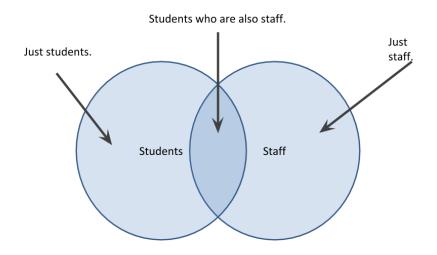
MergingDataFrame_ed

November 15, 2021

In this lecture we're going to address how you can bring multiple dataframe objects together, either by merging them horizontally, or by concatenating them vertically. Before we jump into the code, we need to address a little relational theory and to get some language conventions down. I'm going to bring in an image to help explain some concepts.

6: Venn Diagram



Venn Diagram

Ok, this is a Venn Diagram. A Venn Diagram is traditionally used to show set membership. For example, the circle on the left is the population of students at a university. The circle on the right is the population of staff at a university. And the overlapping region in the middle are all of those students who are also staff. Maybe these students run tutorials for a course, or grade assignments, or engage in running research experiments.

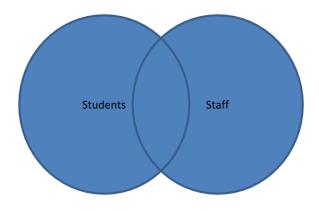
So, this diagram shows two populations whom we might have data about, but there is overlap between those populations.

When it comes to translating this to pandas, we can think of the case where we might have these two populations as indices in separate DataFrames, maybe with the label of Person Name.

When we want to join the DataFrames together, we have some choices to make. First what if we want a list of all the people regardless of whether they're staff or student, and all of the information we can get on them? In database terminology, this is called a full outer join. And in set theory, it's called a union. In the Venn diagram, it represents everyone in any circle.

Here's an image of what that would look like in the Venn diagram.

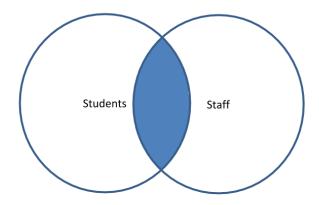
7: Full outer join (union)



Union

It's quite possible though that we only want those people who we have maximum information for, those people who are both staff and students. Maybe being a staff member and a student involves getting a tuition waiver, and we want to calculate the cost of this. In database terminology, this is called an inner join. Or in set theory, the intersection. It is represented in the Venn diagram as the overlapping parts of each circle.

7: Inner join (intersection)



Here's what that looks like:

```
[1]: # With that background, let's see an example of how we would do this in pandas,
     →where we would use the merge
    # function.
    import pandas as pd
    # First we create two DataFrames, staff and students.
    staff_df = pd.DataFrame([{'Name': 'Kelly', 'Role': 'Director of HR'},
                              {'Name': 'Sally', 'Role': 'Course liasion'},
                              {'Name': 'James', 'Role': 'Grader'}])
    # And lets index these staff by name
    staff_df = staff_df.set_index('Name')
    # Now we'll create a student dataframe
    student_df = pd.DataFrame([{'Name': 'James', 'School': 'Business'},
                                {'Name': 'Mike', 'School': 'Law'},
{'Name': 'Sally', 'School': 'Engineering'}])
    # And we'll index this by name too
    student_df = student_df.set_index('Name')
    # And lets just print out the dataframes
    print(staff_df.head())
    print(student_df.head())
```

Role
Name
Kelly Director of HR
Sally Course liasion
James Grader

School

Name

James Business
Mike Law
Sally Engineering

- [2]: # There's some overlap in these DataFrames in that James and Sally are bothustudents and staff, but Mike and
 # Kelly are not. Importantly, both DataFrames are indexed along the value we want to merge them on, which is
 # called Name.
- [3]: # If we want the union of these, we would call merge() passing in the DataFrame

 on the left and the DataFrame

 # on the right and telling merge that we want it to use an outer join. We want

 to use the left and right

 # indices as the joining columns.

 pd.merge(staff_df, student_df, how='outer', left_index=True, right_index=True)
- [3]: Role School
 Name
 James Grader Business
 Kelly Director of HR NaN

Mike NaN Law

Sally Course liasion Engineering

- [4]: Role School
 Name
 Sally Course liasion Engineering
 James Grader Business
- [5]: # And we see the resulting DataFrame has only James and Sally in it. Now there

 → are two other common use cases

 # when merging DataFrames, and both are examples of what we would call set

 → addition. The first is when we

 # would want to get a list of all staff regardless of whether they were

 → students or not. But if they were

```
# students, we would want to get their student details as well. To do this we
    →would use a left join. It is
    # important to note the order of dataframes in this function: the first _{f \sqcup}
    \rightarrow dataframe is the left dataframe and
    # the second is the right
    pd.merge(staff_df, student_df, how='left', left_index=True, right_index=True)
[5]:
                    Role
                               School
   Name
   Kelly Director of HR
                                  NaN
   Sally Course liasion
                         Engineering
    James
                  Grader
                             Business
[6]: # You could probably guess what comes next. We want a list of all of the
    ⇒students and their roles if they were
    # also staff. To do this we would do a right join.
    pd.merge(staff_df, student_df, how='right', left_index=True, right_index=True)
[6]:
                    Role
                               School
   Name
    James
                  Grader
                             Business
   Mike
                     NaN
                                  Law
   Sally Course liasion Engineering
[7]: # We can also do it another way. The merge method has a couple of other
    →interesting parameters. First, you
    →example. Here we have a
    # parameter called "on", and we can assign a column that both dataframe has as \Box
    → the joining column
    # First, lets remove our index from both of our dataframes
    staff_df = staff_df.reset_index()
    student_df = student_df.reset_index()
    # Now lets merge using the on parameter
   pd.merge(staff_df, student_df, how='right', on='Name')
[7]:
       Name
                       Role
                                  School
   0 Sally
             Course liasion Engineering
    1 James
                     Grader
                                Business
      Mike
                        NaN
                                     Law
[8]: # Using the "on" parameter instead of a the index is how I find myself using \Box
    \rightarrowmerge() the most.
[9]: # So what happens when we have conflicts between the DataFrames? Let's take a_{\sqcup}
    → look by creating new staff and
    # student DataFrames that have a location information added to them.
```

```
staff_df = pd.DataFrame([{'Name': 'Kelly', 'Role': 'Director of HR',
                                'Location': 'State Street'},
                              {'Name': 'Sally', 'Role': 'Course liasion',
                                'Location': 'Washington Avenue'},
                              {'Name': 'James', 'Role': 'Grader',
                                'Location': 'Washington Avenue'}])
     student_df = pd.DataFrame([{'Name': 'James', 'School': 'Business',
                                  'Location': '1024 Billiard Avenue'},
                                {'Name': 'Mike', 'School': 'Law',
                                  'Location': 'Fraternity House #22'},
                                 {'Name': 'Sally', 'School': 'Engineering',
                                  'Location': '512 Wilson Crescent'}])
     # In the staff DataFrame, this is an office location where we can find the L
     ⇒staff person. And we can see the
     # Director of HR is on State Street, while the two students are on Washington
     → Avenue, and these locations just
     # happen to be right outside my window as I film this. But for the student \Box
     → DataFrame, the location information
     # is actually their home address.
     # The merge function preserves this information, but appends an x or y to
     →help differentiate between which
     # index went with which column of data. The \_x is always the left DataFrame_{\sqcup}
     →information, and the _y is always
     # the right DataFrame information.
     # Here, if we want all the staff information regardless of whether they were
     →students or not. But if they were
     \# students, we would want to get their student details as well. Then we can do a_{\sqcup}
     \rightarrow left join and on the column of
     # Name
     pd.merge(staff_df, student_df, how='left', on='Name')
[9]:
       Name
                         Role
                                      Location_x
                                                        School
                                                                           Location_y
     O Kelly Director of HR
                                    State Street
                                                           NaN
                                                                                  NaN
     1 Sally Course liasion Washington Avenue Engineering
                                                                 512 Wilson Crescent
     2 James
                       Grader Washington Avenue
                                                      Business 1024 Billiard Avenue
[10]: # From the output, we can see there are columns Location_x and Location_y._{\sqcup}
     \rightarrowLocation_x refers to the Location
     # column in the left dataframe, which is staff dataframe and Location_y refers_
     →to the Location column in the
     # right dataframe, which is student dataframe.
```

```
# Before we leave merging of DataFrames, let's talk about multi-indexing and
     →multiple columns. It's quite
     # possible that the first name for students and staff might overlap, but the
      \rightarrow last name might not. In this
     # case, we use a list of the multiple columns that should be used to join keys_{\sqcup}
     → from both dataframes on the on
     \# parameter. Recall that the column name(s) assigned to the on parameter needs \sqcup
     →to exist in both dataframes.
     # Here's an example with some new student and staff data
     staff_df = pd.DataFrame([{'First Name': 'Kelly', 'Last Name': 'Desjardins',
                               'Role': 'Director of HR'},
                              {'First Name': 'Sally', 'Last Name': 'Brooks',
                               'Role': 'Course liasion'},
                              {'First Name': 'James', 'Last Name': 'Wilde',
                               'Role': 'Grader'}])
     student_df = pd.DataFrame([{'First Name': 'James', 'Last Name': 'Hammond',
                                  'School': 'Business'},
                                {'First Name': 'Mike', 'Last Name': 'Smith',
                                 'School': 'Law'},
                                {'First Name': 'Sally', 'Last Name': 'Brooks',
                                  'School': 'Engineering'}])
     # As you see here, James Wilde and James Hammond don't match on both keys since
     → they have different last
     # names. So we would expect that an inner join doesn't include these
     →individuals in the output, and only Sally
     # Brooks will be retained.
    pd.merge(staff_df, student_df, how='inner', on=['First Name','Last Name'])
[10]: First Name Last Name
                                       Role
                                                   School
            Sallv
                     Brooks Course liasion Engineering
[11]: # Joining dataframes through merging is incredibly common, and you'll need to
     →know how to pull data from
     # different sources, clean it, and join it for analysis. This is a staple not_{\sqcup}
     →only of pandas, but of database
     # technologies as well.
[12]: # If we think of merging as joining "horizontally", meaning we join on similar
     →values in a column found in two
     \# dataframes then concatenating is joining "vertically", meaning we put
     \rightarrow dataframes on top or at the bottom of
     # each other
     # Let's understand this from an example. You have a dataset that tracks some
     → information over the years. And
```

```
→ has the exactly same columns.
     # What happens if you want to put all the data, from all years' record,
      →together? You can concatenate them.
[13]: # Let's take a look at the US Department of Education College Scorecard data Itu
      →has each US university's data
     # on student completion, student debt, after-graduation income, etc. The data_{\sqcup}
     → is stored in separate CSV's with
     # each CSV containing a year's record Let's say we want the records from 2011_{\sqcup}
     →to 2013 we first create three
     # dataframe, each containing one year's record. And, because the csv files_
     \rightarrow we're working with are messy, I
     # want to supress some of the jupyter warning messages and just tell read_csvu
     →to ignore bad lines, so I'm
     # going to start the cell with a cell magic called %%capture
[14]: | %%capture
     df_2011 = pd.read_csv("datasets/college_scorecard/MERGED2011_12_PP.csv", __
      →error_bad_lines=False)
     df_2012 = pd.read_csv("datasets/college_scorecard/MERGED2012_13_PP.csv", __
      →error_bad_lines=False)
     df_2013 = pd.read_csv("datasets/college_scorecard/MERGED2013_14_PP.csv", __
      →error_bad_lines=False)
[15]: # Let's get a view of one of the dataframes
     df_2011.head(3)
[15]:
          UNITID
                      OPEID OPEID6
                                                                    INSTNM
     0 100654.0
                   100200.0
                               1002
                                                 Alabama A & M University
     1 100663.0
                   105200.0
                               1052 University of Alabama at Birmingham
     2 100690.0
                  2503400.0 25034
                                                       Amridge University
              CITY STABBR
                                        ACCREDAGENCY INSTURL NPCURL
                                   ZIP
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     2 Montgomery
                        AL
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```

each year's record is a separate CSV and every CSV ofr every year's record

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                                                               NaN
     [3 rows x 1977 columns]
[16]: # We see that there is a whopping number of columns - more than 1900! We can
      →calculate the length of each
     # dataframe as well
     print(len(df_2011))
     print(len(df_2012))
     print(len(df_2013))
    15235
    7793
    7804
[17]: # That's a bit surprising that the number of schools in the scorecard for 2011_{\square}
      → is almost double that of the
     # next two years. But let's not worry about that. Instead, let's just put allu
      →three dataframes in a list and
     # call that list frames and pass the list into the concat() function Let's see __
      →what it looks like
     frames = [df_2011, df_2012, df_2013]
     pd.concat(frames)
[17]:
               UNITID
                            OPEID OPEID6 \
     0
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                        100200.0
                                    1002
     1
             100663.0
                        105200.0
                                    1052
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             100690.0
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                                    1055
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                        100500.0
                                    1005
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                                    1571
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4		Ala	bama State	e Unive	rsity	Montgomery	AL	
779	99 Georgi	la Military Co	llege-Col	umbus C	ampus	Columbus	GA	
780	00 Georgi	la Military Co	llege-Valo	dosta C	ampus	Valdosta	GA	
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780)2	Georgia Mi	litary Col	llege-0	nline	Milledgeville	GA	
780)3 Georg	gia Military C	College-Sto	one Mou	ntain	Stone Mountain	GA	
	ZIP	ACCREDAGENCY	INSTURL N	NPCURL		\		
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1	35294-0110	NaN	NaN	NaN				
2	36117-3553	NaN	NaN	NaN				
3	35899	NaN	NaN	NaN				
4	36104-0271	NaN	NaN	NaN				
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OMENRYP_FULLTIME_POOLED_SUPP OMENRAP_FULLTIME_POOLED_SUPP \

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7700	 N – N	 N - N
7799		NaN NaN
7800		NaN
7801	NaN	NaN
7802		NaN
7803	NaN	NaN

[30832 rows x 1977 columns]

7802

