hyphens or any other special symbols. We will stick to providing names as strings in square brackets.