```
[18]: # As you can see, we have more observations in one dataframe and columns remain
     → the same. If we scroll down to
     # the bottom of the output, we see that there are a total of 30,832 rows after \Box
     ⇔concatenating three dataframes.
     # Let's add the number of rows of the three dataframes and see if the two_{\sf L}
     →numbers match
     len(df_2011)+len(df_2012)+len(df_2013)
[18]: 30832
[19]: \# The two numbers match! Which means our concatenation is successful. But wait,
     →now that all the data is
     # concatenated together, we don't know what observations are from what year_
     → anymore! Actually the concat
     # function has a parameter that solves such problem with the keys parameter, we_
     →can set an extra level of
     → dataframes into the keys parameter
     # Now let's try it out
     pd.concat(frames, keys=['2011','2012','2013'])
[19]:
                   UNITID
                               OPEID OPEID6 \
    2011 0
                 100654.0
                            100200.0
                                       1002
                 100663.0
                            105200.0
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         1
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    2013 7799 48285703.0
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    2011 0
                                    Alabama A & M University
                                                                     Normal
         1
                         University of Alabama at Birmingham
                                                                 Birmingham
         2
                                          Amridge University
                                                                 Montgomery
         3
                         University of Alabama in Huntsville
                                                                 Huntsville
         4
                                    Alabama State University
                                                                 Montgomery
    2013 7799
                    Georgia Military College-Columbus Campus
                                                                   Columbus
         7800
                    Georgia Military College-Valdosta Campus
                                                                   Valdosta
         7801
               Georgia Military College-Warner Robins Campus
                                                              Warner Robins
```

Georgia Military College-Online

Milledgeville

	7803		Georgia Mili	tary Coll	ege-S	tone Mo	untain	Stone	Mountai	l n	
		STABBR	ZIP	ACCREDAG	ENCY	INSTURL	NPCURL		\		
2011	0	AL	35762		NaN	NaN	NaN				
	1	AL	35294-0110		${\tt NaN}$	NaN	NaN				
	2	AL	36117-3553		${\tt NaN}$	NaN	NaN				
	3	AL	35899		${\tt NaN}$	NaN	NaN				
	4	AL	36104-0271		NaN	NaN	NaN	• • •			
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2013	7799	GA	31909		NaN	NaN		• • •			
	7800	GA	31605		NaN	NaN		• • •			
	7801	GA	31093		NaN	NaN		• • •			
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. . .
2013 7799
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2011 0
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2013 7799
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     7803
                                       NaN
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```

[30832 rows x 1977 columns]

```
[20]: # Now we have the indices as the year so we know what observations are from what year. You should know that

# concatenation also has inner and outer method. If you are concatenating two dataframes that do not have

# identical columns, and choose the outer method, some cells will be NaN. If you choose to do inner, then some

# observations will be dropped due to NaN values. You can think of this as analogous to the left and right

# joins of the merge() function.
```

Now you know how to merge and concatenate datasets together. You will find such functions very useful for combining data to get more complex or complicated results and to do analysis with. A solid understanding of how to merge data is absolutely essentially when you are procuring,

cleaning, and manipulating data. It's worth knowing how to join different datasets quickly, and the different options you can use when joining datasets, and I would encourage you to check out the pandas docs for joining and concatenating data.

PandasIdioms_ed

November 15, 2021

Python programmers will often suggest that there many ways the language can be used to solve a particular problem. But that some are more appropriate than others. The best solutions are celebrated as Idiomatic Python and there are lots of great examples of this on StackOverflow and other websites.

A sort of sub-language within Python, Pandas has its own set of idioms. We've alluded to some of these already, such as using vectorization whenever possible, and not using iterative loops if you don't need to. Several developers and users within the Panda's community have used the term **pandorable** for these idioms. I think it's a great term. So, I wanted to share with you a couple of key features of how you can make your code pandorable.

[1]: # Let's start by bringing in our data processing libraries

```
import pandas as pd
    import numpy as np
    # And we'll bring in some timing functionality too, from the timeit module
    import timeit
    # And lets look at some census data from the US
    df = pd.read_csv('datasets/census.csv')
    df.head()
[1]:
       SUMLEV
               REGION DIVISION STATE
                                         COUNTY
                                                  STNAME
                                                                  CTYNAME
    0
           40
                    3
                                      1
                                              0 Alabama
                                                                  Alabama
           50
                    3
                                                 Alabama Autauga County
    1
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                                                 Alabama Baldwin County
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    3
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                                                 Alabama Barbour County
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    4
           50
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                                                 Alabama
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       CENSUS2010POP ESTIMATESBASE2010 POPESTIMATE2010
                                                                 RDOMESTICMIG2011
    0
             4779736
                                4780127
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    1
               54571
                                   54571
                                                    54660 ...
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    2
              182265
                                                                        14.832960
                                  182265
                                                   183193
    3
               27457
                                   27457
                                                    27341
                                                                        -4.728132
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    4
               22915
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                                                                        -5.527043
       RDOMESTICMIG2012 RDOMESTICMIG2013 RDOMESTICMIG2014 RDOMESTICMIG2015
    0
              -0.193196
                                 0.381066
                                                    0.582002
                                                                      -0.467369
    1
              -2.915927
                                 -3.012349
                                                    2.265971
                                                                      -2.530799
    2
              17.647293
                                21.845705
                                                   19.243287
                                                                      17.197872
```

```
-2.500690
                                -7.056824
                                                  -3.904217
                                                                   -10.543299
   3
              -5.068871
                                -6.201001
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      RNETMIG2011 RNETMIG2012 RNETMIG2013 RNETMIG2014 RNETMIG2015
   0
         1.030015
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                                   1.383282
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   1
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                      -2.626146
                                   -2.722002
                                                 2.592270
                                                             -2.187333
   2
        15.844176
                    18.559627
                                   22.727626
                                             20.317142
                                                            18.293499
   3
        -4.874741
                     -2.758113
                                  -7.167664
                                                -3.978583
                                                            -10.543299
        -5.088389
                     -4.363636
                                   -5.403729
                                                 0.754533
                                                              1.107861
    [5 rows x 100 columns]
[2]: # The first of the pandas idioms I would like to talk about is called method
     → chaining. The general idea behind
    # method chaining is that every method on an object returns a reference to that
    →object. The beauty of this is
    # that you can condense many different operations on a DataFrame, for instance, __
    ⇒into one line or at least one
    # statement of code.
    # Here's the pandorable way to write code with method chaining. In this code_
    \hookrightarrow I'm going to pull out the state
    # and city names as a multiple index, and I'm going to do so only for data,
    →which has a summary level of 50,
    # which in this dataset is county-level data. I'll rename a column too, just tou
    \rightarrow make it a bit more readable.
    (df.where(df['SUMLEV']==50)
        .dropna()
        .set_index(['STNAME','CTYNAME'])
        .rename(columns={'ESTIMATESBASE2010': 'Estimates Base 2010'}))
[2]:
                               SUMLEV REGION DIVISION STATE COUNTY \
   STNAME CTYNAME
   Alabama Autauga County
                                 50.0
                                          3.0
                                                    6.0
                                                           1.0
                                                                   1.0
            Baldwin County
                                 50.0
                                          3.0
                                                           1.0
                                                                   3.0
                                                    6.0
            Barbour County
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                                          3.0
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                                                                   5.0
            Bibb County
                                 50.0
                                          3.0
                                                    6.0
                                                           1.0
                                                                   7.0
            Blount County
                                 50.0
                                          3.0
                                                    6.0
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                                  . . .
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                                          4.0
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                                                                  37.0
    Wyoming Sweetwater County
                                 50.0
                                                    8.0
            Teton County
                                 50.0
                                          4.0
                                                    8.0
                                                          56.0
                                                                  39.0
            Uinta County
                                 50.0
                                          4.0
                                                    8.0
                                                          56.0
                                                                  41.0
            Washakie County
                                 50.0
                                          4.0
                                                    8.0
                                                          56.0
                                                                   43.0
```

2

4.0

8.0

CENSUS2010POP Estimates Base 2010 \

56.0

45.0

50.0

Weston County

STNAME CTYNAME

Alabama	Autauga County		54571.0		54571.	0		
	Baldwin County		182265.0		182265.			
	Barbour County		27457.0		27457.			
	Bibb County		22915.0		22919.			
	Blount County		57322.0		57322.			
			•••		•••			
Wvoming	Sweetwater County		43806.0		43806.			
,	Teton County		21294.0		21294.			
	Uinta County		21118.0		21118.			
	Washakie County		8533.0		8533.			
	Weston County		7208.0		7208.			
	weston country		1200.0		7200.	O		
		PUPE	STIMATE2010	POPES	TTMATE2011	POPEST	TMATE2012	\
STNAME	CTYNAME	1 01 1	D111111112010	1 01 110	711111111111111111111111111111111111111	1 01 201	111111111111111111111111111111111111111	`
	Autauga County		54660.0		55253.0		55175.0	
ATADAMA	Baldwin County		183193.0		186659.0		190396.0	
	Barbour County		27341.0		27226.0		27159.0	
	Bibb County		22861.0		22733.0		22642.0	
	· · · · · · · · · · · · · · · · · · ·		57373.0		57711.0			
	Blount County						57776.0	
	Sweetwater County		43593.0		44041.0		45104.0	
wyoming	•						21697.0	
	Teton County		21297.0		21482.0			
	Uinta County		21102.0		20912.0		20989.0	
	Washakie County		8545.0		8469.0		8443.0	
	Weston County		7181.0		7114.0		7065.0	
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STNAME	CTYNAME		IDUNESTICHT	G2011	TOUTEDITO	142012	`	
	Autauga County	• • •	7 0	42091	-2	915927		
Alabalia	Baldwin County	• • •		32960		647293		
	Barbour County			28132		500690		
	Bibb County			27043		068871		
	Blount County	• • •		07375		177622		
	brount County	• • •	1.0	01313	-1.	111022		
Urramina	Crostroton Country	• • •	1 0	70642	16	042100		
wyoming	Sweetwater County	• • •		72643 89565		243199 972695		
	Teton County	• • •						
	Uinta County	• • •		55986		916350		
	Washakie County	• • •		37475		827815		
	Weston County	• • •	-11.7	52361	-8.	040059		
		мОлд	ESTICMIG2013	Вром	ESTICMIG201	1		
STNAME	CTYNAME	מוטעה	EDITORIGZ013	เมบเฟ	IPOITGE/I	4 \		
	Autauga County		-3.012349	ı	2.26597	1		
111 abana	Baldwin County		21.845705		19.24328			
	Barbour County		-7.056824		-3.90421			
	Bibb County		-6.201001		-0.17753			
	•							
	Blount County		-1.748766	1	-2.06253	Ð		

 Uzomina	Sweetwater County	-5.33	0774	_	14.25	2880	
wyoming	Teton County	19.52		14.143021			
	•			_	14.14		
	Uinta County	-6.90 -2.01			17.78		
	Washakie County			_			
	Weston County	12.37	2563		1.53	3035	
		RDOMESTICMIG	2015	RNETMIG	2011	RNETMIG2012	\
STNAME	CTYNAME	RDUNESTICKIG	2015	RNEIMIG	2011	RNEIMIG2012	\
	Autauga County	-2.53	0700	7 60	6016	-2.626146	
ATADAMA	Baldwin County	17.19		15.84		18.559627	
	Barbour County	-10.54		-4.87		-2.758113	
	Bibb County	0.17		-5.08		-4.363636	
	Blount County	-1.36			9511	-0.848580	
	brount Country	1.50		1.00		0.040000	
Www.ing	Sweetwater County	-14.24		1 25	5221	16.243199	
wyoming	Teton County	-0.56			4527	2.408578	
	Uinta County	-12.12		-18.13		-5.536861	
	Washakie County		2288			-1.182592	
	Weston County		5294	-12.03		-8.040059	
	weston County	0.93	52 54	-12.03	2119	-0.040039	
		RNETMIG2013	RNET	MIG2014	RNET	MIG2015	
STNAME	CTYNAME	IMMETHICZOTO	1011111	IIIUZUII	101011	11142010	
	Autauga County	-2.722002	2	.592270	-2	.187333	
III ab ama	Baldwin County	22.727626	_	.317142		.293499	
	Barbour County	-7.167664		.978583		.543299	
	Bibb County	-5.403729		.754533		.107861	
	Blount County	-1.402476		.577232		.884411	
	Broans councy		_		Ū		
Wvoming	Sweetwater County	-5.295460	-14	.075283	-14	.070195	
,8	Teton County	21.160658		.308671		.520747	
	Uinta County	-7.521840		.740608		.606351	
	Washakie County	-2.250385		.020168		.441961	
	Weston County	12.372583		.533635		.935294	
		: 3. 2000	_		Ū		

[3142 rows x 98 columns]

```
\hookrightarrow I could have done, I began the
    # statement with a parenthesis, which tells python I'm going to span the
    ⇒statement over multiple lines for
    # readability.
[4]: # Here's a more traditional, non-pandorable way, of writing this. There's
    →nothing wrong with this code in the
    \# functional sense, you might even be able to understand it better as a new_l
    ⇒person to the language. It's just
    # not as pandorable as the first example.
    # First create a new dataframe from the original
    df = df[df['SUMLEV'] == 50] # I'll use the overloaded indexing operator [] which_
    → drops nans
    # Update the dataframe to have a new index, we use inplace=True to do this in
     \rightarrow place
    df.set_index(['STNAME','CTYNAME'], inplace=True)
    # Set the column names
    df.rename(columns={'ESTIMATESBASE2010': 'Estimates Base 2010'})
[4]:
                               SUMLEV REGION DIVISION STATE COUNTY \
   STNAME CTYNAME
   Alabama Autauga County
                                   50
                                             3
                                                       6
                                                                      1
           Baldwin County
                                   50
                                             3
                                                       6
                                                                      3
                                             3
                                                       6
                                                                      5
            Barbour County
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            Bibb County
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            Blount County
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                                                            . . .
                                  . . .
                                           . . .
                                                    . . .
    Wyoming Sweetwater County
                                           4
                                                            56
                                                                     37
                                  50
                                                     8
            Teton County
                                  50
                                                             56
                                                                     39
                                            4
                                                      8
            Uinta County
                                   50
                                            4
                                                       8
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                                                                     41
            Washakie County
                                   50
                                             4
                                                       8
                                                             56
                                                                     43
            Weston County
                                   50
                                                       8
                                                             56
                                                                     45
                               CENSUS2010POP Estimates Base 2010 \
    STNAME CTYNAME
                                                             54571
    Alabama Autauga County
                                       54571
            Baldwin County
                                                            182265
                                      182265
            Barbour County
                                       27457
                                                             27457
            Bibb County
                                       22915
                                                             22919
            Blount County
                                       57322
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                                                               . . .
    Wyoming Sweetwater County
                                       43806
                                                             43806
            Teton County
                                       21294
                                                             21294
            Uinta County
                                       21118
                                                             21118
            Washakie County
                                       8533
                                                              8533
```

make it more readable. Note that instead of writing this all on one line, as ____

	Weston County	7208	72	08	
STNAME	CTYNAME	POPESTIMATE2010	POPESTIMATE2011	POPESTIMATE2012	2 \
	Autauga County	54660	55253	5517	5
	Baldwin County	183193	186659		
	Barbour County	27341	27226		
	Bibb County	22861	22733		
	Blount County	57373	57711		
	Diouno Councy				
Wwoming	Sweetwater County	43593	44041		
wyoming	Teton County	21297	21482		
	Uinta County	21102	20912		
	•	8545	8469		
	Washakie County				
	Weston County	7181	7114	7008	5
OTNI A ME	OTWNAME	RDOMESTICMI	G2011 RDOMESTIC	MIG2012 \	
STNAME	CTYNAME	7.0	40004	045007	
Alabama	Autauga County			.915927	
	Baldwin County			.647293	
	Barbour County			.500690	
	Bibb County			.068871	
	Blount County	1.8	07375 -1	.177622	
• • •		• • •		• • •	
Wyoming	Sweetwater County			.243199	
	Teton County			.972695	
	Uinta County	-17.7		.916350	
	Washakie County	-11.6		.827815	
	Weston County	-11.7	52361 -8	.040059	
STNAME	CTYNAME	RDOMESTICMIG2013	RDOMESTICMIG20	14 \	
	Autauga County	-3.012349	2.2659	71	
ATADAMA	Baldwin County	21.845705			
	•				
	Barbour County	-7.056824			
	Bibb County	-6.201001			
	Blount County	-1.748766			
 Umomina	Sweetwater County	-5.339774			
w y omiting	Teton County	19.525929			
	•				
	Uinta County	-6.902954 -2.013502			
	Washakie County	-2.013502			
	Weston County	12.372583	1.5336	3 0	
STNAME	CTYNAME	RDOMESTICMIG2015	RNETMIG2011 R	NETMIG2012 \	
	Autauga County	-2.530799	7.606016	-2.626146	

```
Baldwin County
                                 17.197872
                                             15.844176 18.559627
                              -10.543299
          Barbour County
                                             -4.874741 -2.758113
                                             -5.088389
          Bibb County
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          Blount County
                                 -1.369970
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   Wyoming Sweetwater County
                                -14.248864
                                              1.255221 16.243199
          Teton County
                                 -0.564849
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                                                         2.408578
                                 -12.127022 -18.136812 -5.536861
          Uinta County
          Washakie County
                                  1.682288 -11.990126
                                                         -1.182592
          Weston County
                                   6.935294 -12.032179
                                                         -8.040059
                            RNETMIG2013 RNETMIG2014 RNETMIG2015
   STNAME CTYNAME
                                         2.592270 -2.187333
   Alabama Autauga County
                            -2.722002
                            22.727626
          Baldwin County
                                         20.317142
                                                    18.293499
          Barbour County
                            -7.167664 -3.978583 -10.543299
                             -5.403729
                                                     1.107861
          Bibb County
                                         0.754533
          Blount County -1.402476 -1.577232 -0.884411
                                   . . .
   Wyoming Sweetwater County -5.295460 -14.075283 -14.070195
          Teton County
                            21.160658 16.308671
                                                     1.520747
                            -7.521840 -14.740608 -12.606351
          Uinta County
                                                    1.441961
          Washakie County
                             -2.250385 -18.020168
          Weston County
                            12.372583
                                         1.533635
                                                     6.935294
   [3142 rows x 98 columns]
[5]: # Now, the key with any good idiom is to understand when it isn't helping you.
    → In this case, you can actually
   # time both methods and see which one runs faster
   # We can put the approach into a function and pass the function into the timeit_{\sf L}
    →function to count the time the
   # parameter number allows us to choose how many times we want to run the
    → function. Here we will just set it to
   # 10
   # Lets write a wrapper for our first function
```

.rename(columns={'ESTIMATESBASE2010': 'Estimates Base 2010'}))

def first_approach():
 global df

And we'll just paste our code right here

.set_index(['STNAME','CTYNAME'])

return (df.where(df['SUMLEV']==50)

.dropna()

Read in our dataset anew

```
df = pd.read_csv('datasets/census.csv')
# And now lets run it
timeit.timeit(first_approach, number=10)
```

[5]: 1.1084833510685712

```
[6]: # Now let's test the second approach. As you may notice, we use our global
    →variable df in the function.
   # However, changing a global variable inside a function will modify the
    →variable even in a global scope and we
   # do not want that to happen in this case. Therefore, for selecting summary
    →levels of 50 only, I create a new
   # dataframe for those records
   # Let's run this for once and see how fast it is
   def second_approach():
       global df
       new_df = df[df['SUMLEV']==50]
       new_df.set_index(['STNAME','CTYNAME'], inplace=True)
       return new df.rename(columns={'ESTIMATESBASE2010': 'Estimates Base 2010'})
   # Read in our dataset anew
   df = pd.read_csv('datasets/census.csv')
   # And now lets run it
   timeit.timeit(second_approach, number=10)
```

[6]: 0.10386669298168272

```
[7]: # As you can see, the second approach is much faster! So, this is a particular ⇒ example of a classic time
# readability trade off.

# You'll see lots of examples on stack overflow and in documentation of people ⇒ using method chaining in their
# pandas. And so, I think being able to read and understand the syntax is ⇒ really worth your time. But keep in
# mind that following what appears to be stylistic idioms might have ⇒ performance issues that you need to
# consider as well.
```

```
[8]: # Here's another pandas idiom. Python has a wonderful function called map, which is sort of a basis for # functional programming in the language. When you want to use map in Python, you pass it some function you # want called, and some iterable, like a list, that you want the function to be applied to. The results are
```

```
# that the function is called against each item in the list, and there's a_{\sqcup}
      →resulting list of all of the
     # evaluations of that function.
     # Pandas has a similar function called applymap. In applymap, you provide some_
      → function which should operate
     # on each cell of a DataFrame, and the return set is itself a DataFrame. Now I_{f \sqcup}
     \hookrightarrow think applymap is fine, but I
     \# actually rarely use it. Instead, I find myself often wanting to map across_{\sqcup}
      →all of the rows in a DataFrame.
     # And pandas has a function that I use heavily there, called apply. Let's look_
     \rightarrowat an example.
 [9]: # Let's take a look at our census DataFrame. In this DataFrame, we have five
      ⇔columns for population estimates,
     # with each column corresponding with one year of estimates. It's quite,
      →reasonable to want to create some new
     # columns for minimum or maximum values, and the apply function is an easy way_{\sqcup}
     →to do this.
     # First, we need to write a function which takes in a particular row of data,
      → finds a minimum and maximum
     # values, and returns a new row of data nd returns a new row of data. We'll
      →call this function min_max, this
     # is pretty straight forward. We can create some small slice of a row by \Box
      →projecting the population columns.
     # Then use the NumPy min and max functions, and create a new series with a_{\sqcup}
     → label values represent the new
     # values we want to apply.
     def min_max(row):
         data = row[['POPESTIMATE2010',
                      'POPESTIMATE2011',
                      'POPESTIMATE2012'.
                      'POPESTIMATE2013'.
                      'POPESTIMATE2014',
                      'POPESTIMATE2015']]
         return pd.Series({'min': np.min(data), 'max': np.max(data)})
[10]: # Then we just need to call apply on the DataFrame.
     # Apply takes the function and the axis on which to operate as parameters. Now, __
      →we have to be a bit careful,
     # we've talked about axis zero being the rows of the DataFrame in the past. But \Box
      → this parameter is really the
     # parameter of the index to use. So, to apply across all rows, which is \Box
      →applying on all columns, you pass axis
```

```
# equal to 'columns'.
     df.apply(min_max, axis='columns').head()
[10]:
            min
                     max
       4785161
                4858979
     1
          54660
                   55347
     2
         183193
                  203709
     3
          26489
                   27341
          22512
                    22861
[11]: # Of course there's no need to limit yourself to returning a new series object.
      → If you're doing this as part
     # of data cleaning your likely to find yourself wanting to add new data to the
      →existing DataFrame. In that
     # case you just take the row values and add in new columns indicating the max_
      \hookrightarrow and minimum scores. This is a
     # regular part of my workflow when bringing in data and building summary or \Box
      \rightarrow descriptive statistics, and is
     # often used heavily with the merging of DataFrames.
[12]: # Here's an example where we have a revised version of the function min_max_1
      → Instead of returning a separate
     # series to display the min and max we add two new columns in the original \Box
      \rightarrow dataframe to store min and max
     def min_max(row):
         data = row[['POPESTIMATE2010',
                      'POPESTIMATE2011',
                      'POPESTIMATE2012',
                      'POPESTIMATE2013',
                      'POPESTIMATE2014',
                      'POPESTIMATE2015']]
         # Create a new entry for max
         row['max'] = np.max(data)
         # Create a new entry for min
         row['min'] = np.min(data)
     # Now just apply the function across the dataframe
     df.apply(min_max, axis='columns')
[12]:
           SUMLEV REGION DIVISION STATE
                                              COUNTY
                                                       STNAME
                                                                           CTYNAME \
                         3
                                                   0 Alabama
     0
               40
                                   6
                                           1
                                                                           Alabama
     1
               50
                         3
                                   6
                                           1
                                                   1 Alabama
                                                                   Autauga County
     2
               50
                         3
                                   6
                                                                   Baldwin County
                                           1
                                                   3 Alabama
                         3
     3
               50
                                   6
                                           1
                                                   5
                                                      Alabama
                                                                   Barbour County
               50
                         3
                                   6
                                           1
                                                   7
                                                      Alabama
                                                                      Bibb County
                                  . . .
                                         . . .
                         4
     3188
               50
                                   8
                                          56
                                                  37 Wyoming Sweetwater County
```

3189	50	4 8	56	39	Wyoming	Teton	County
3190	50	4 8		41	Wyoming		County
3191	50	4 8	56	43	Wyoming	Washakie	-
3192	50	4 8	56	45	Wyoming		County
					, 0		•
	CENSUS2010PO	P ESTIMATES	BASE2010	POPEST	ΓΙΜΑΤΕ2010	\	
0	477973	36	4780127		4785161		
1	5457	71	54571		54660		
2	18226	35	182265		183193		
3	2745	57	27457		27341		
4	2291	15	22919		22861		
	• •	•					
3188	4380)6	43806		43593		
3189	2129	94	21294		21297		
3190	2111	18	21118		21102		
3191	853	33	8533		8545		
3192	720)8	7208		7181		
	RDOMESTICMIC	32013 RDOMES	TICMIG2014	RDON	MESTICMIG20	15 RNETMI	G2011 \
0	0.38	31066	0.582002	?	-0.4673	69 1.0	30015
1	-3.01	12349	2.265971		-2.5307	99 7.6	06016
2	21.84	15705	19.243287	•	17.1978	72 15.8	44176
3	-7.05	56824	-3.904217	•	-10.5432	99 -4.8	74741
4	-6.20	1001	-0.177537	•	0.1772	58 -5.0	88389
3188	-5.33	39774	-14.252889)	-14.2488	64 1.2	55221
3189	19.52	25929	14.143021		-0.5648	49 0.6	54527
3190	-6.90)2954	-14.215862	?	-12.1270	22 -18.1	36812
3191	-2.01	13502	-17.781491		1.6822	88 -11.9	90126
3192	12.37	72583	1.533635	,	6.9352	94 -12.0	32179
	RNETMIG2012	RNETMIG2013	RNETMIG2	0014 E	RNETMIG2015	max	min
0	0.826644	1.383282			0.712594		4785161
1	-2.626146	-2.722002			-2.187333	55347	54660
2	18.559627	22.727626			18.293499		183193
3	-2.758113	-7.167664			-10.543299		26489
4	-4.363636	-5.403729			1.107861		22512
•	4.000000	0.400723	0.704		1.107001		
3188	16.243199	-5.295460	-14.075		-14.070195		43593
3189	2.408578	21.160658			1.520747		21297
3190	-5.536861	-7.521840			-12.606351	21102	20822
3191	-1.182592	-2.250385			1.441961		8316
3192	-8.040059	12.372583			6.935294		7065

[3193 rows x 102 columns]

```
[13]: # Apply is an extremely important tool in your toolkit. The reason I introduced.
     →apply here is because you
     # rarely see it used with large function definitions, like we did. Instead, you
     →typically see it used with
     # lambdas. To get the most of the discussions you'll see online, you're going_
     →to need to know how to at least
     # read lambdas.
     # Here's You can imagine how you might chain several apply calls with lambdasu
     →together to create a readable
     # yet succinct data manipulation script. One line example of how you might_
     ⇔calculate the max of the columns
     # using the apply function.
     rows = ['POPESTIMATE2010', 'POPESTIMATE2011', 'POPESTIMATE2012',
     → 'POPESTIMATE2013', 'POPESTIMATE2014',
             'POPESTIMATE2015']
     # Now we'll just apply this across the dataframe with a lambda
     df.apply(lambda x: np.max(x[rows]), axis=1).head()
[13]: 0
         4858979
    1
            55347
    2
           203709
           27341
    3
    4
            22861
    dtype: int64
[14]: # If you don't remember lambdas just pause the video for a moment and look up_{\sqcup}
     →the syntax. A lambda is just an
     # unnamed function in python, in this case it takes a single parameter, x, and
     →returns a single value, in this
     \# case the maximum over all columns associated with row x.
[15]: # The beauty of the apply function is that it allows flexibility in doing.
      →whatever manipulation that you
     # desire, as the function you pass into apply can be any customized however you\square
     →want. Let's say we want to
     # divide the states into four categories: Northeast, Midwest, South, and West_{\sqcup}
     →We can write a customized
     # function that returns the region based on the state the state regions \Box
     →information is obtained from Wikipedia
     def get_state_region(x):
         northeast = ['Connecticut', 'Maine', 'Massachusetts', 'New Hampshire',
                      'Rhode Island', 'Vermont', 'New York', 'New_
      →Jersey', 'Pennsylvania']
         midwest = ['Illinois','Indiana','Michigan','Ohio','Wisconsin','Iowa',
                    'Kansas', 'Minnesota', 'Missouri', 'Nebraska', 'North Dakota',
```

```
'South Dakota']
         south = ['Delaware', 'Florida', 'Georgia', 'Maryland', 'North Carolina',
                  'South Carolina', 'Virginia', 'District of Columbia', 'West Virginia',
                  'Alabama', 'Kentucky', 'Mississippi', 'Tennessee', 'Arkansas',
                  'Louisiana','Oklahoma','Texas']
         west = ['Arizona','Colorado','Idaho','Montana','Nevada','New Mexico','Utah',
                 'Wyoming','Alaska','California','Hawaii','Oregon','Washington']
         if x in northeast:
             return "Northeast"
         elif x in midwest:
             return "Midwest"
         elif x in south:
             return "South"
         else:
             return "West"
[16]: # Now we have the customized function, let's say we want to create a new column
      →called Region, which shows the
     # state's region, we can use the customized function and the apply function to \sqcup
     \rightarrow do so. The customized function
     # is supposed to work on the state name column STNAME. So we will set the apply_
     → function on the state name
     # column and pass the customized function into the apply function
     df['state_region'] = df['STNAME'].apply(lambda x: get_state_region(x))
[17]: # Now let's see the results
     df[['STNAME','state_region']].head()
         STNAME state_region
[17]:
     0 Alabama
                       South
     1 Alabama
                       South
     2 Alabama
                       South
     3 Alabama
                       South
     4 Alabama
                       South
```

So there are a couple of Pandas idioms. But I think there's many more, and I haven't talked about them here. So here's an unofficial assignment for you. Go look at some of the top ranked questions on pandas on Stack Overflow, and look at how some of the more experienced authors, answer those questions. Do you see any interesting patterns? Feel free to share them with myself and others in the class.

GroupBy_ed

November 15, 2021

Sometimes we want to select data based on groups and understand aggregated data on a group level. We have seen that even though Pandas allows us to iterate over every row in a dataframe, it is geneally very slow to do so. Fortunately Pandas has a groupby() function to speed up such task. The idea behind the groupby() function is that it takes some dataframe, splits it into chunks based on some key values, applies computation on those chunks, then combines the results back together into another dataframe. In pandas this is referred to as the split-apply-combine pattern.

1 Splitting

```
[1]: # Let's look at an example. First, we'll bring in our pandas and numpy_
     \rightarrow libraries
    import pandas as pd
    import numpy as np
[2]: # Let's look at some US census data
    df = pd.read_csv('datasets/census.csv')
    # And exclude state level summarizations, which have sum level value of 40
    df = df[df['SUMLEV']==50]
    df.head()
               REGION DIVISION STATE
[2]:
       SUMLEV
                                        COUNTY
                                                  STNAME
                                                                  CTYNAME
           50
                    3
                               6
                                      1
                                              1 Alabama Autauga County
   1
   2
           50
                    3
                                              3 Alabama Baldwin County
                                      1
   3
                    3
                               6
                                              5 Alabama Barbour County
           50
                                      1
   4
                    3
                               6
                                              7
           50
                                      1
                                                 Alabama
                                                              Bibb County
                                                 Alabama
   5
           50
                                      1
                                                            Blount County
       CENSUS2010POP ESTIMATESBASE2010 POPESTIMATE2010
                                                                 RDOMESTICMIG2011
   1
               54571
                                   54571
                                                     54660
                                                                         7.242091
   2
              182265
                                  182265
                                                    183193
                                                                        14.832960
    3
               27457
                                   27457
                                                     27341
                                                                        -4.728132
   4
               22915
                                   22919
                                                     22861
                                                                        -5.527043
   5
               57322
                                   57322
                                                     57373
                                                                          1.807375
       RDOMESTICMIG2012 RDOMESTICMIG2013 RDOMESTICMIG2014 RDOMESTICMIG2015
                                 -3.012349
   1
              -2.915927
                                                     2.265971
                                                                      -2.530799
   2
              17.647293
                                 21.845705
                                                    19.243287
                                                                      17.197872
```

```
3
             -2.500690
                               -7.056824
                                                 -3.904217
                                                                  -10.543299
   4
             -5.068871
                               -6.201001
                                                 -0.177537
                                                                    0.177258
                               -1.748766
             -1.177622
                                                 -2.062535
                                                                   -1.369970
      RNETMIG2011 RNETMIG2012 RNETMIG2013 RNETMIG2014 RNETMIG2015
   1
         7.606016
                     -2.626146
                                  -2.722002
                                                2.592270
                                                            -2.187333
   2
        15.844176
                    18.559627
                                  22.727626
                                               20.317142 18.293499
   3
        -4.874741
                     -2.758113 -7.167664
                                               -3.978583 -10.543299
   4
        -5.088389
                     -4.363636
                                  -5.403729
                                                0.754533
                                                            1.107861
         1.859511
                     -0.848580
                                  -1.402476
                                               -1.577232
                                                            -0.884411
    [5 rows x 100 columns]
[3]: # In the first example for groupby() I want to use the census date. Let's get a_{\perp}
    → list of the unique states,
    # then we can iterate over all the states and for each state we reduce the \mathrm{data}_{\sqcup}
    → frame and calculate the
    # average.
    \# Let's run such task for 3 times and time it. For this we'll use the cellu
    → magic function %%timeit
[4]: %%timeit -n 3
    for state in df['STNAME'].unique():
        # We'll just calculate the average using numpy for this particular state
```

avg = np.average(df.where(df['STNAME']==state).dropna()['CENSUS2010POP'])

' have an average population of ' + str(avg))

And we'll print it to the screen
print('Counties in state ' + state +

```
Counties in state Alabama have an average population of 71339.34328358209
Counties in state Alaska have an average population of 24490.724137931036
Counties in state Arizona have an average population of 426134.4666666667
Counties in state Arkansas have an average population of 38878.906666666667
Counties in state California have an average population of 642309.5862068966
Counties in state Colorado have an average population of 78581.1875
Counties in state Connecticut have an average population of 446762.125
Counties in state Delaware have an average population of 299311.3333333333
Counties in state District of Columbia have an average population of 601723.0
Counties in state Florida have an average population of 280616.5671641791
Counties in state Georgia have an average population of 60928.63522012578
Counties in state Hawaii have an average population of 272060.2
Counties in state Idaho have an average population of 35626.86363636364
Counties in state Illinois have an average population of 125790.50980392157
Counties in state Indiana have an average population of 70476.10869565218
Counties in state Iowa have an average population of 30771.262626262625
Counties in state Kansas have an average population of 27172.55238095238
```

```
Counties in state Kentucky have an average population of 36161.39166666667
Counties in state Louisiana have an average population of 70833.9375
Counties in state Maine have an average population of 83022.5625
Counties in state Maryland have an average population of 240564.6666666666
Counties in state Massachusetts have an average population of 467687.78571428574
Counties in state Michigan have an average population of 119080.0
Counties in state Minnesota have an average population of 60964.65517241379
Counties in state Mississippi have an average population of 36186.54878048781
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Counties in state Nebraska have an average population of 19638.075268817203
Counties in state Nevada have an average population of 158855.9411764706
Counties in state New Hampshire have an average population of 131647.0
Counties in state New Jersey have an average population of 418661.61904761905
Counties in state New Mexico have an average population of 62399.36363636364
Counties in state New York have an average population of 312550.03225806454
Counties in state North Carolina have an average population of 95354.83
Counties in state North Dakota have an average population of 12690.396226415094
Counties in state Ohio have an average population of 131096.63636363635
Counties in state Oklahoma have an average population of 48718.844155844155
Counties in state Oregon have an average population of 106418.72222222222
Counties in state Pennsylvania have an average population of 189587.74626865672
Counties in state Rhode Island have an average population of 210513.4
Counties in state South Carolina have an average population of
100551.39130434782
Counties in state South Dakota have an average population of 12336.060606060606
Counties in state Tennessee have an average population of 66801.1052631579
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Counties in state New Mexico have an average population of 62399.36363636364
Counties in state New York have an average population of 312550.03225806454
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Counties in state North Carolina have an average population of 95354.83
Counties in state North Dakota have an average population of 12690.396226415094
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100551.39130434782
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   4.62 s ś 230 ms per loop (mean ś std. dev. of 7 runs, 3 loops each)
[5]: # If you scroll down to the bottom of that output you can see it takes a fair.
    \rightarrow bit of time to finish.
    # Now let's try another approach using groupby()
[6]: | %%timeit -n 3
    # For this method, we start by telling pandas we're interested in grouping by \Box
    \hookrightarrowstate name, this is the "split"
    for group, frame in df.groupby('STNAME'):
        # You'll notice there are two values we set here. groupby() returns a_
     \rightarrow tuple, where the first value is the
        # value of the key we were trying to group by, in this case a specificu
     ⇒state name, and the second one is
        # projected dataframe that was found for that group
        # Now we include our logic in the "apply" step, which is to calculate an
     →average of the census2010pop
        avg = np.average(frame['CENSUS2010POP'])
        # And print the results
        print('Counties in state ' + group +
              ' have an average population of ' + str(avg))
```

Counties in state Alabama have an average population of 71339.34328358209 Counties in state Alaska have an average population of 24490.724137931036

→of our data transformation is # actually printing out results.

And we don't have to worry about the combine step in this case, because all \sqcup

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100551.39130434782
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```

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Counties in state South Dakota have an average population of 12336.060606060606
Counties in state Tennessee have an average population of 66801.1052631579
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Counties in state Utah have an average population of 95306.37931034483
Counties in state Vermont have an average population of 44695.78571428572
Counties in state Virginia have an average population of 60111.29323308271
Counties in state Washington have an average population of 172424.10256410256
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Counties in state Montana have an average population of 17668.125
Counties in state Nebraska have an average population of 19638.075268817203
Counties in state Nevada have an average population of 158855.9411764706
Counties in state New Hampshire have an average population of 131647.0
Counties in state New Jersey have an average population of 418661.61904761905
Counties in state New Mexico have an average population of 62399.36363636364
Counties in state New York have an average population of 312550.03225806454
Counties in state North Carolina have an average population of 95354.83
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29.6 ms ś 3.65 ms per loop (mean ś std. dev. of 7 runs, 3 loops each)
```

- [7]: # Wow, what a huge difference in speed. An improve by roughly by two factors!

```
→ that in order to do this you need
     # to set the index of the data frame to be the column that you want to group by
      \hookrightarrow first.
     # We'll create some new function called set_batch_number and if the first_{\sqcup}
     → letter of the parameter is a capital
     # M we'll return a 0. If it's a capital Q we'll return a 1 and otherwise we'll \Box
      →return a 2. Then we'll pass
     # this function to the data frame
     df = df.set_index('STNAME')
     def set_batch_number(item):
         if item[0]<'M':</pre>
             return 0
         if item[0]<'Q':</pre>
             return 1
         return 2
     # The dataframe is supposed to be grouped by according to the batch number Andu
      →we will loop through each batch
     # group
     for group, frame in df.groupby(set_batch_number):
         print('There are ' + str(len(frame)) + ' records in group ' + str(group) + L
      There are 1177 records in group 0 for processing.
    There are 1134 records in group 1 for processing.
    There are 831 records in group 2 for processing.
 [9]: # Notice that this time I didn't pass in a column name to groupby(). Instead, I_{\sqcup}
      \rightarrowset the index of the dataframe
     # to be STNAME, and if no column identifier is passed groupby() will_
     →automatically use the index.
[10]: # Let's take one more look at an example of how we might group data. In this
      \rightarrow example, I want to use a dataset
     # of housing from airbnb. In this dataset there are two columns of interest, __
     → one is the cancellation_policy
     # and the other is the review scores value.
     df=pd.read_csv("datasets/listings.csv")
     df.head()
[10]:
              id
                                             listing url
                                                                scrape_id \
    0 12147973 https://www.airbnb.com/rooms/12147973 20160906204935
         3075044
                  https://www.airbnb.com/rooms/3075044 20160906204935
     1
     2
            6976
                      https://www.airbnb.com/rooms/6976 20160906204935
```

by to use this function to split up our data frame. It's important to note,

```
https://www.airbnb.com/rooms/1436513 20160906204935
3
    1436513
    7651065
              https://www.airbnb.com/rooms/7651065 20160906204935
  last_scraped
0
    2016-09-07
                                   Sunny Bungalow in the City
    2016-09-07
                            Charming room in pet friendly apt
1
2
    2016-09-07
                             Mexican Folk Art Haven in Boston
3
    2016-09-07
                Spacious Sunny Bedroom Suite in Historic Home
    2016-09-07
                                          Come Home to Boston
                                              summary
O Cozy, sunny, family home. Master bedroom high...
1 Charming and quiet room in a second floor 1910...
2 Come stay with a friendly, middle-aged guy in ...
3 Come experience the comforts of home away from...
4 My comfy, clean and relaxing home is one block...
                                                space
O The house has an open and cozy feel at the sam...
1 Small but cozy and quite room with a full size...
2 Come stay with a friendly, middle-aged guy in ...
3 Most places you find in Boston are small howev...
4 Clean, attractive, private room, one block fro...
                                          description experiences_offered
O Cozy, sunny, family home. Master bedroom high...
                                                                     none
1 Charming and quiet room in a second floor 1910...
                                                                     none
2 Come stay with a friendly, middle-aged guy in ...
                                                                     none
3 Come experience the comforts of home away from...
                                                                     none
4 My comfy, clean and relaxing home is one block...
                                                                     none
                               neighborhood_overview ... review_scores_value
O Roslindale is quiet, convenient and friendly. ...
                                                                           NaN
1 The room is in Roslindale, a diverse and prima...
                                                                           9.0
2 The LOCATION: Roslindale is a safe and diverse...
                                                                          10.0
3 Roslindale is a lovely little neighborhood loc...
                                                                          10.0
4 I love the proximity to downtown, the neighbor...
                                                                          10.0
  requires_license license jurisdiction_names instant_bookable
0
                       NaN
                 f
                                          NaN
                 f
                       NaN
1
                                          NaN
                                                              t
                                                              f
2
                 f
                       NaN
                                          NaN
3
                 f
                       NaN
                                          NaN
                                                              f
4
                 f
                       \mathtt{NaN}
                                          \mathtt{NaN}
                                                              f
  cancellation_policy require_guest_profile_picture
0
             moderate
```

```
2
                 moderate
                                                     t
                                                     f
    3
                 moderate
                                                     f
    4
                 flexible
      require_guest_phone_verification calculated_host_listings_count
                                     f
    1
                                                                   1
                                     f
    2
                                                                   1
    3
                                     f
                                                                   1
    4
                                     f
                                                                   1
       reviews_per_month
    0
                     \mathtt{NaN}
                    1.30
    1
                    0.47
    2
    3
                    1.00
    4
                    2.25
    [5 rows x 95 columns]
[11]: \# So, how would I group by both of these columns? A first approach might be to
     \rightarrowpromote them to a multiindex
    # and just call groupby()
    df=df.set_index(["cancellation_policy", "review_scores_value"])
    \rightarrow grouping by
    for group, frame in df.groupby(level=(0,1)):
        print(group)
    ('flexible', 2.0)
    ('flexible', 4.0)
    ('flexible', 5.0)
    ('flexible', 6.0)
    ('flexible', 7.0)
    ('flexible', 8.0)
    ('flexible', 9.0)
    ('flexible', 10.0)
    ('moderate', 2.0)
    ('moderate', 4.0)
    ('moderate', 6.0)
    ('moderate', 7.0)
    ('moderate', 8.0)
    ('moderate', 9.0)
    ('moderate', 10.0)
    ('strict', 2.0)
    ('strict', 3.0)
```

1

moderate

f

```
('strict', 4.0)
    ('strict', 5.0)
    ('strict', 6.0)
    ('strict', 7.0)
    ('strict', 8.0)
    ('strict', 9.0)
    ('strict', 10.0)
    ('super_strict_30', 6.0)
    ('super_strict_30', 7.0)
    ('super_strict_30', 8.0)
    ('super_strict_30', 9.0)
    ('super_strict_30', 10.0)
[12]: # This seems to work ok. But what if we wanted to group by the cancelation \Box
     →policy and review scores, but
     # separate out all the 10's from those under ten? In this case, we could use a_{\sqcup}
     → function to manage the
     # groupings
     def grouping_fun(item):
         # Check the "review_scores_value" portion of the index. item is in the \Box
      \rightarrow format of
         # (cancellation_policy, review_scores_value
         if item[1] == 10.0:
             return (item[0],"10.0")
             return (item[0], "not 10.0")
     for group, frame in df.groupby(by=grouping_fun):
         print(group)
    ('flexible', '10.0')
    ('flexible', 'not 10.0')
    ('moderate', '10.0')
    ('moderate', 'not 10.0')
    ('strict', '10.0')
    ('strict', 'not 10.0')
    ('super_strict_30', '10.0')
    ('super_strict_30', 'not 10.0')
[13]: df.head()
[13]:
                                                      id \
     cancellation_policy review_scores_value
    moderate
                                                12147973
                          NaN
                                                 3075044
                          9.0
                          10.0
                                                    6976
                          10.0
                                                 1436513
```

flexible	10.0	7651065
\		listing_url
•	review_scores_value NaN 9.0 10.0 10.0	https://www.airbnb.com/rooms/12147973 https://www.airbnb.com/rooms/3075044 https://www.airbnb.com/rooms/6976 https://www.airbnb.com/rooms/1436513
flexible	10.0	https://www.airbnb.com/rooms/7651065
<pre>cancellation_policy moderate flexible</pre>	review_scores_value NaN 9.0 10.0 10.0 10.0	scrape_id last_scraped \ 20160906204935
name \	10.0	20100000201000 2010 00 01
cancellation_policy moderate the City	review_scores_value NaN	Sunny Bungalow in
friendly apt	9.0	Charming room in pet
Boston	10.0	Mexican Folk Art Haven in
Historic Home flexible Boston	10.0	Spacious Sunny Bedroom Suite in Come Home to
<pre>summary \ cancellation_policy moderate bedroom high</pre>	review_scores_value	Cozy, sunny, family home. Master
floor 1910	9.0	Charming and quiet room in a second
guy in	10.0	Come stay with a friendly, middle-aged
away from flexible one block	10.0	Come experience the comforts of home My comfy, clean and relaxing home is
<pre>space \ cancellation_policy</pre>	review_scores_value	

moderate	NaN	The house has an open and cozy feel at			
the sam					
	9.0	Small but cozy and quite room with a			
full size					
	10.0	Come stay with a friendly, middle-aged			
guy in					
	10.0	Most places you find in Boston are			
small howev					
flexible	10.0	Clean, attractive, private room, one			
block fro					
description \					
	review_scores_value				
moderate	NaN	Cozy, sunny, family home. Master			
bedroom high					
	9.0	Charming and quiet room in a second			
floor 1910					
	10.0	Come stay with a friendly, middle-aged			
guy in					
	10.0	Come experience the comforts of home			
away from					
flexible	10.0	My comfy, clean and relaxing home is			
one block					
		experiences_offered \			
	review_scores_value	experiences_offered \			
cancellation_policy		experiences_offered \ none			
	review_scores_value	•			
	review_scores_value NaN	none			
	review_scores_value NaN 9.0	none none			
	review_scores_value NaN 9.0 10.0	none none none			
moderate	review_scores_value NaN 9.0 10.0 10.0	none none none none			
moderate flexible neighborhood_overvi	review_scores_value NaN 9.0 10.0 10.0 10.0	none none none none			
moderate flexible neighborhood_overvi	review_scores_value NaN 9.0 10.0 10.0	none none none none			
moderate flexible neighborhood_overvi	review_scores_value NaN 9.0 10.0 10.0 10.0	none none none none			
moderate flexible neighborhood_overvicancellation_policy	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value	none none none none none			
moderate flexible neighborhood_overvicancellation_policy moderate	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value	none none none none none			
moderate flexible neighborhood_overvicancellation_policy moderate	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN	none none none none none Roslindale is quiet, convenient and			
moderate flexible neighborhood_overvicancellation_policy moderate friendly	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN	none none none none none Roslindale is quiet, convenient and			
moderate flexible neighborhood_overvicancellation_policy moderate friendly	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN 9.0	none none none none none none Roslindale is quiet, convenient and The room is in Roslindale, a diverse			
flexible neighborhood_overvicancellation_policy moderate friendly and prima	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN 9.0	none none none none none none Roslindale is quiet, convenient and The room is in Roslindale, a diverse			
flexible neighborhood_overvicancellation_policy moderate friendly and prima	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN 9.0 10.0	none none none none none none The room is in Roslindale, a diverse The LOCATION: Roslindale is a safe and			
flexible neighborhood_overvicancellation_policy moderate friendly and prima diverse	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN 9.0 10.0	none none none none none none The room is in Roslindale, a diverse The LOCATION: Roslindale is a safe and			
flexible neighborhood_overvicancellation_policy moderate friendly and prima diverse neighborhood loc	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN 9.0 10.0	none none none none none none Roslindale is quiet, convenient and The room is in Roslindale, a diverse The LOCATION: Roslindale is a safe and Roslindale is a lovely little			
flexible neighborhood_overvicancellation_policy moderate friendly and prima diverse neighborhood loc flexible	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN 9.0 10.0	none none none none none none Roslindale is quiet, convenient and The room is in Roslindale, a diverse The LOCATION: Roslindale is a safe and Roslindale is a lovely little			
flexible neighborhood_overvicancellation_policy moderate friendly and prima diverse neighborhood loc flexible	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN 9.0 10.0	none none none none none none Roslindale is quiet, convenient and The room is in Roslindale, a diverse The LOCATION: Roslindale is a safe and Roslindale is a lovely little			
flexible neighborhood_overvicancellation_policy moderate friendly and prima diverse neighborhood loc flexible neighbor	review_scores_value NaN 9.0 10.0 10.0 10.0 ew \ review_scores_value NaN 9.0 10.0	none none none none none none Roslindale is quiet, convenient and The room is in Roslindale, a diverse The LOCATION: Roslindale is a safe and Roslindale is a lovely little I love the proximity to downtown, the			

```
moderate
                     {\tt NaN}
                                                                         NaN
                                            . . .
                     9.0
                                                                        10.0
                     10.0
                                                                        10.0
                     10.0
                                                                        10.0
                                            . . .
                                                                        10.0
                     10.0
flexible
                                          review_scores_location \
cancellation_policy review_scores_value
                     NaN
moderate
                                                              NaN
                     9.0
                                                              9.0
                     10.0
                                                              9.0
                     10.0
                                                              10.0
flexible
                     10.0
                                                              9.0
                                          requires_license license \
cancellation_policy review_scores_value
                                                          f
                                                                 NaN
moderate
                     NaN
                     9.0
                                                          f
                                                                 NaN
                     10.0
                                                                NaN
                                                          f
                     10.0
                                                          f
                                                                 NaN
flexible
                     10.0
                                                          f
                                                                 NaN
                                          jurisdiction_names instant_bookable \
cancellation_policy review_scores_value
moderate
                     NaN
                                                          NaN
                                                                              f
                     9.0
                                                          NaN
                                                                              t
                     10.0
                                                          NaN
                                                                              f
                     10.0
                                                          NaN
                                                                              f
flexible
                     10.0
                                                          NaN
                                                                              f
                                          require_guest_profile_picture \
cancellation_policy review_scores_value
moderate
                     NaN
                                                                        f
                     9.0
                                                                        f
                     10.0
                                                                        t
                     10.0
                                                                        f
flexible
                     10.0
                                                                        f
                                          require_guest_phone_verification \
cancellation_policy review_scores_value
                                                                           f
moderate
                     NaN
                     9.0
                                                                           f
                     10.0
                                                                           f
                     10.0
                                                                           f
flexible
                     10.0
                                                                           f
                                          calculated_host_listings_count \
```

```
cancellation_policy review_scores_value
moderate
                     NaN
                                                                          1
                     9.0
                                                                          1
                     10.0
                                                                          1
                     10.0
                                                                          1
flexible
                     10.0
                                                                          1
                                            reviews_per_month
cancellation_policy review_scores_value
moderate
                     NaN
                                                          NaN
                                                          1.30
                     9.0
                     10.0
                                                          0.47
                     10.0
                                                          1.00
```

[5 rows x 93 columns]

10.0

2 Applying

flexible

```
[14]: # To this point we have applied very simple processing to our data after

⇒splitting, really just outputting

# some print statements to demonstrate how the splitting works. The pandas

⇒developers have three broad

# categories of data processing to happen during the apply step, Aggregation of

⇒group data, Transformation of

# group data, and Filtration of group data
```

2.25

2.1 Aggregation

```
cancellation_policy
    flexible
                                          NaN
    moderate
                                          NaN
    strict
                                          NaN
    super_strict_30
                                          NaN
[16]: # Hrm. That didn't seem to work at all. Just a bunch of not a numbers. The
     → issue is actually in the function
     # that we sent to aggregate. np.average does not ignore nans! However, there is __
     →a function we can use for this
     df.groupby("cancellation_policy").agg({"review_scores_value":np.nanmean})
[16]:
                          review_scores_value
    cancellation_policy
    flexible
                                     9.237421
    moderate
                                     9.307398
    strict
                                     9.081441
    super_strict_30
                                     8.537313
[17]: # We can just extend this dictionary to aggregate by multiple functions or
     → multiple columns.
     df.groupby("cancellation_policy").agg({"review_scores_value":(np.nanmean,np.
      →nanstd),
                                            "reviews_per_month":np.nanmean})
[17]:
                         review_scores_value
                                                        reviews_per_month
                                     nanmean
                                                nanstd
                                                                  nanmean
     cancellation_policy
                                    9.237421 1.096271
                                                                 1.829210
    flexible
                                    9.307398 0.859859
                                                                 2.391922
    moderate
                                    9.081441 1.040531
                                                                 1.873467
    strict
                                    8.537313 0.840785
                                                                 0.340143
    super_strict_30
[18]: # Take a moment to make sure you understand the previous cell, since it's
     →somewhat complex. First we're doing
     # a group by on the dataframe object by the column "cancellation_policy". This __
     →creates a new GroupBy object.
     # Then we are invoking the agg() function on that object. The agg function is \Box
     → qoing to apply one or more
     # functions we specify to the group dataframes and return a single row peru
      → dataframe/group. When we called
     # this function we sent it two dictionary entries, each with the key indicating
     →which column we wanted
     # functions applied to. For the first column we actually supplied a tuple of \Box
     \rightarrow two functions. Note that these
     # are not function invocations, like np.nanmean(), or function names, like_
     → "nanmean" they are references to
```

review_scores_value

[15]:

```
# functions which will return single values. The groupby object will recognize the tuple and call each
# function in order on the same column. The results will be in a heirarchical index, but since they are
# columns they don't show as an index per se. Then we indicated another column and a single function we wanted
# to run.
```

2.2 Transformation

```
[19]: # Transformation is different from aggregation. Where agg() returns a single_
     →value per column, so one row per
     # group, tranform() returns an object that is the same size as the group.
     →Essentially, it broadcasts the
     # function you supply over the grouped dataframe, returning a new dataframe.
     → This makes combining data later
     # easy.
[20]: # For instance, suppose we want to include the average rating values in a given
     → group by cancellation policy,
     # but preserve the dataframe shape so that we could generate a difference_
     →between an individual observation
     # and the sum.
     # First, lets define just some subset of columns we are interested in
     cols=['cancellation_policy','review_scores_value']
     # Now lets transform it, I'll store this in its own dataframe
     transform_df=df[cols].groupby('cancellation_policy').transform(np.nanmean)
    transform_df.head()
[20]:
       review_scores_value
                  9.307398
                  9.307398
    1
    2
                  9.307398
                  9.307398
    3
                  9.237421
[21]: # So we can see that the index here is actually the same as the original
     → dataframe. So lets just join this
     # in. Before we do that, lets rename the column in the transformed version
     transform_df.rename({'review_scores_value':'mean_review_scores'},_
      →axis='columns', inplace=True)
     df=df.merge(transform_df, left_index=True, right_index=True)
     df.head()
                                                       id \
```

```
2
             moderate
                                      10.0
                                                 6976
3
             moderate
                                      10.0
                                              1436513
4
             flexible
                                      10.0
                                              7651065
                             listing_url
                                                scrape_id last_scraped
   https://www.airbnb.com/rooms/12147973
                                          20160906204935
                                                            2016-09-07
0
    https://www.airbnb.com/rooms/3075044
                                           20160906204935
                                                            2016-09-07
1
2
       https://www.airbnb.com/rooms/6976
                                           20160906204935
                                                            2016-09-07
3
    https://www.airbnb.com/rooms/1436513
                                           20160906204935
                                                            2016-09-07
    https://www.airbnb.com/rooms/7651065
                                           20160906204935
                                                            2016-09-07
                                             name
0
                      Sunny Bungalow in the City
1
               Charming room in pet friendly apt
2
                Mexican Folk Art Haven in Boston
3
   Spacious Sunny Bedroom Suite in Historic Home
                             Come Home to Boston
                                              summary \
O Cozy, sunny, family home. Master bedroom high...
1 Charming and quiet room in a second floor 1910...
2 Come stay with a friendly, middle-aged guy in ...
3 Come experience the comforts of home away from...
4 My comfy, clean and relaxing home is one block...
                                                space
 The house has an open and cozy feel at the sam...
1 Small but cozy and quite room with a full size...
2 Come stay with a friendly, middle-aged guy in ...
3 Most places you find in Boston are small howev...
4 Clean, attractive, private room, one block fro...
                                          description
O Cozy, sunny, family home. Master bedroom high...
1 Charming and quiet room in a second floor 1910...
2 Come stay with a friendly, middle-aged guy in ...
3 Come experience the comforts of home away from...
4 My comfy, clean and relaxing home is one block...
  review_scores_location requires_license license jurisdiction_names
0
                     NaN
                                         f
                                               NaN
                                                                  NaN
1
                     9.0
                                         f
                                               NaN
                                                                  NaN
2
                     9.0
                                         f
                                               NaN
                                                                  NaN
3
                    10.0
                                         f
                                               NaN
                                                                  NaN
4
                     9.0
                                         f
                                              NaN
                                                                  NaN
```

instant_bookable require_guest_profile_picture \

```
1
                                                        f
                       t
     2
                       f
                                                        t
     3
                       f
                                                        f
     4
                                                        f
                       f
       require_guest_phone_verification calculated_host_listings_count
     0
                                         f
     1
                                         f
                                                                           1
     2
                                         f
                                                                           1
     3
                                         f
                                                                           1
     4
                                         f
                                                                           1
       reviews_per_month mean_review_scores
     0
                      {\tt NaN}
                                      9.307398
     1
                     1.30
                                      9.307398
     2
                     0.47
                                      9.307398
     3
                     1.00
                                      9.307398
     4
                     2.25
                                      9.237421
     [5 rows x 96 columns]
[22]: # Great, we can see that our new column is in place, the mean_review_scores. Sou
      →now we could create, for
     # instance, the difference between a given row and it's group (the cancellation \Box
      \rightarrowpolicy) means.
     df['mean_diff'] = np.absolute(df['review_scores_value'] - df['mean_review_scores'])
     df['mean_diff'].head()
[22]: 0
                NaN
          0.307398
     1
```

f

2.3 Filtering

0.692602 0.692602 0.762579

Name: mean_diff, dtype: float64

2

0

f

```
[23]: # The GroupBy object has build in support for filtering groups as well. It's

→ often that you'll want to group

# by some feature, then make some transformation to the groups, then drop

→ certain groups as part of your

# cleaning routines. The filter() function takes in a function which it applies

→ to each group dataframe and

# returns either a True or a False, depending upon whether that group should be

→ included in the results.
```

```
[24]: # For instance, if we only want those groups which have a mean rating above 9_{\sqcup}
      →included in our results
     df.groupby('cancellation_policy').filter(lambda x: np.
      →nanmean(x['review_scores_value'])>9.2)
          cancellation_policy review_scores_value
[24]:
                                                             id
     0
                     moderate
                                                 NaN
                                                       12147973
     1
                     moderate
                                                 9.0
                                                        3075044
     2
                      moderate
                                                10.0
                                                           6976
     3
                      moderate
                                                10.0
                                                        1436513
                      flexible
                                                10.0
                                                        7651065
                           . . .
                                                 . . .
                                                            . . .
     . . .
     3576
                      flexible
                                                 {\tt NaN}
                                                      14689681
     3577
                      flexible
                                                      13750763
                                                 {\tt NaN}
                      flexible
                                                      14852179
     3579
                                                 \mathtt{NaN}
                      flexible
                                                      14585486
     3582
                                                 NaN
                      flexible
                                                      14504422
     3584
                                                 {\tt NaN}
                                      listing_url
                                                          scrape_id last_scraped
     0
           https://www.airbnb.com/rooms/12147973
                                                     20160906204935
                                                                       2016-09-07
            https://www.airbnb.com/rooms/3075044
     1
                                                     20160906204935
                                                                       2016-09-07
     2
               https://www.airbnb.com/rooms/6976
                                                     20160906204935
                                                                       2016-09-07
     3
            https://www.airbnb.com/rooms/1436513
                                                     20160906204935
                                                                       2016-09-07
     4
            https://www.airbnb.com/rooms/7651065
                                                     20160906204935
                                                                       2016-09-07
     3576 https://www.airbnb.com/rooms/14689681
                                                     20160906204935
                                                                       2016-09-07
           https://www.airbnb.com/rooms/13750763
     3577
                                                     20160906204935
                                                                       2016-09-07
           https://www.airbnb.com/rooms/14852179
     3579
                                                     20160906204935
                                                                       2016-09-07
           https://www.airbnb.com/rooms/14585486
     3582
                                                     20160906204935
                                                                       2016-09-07
     3584
           https://www.airbnb.com/rooms/14504422
                                                     20160906204935
                                                                       2016-09-07
                                                           name
     0
                                   Sunny Bungalow in the City
     1
                            Charming room in pet friendly apt
     2
                             Mexican Folk Art Haven in Boston
     3
               Spacious Sunny Bedroom Suite in Historic Home
     4
                                           Come Home to Boston
            Beautiful loft style bedroom with large bathroom
     3576
                 Comfortable Space in the Heart of Brookline
     3577
           Spacious Queen Bed Room Close to Boston Univer...
     3579
     3582
                                      Gorgeous funky apartment
     3584
                           (K1) Private Room near Harvard/MIT
                                                        summary
     0
           Cozy, sunny, family home. Master bedroom high...
     1
           Charming and quiet room in a second floor 1910...
```

```
2
      Come stay with a friendly, middle-aged guy in ...
3
      Come experience the comforts of home away from...
4
      My comfy, clean and relaxing home is one block...
3576 You'd be living on the top floor of a four sto...
     Our place is close to Coolidge Corner, Allston...
3577
3579
     - Grocery: A full-size Star market is 2 minute...
3582 Funky little apartment close to public transpo...
     My place is close to My home is a warm and fri...
3584
                                                    space \
0
      The house has an open and cozy feel at the sam...
1
      Small but cozy and quite room with a full size...
2
      Come stay with a friendly, middle-aged guy in ...
3
      Most places you find in Boston are small howev...
4
      Clean, attractive, private room, one block fro...
. . .
3576
                                                      NaN
3577
      This space consists of 2 Rooms and a private b...
3579
3582
     Modern and relaxed space with many facilities ...
3584
     To ensure a smooth check in: 1. You MUST have ...
                                              description
                                                           ... requires_license
0
      Cozy, sunny, family home. Master bedroom high...
                                                                               f
1
      Charming and quiet room in a second floor 1910...
                                                                               f
2
      Come stay with a friendly, middle-aged guy in ...
                                                                               f
3
      Come experience the comforts of home away from...
                                                                               f
4
      My comfy, clean and relaxing home is one block...
                                                                               f
. . .
                                                      . . .
                                                            . . .
                                                                             . . .
3576 You'd be living on the top floor of a four sto...
                                                                               f
      Our place is close to Coolidge Corner, Allston...
                                                                               f
3577
     - Grocery: A full-size Star market is 2 minute...
                                                                               f
3579
                                                                               f
3582 Funky little apartment close to public transpo...
3584 My place is close to My home is a warm and fri...
                                                                               f
     license jurisdiction_names instant_bookable
0
         NaN
                             NaN
                                                 f
1
         NaN
                             NaN
                                                 t.
2
         NaN
                             NaN
                                                 f
3
         NaN
                             NaN
                                                 f
4
         NaN
                             NaN
                                                 f
         . . .
                             . . .
. . .
                                               . . .
3576
         \mathtt{NaN}
                             NaN
                                                 f
3577
         {\tt NaN}
                             NaN
                                                 f
         NaN
                                                 f
3579
                             NaN
3582
         NaN
                             NaN
                                                 f
```

```
3584
          {\tt NaN}
                                 {\tt NaN}
                                                       t
     require_guest_profile_picture require_guest_phone_verification
0
                                      f
                                                                             f
1
2
                                      t
                                                                             f
3
                                                                             f
                                      f
4
                                      f
                                                                             f
. . .
3576
                                      f
                                                                             f
3577
                                      f
                                                                             f
3579
                                      f
                                                                             f
3582
                                      f
                                                                             f
3584
                                       f
                                                                             f
     calculated_host_listings_count reviews_per_month mean_review_scores \
0
                                        1
                                                          {\tt NaN}
                                                                           9.307398
                                                         1.30
1
                                        1
                                                                           9.307398
2
                                                         0.47
                                        1
                                                                           9.307398
3
                                        1
                                                         1.00
                                                                           9.307398
4
                                        1
                                                         2.25
                                                                           9.237421
                                                           . . .
3576
                                        1
                                                          {\tt NaN}
                                                                           9.237421
3577
                                                          NaN
                                                                           9.237421
                                        1
3579
                                        1
                                                          {\tt NaN}
                                                                           9.237421
3582
                                        1
                                                          NaN
                                                                           9.237421
3584
                                        3
                                                          {\tt NaN}
                                                                           9.237421
     {\tt mean\_diff}
0
            NaN
      0.307398
1
2
      0.692602
3
      0.692602
      0.762579
4
3576
            NaN
3577
            NaN
            NaN
3579
3582
            NaN
3584
            NaN
```

[1918 rows x 97 columns]

[25]: # Notice that the results are still indexed, but that any of the results which

→were in a group with a mean

review score of less than or equal to 9.2 were not copied over.

2.4 Applying

```
[26]: # By far the most common operation I invoke on groupby objects is the apply()_{\sqcup}
     → function. This allows you to
     # apply an arbitrary function to each group, and stitch the results back for
     →each apply() into a single
     # dataframe where the index is preserved.
     # Lets look at an example using our airbnb data, I'm going to get a clean copyu
     \rightarrow of the dataframe
     df=pd.read_csv("datasets/listings.csv")
     # And lets just include some of the columns we were interested in previously
     df=df[['cancellation_policy','review_scores_value']]
     df.head()
[26]:
      cancellation_policy review_scores_value
                  moderate
                                            NaN
                                            9.0
                  moderate
    1
                                           10.0
    2
                  moderate
    3
                  moderate
                                           10.0
    4
                  flexible
                                           10.0
[27]: # In previous work we wanted to find the average review score of a listing and
     →its deviation from the group
     # mean. This was a two step process, first we used transform() on the groupby_
     →object and then we had to
     # broadcast to create a new column. With apply() we could wrap this logic in
      →one place
     def calc_mean_review_scores(group):
         # group is a dataframe just of whatever we have grouped by, e.g._
     ⇔cancellation policy, so we can treat
         # this as the complete dataframe
         avg=np.nanmean(group["review_scores_value"])
         # now broadcast our formula and create a new column
         group["review_scores_mean"]=np.abs(avg-group["review_scores_value"])
        return group
     # Now just apply this to the groups
     df.groupby('cancellation_policy').apply(calc_mean_review_scores).head()
```

[27]:		cancellation_policy	review_scores_value	review_scores_mean
	0	moderate	NaN	NaN
	1	moderate	9.0	0.307398
	2	moderate	10.0	0.692602
	3	moderate	10.0	0.692602
	4	flexible	10.0	0.762579

```
[28]: # Using apply can be slower than using some of the specialized functions, □ ⇒especially agg(). But, if your # dataframes are not huge, it's a solid general purpose approach
```

Groupby is a powerful and commonly used tool for data cleaning and data analysis. Once you have grouped the data by some category you have a dataframe of just those values and you can conduct aggregated analysis on the segments that you are interested. The groupby() function follows a split-apply-combine approach - first the data is split into subgroups, then you can apply some transformation, filtering, or aggregation, then the results are combined automatically by pandas for us.

Scales

November 15, 2021

1 Scales

```
[1]: # Let's bring in pandas as normal
                   import pandas as pd
                   # Heres an example. Lets create a dataframe of letter grades in descending
                      →order. We can also set an index
                   # value and here we'll just make it some human judgement of how good a student \Box
                      →was, like "excellent" or "good"
                   df=pd.DataFrame(['A+', 'A', 'A-', 'B+', 'B', 'B-', 'C+', 'C', 'C-', 'D+', 'D'],
                                                                                                index=['excellent', 'excellent', 'good', 
                       'ok', 'ok', 'ok', 'poor', 'poor'],
                                                                                           columns=["Grades"])
                  df
                                                                   Grades
                  excellent
                                                                                      A+
                  excellent
                                                                                        Α
```

```
[1]:
    excellent
                   A-
    good
                   B+
    good
                    В
                   B-
    good
    ok
                   C+
    ok
                    C
                   C-
    ok
                   D+
    poor
                    D
    poor
```

```
[2]: # Now, if we check the datatype of this column, we see that it's just anu ⇒object, since we set string values df.dtypes
```

```
[2]: Grades object dtype: object
```

```
[3]: # We can, however, tell pandas that we want to change the type to category,
     \rightarrowusing the astype() function
    df ["Grades"] . astype("category") . head()
[3]: excellent
                  A+
    excellent
                   Α
    excellent
                  A-
    good
                  R+
    good
                   В
    Name: Grades, dtype: category
    Categories (11, object): [A, A+, A-, B, ..., C+, C-, D, D+]
[4]: # We see now that there are eleven categories, and pandas is aware of what
     → those categories are. More
    # interesting though is that our data isn't just categorical, but that it's_{\hspace*{-0.1em}\square}
     ⇔ordered. That is, an A- comes
    # after a B+, and B comes before a B+. We can tell pandas that the data is \sqcup
     →ordered by first creating a new
    # categorical data type with the list of the categories (in order) and the
     →ordered=True flag
    my_categories=pd.CategoricalDtype(categories=['D', 'D+', 'C-', 'C', 'C+', 'B-', _
     \hookrightarrow 'B', 'B+', 'A-', 'A', 'A+'],
                                  ordered=True)
    # then we can just pass this to the astype() function
    grades=df["Grades"].astype(my_categories)
    grades.head()
[4]: excellent
    excellent
                  Α
    excellent
                  A-
                  B+
    good
    good
                   В
    Name: Grades, dtype: category
    Categories (11, object): [D < D+ < C- < C \dots B+ < A- < A < A+]
[5]: # Now we see that pandas is not only aware that there are 11 categories, but it _{\sqcup}
     \hookrightarrow is also aware of the order of
    # those categoreies. So, what can you do with this? Well because there is an u
     →ordering this can help with
    # comparisons and boolean masking. For instance, if we have a list of our
     \hookrightarrow grades and we compare them to a C
    # we see that the lexicographical comparison returns results we were not_{f \sqcup}
     \rightarrow intending.
    df [df ["Grades"]>"C"]
[5]:
         Grades
    ok
              C+
              C-
    ok
```

```
D
   poor
[6]: # So a C+ is great than a C, but a C- and D certainly are not. However, if we
    ⇒broadcast over the dataframe
    # which has the type set to an ordered categorical
    grades[grades>"C"]
[6]: excellent
                A+
   excellent
                 Α
   excellent
                A-
   good
                R+
   good
                 В
                B-
   good
   ok
                C+
   Name: Grades, dtype: category
   Categories (11, object): [D < D+ < C- < C \dots B+ < A- < A < A+]
[7]: # We see that the operator works as we would expect. We can then use a certain
    ⇒set of mathematical operators,
    # like minimum, maximum, etc., on the ordinal data.
[8]: # Sometimes it is useful to represent categorical values as each being a column
    →with a true or a false as to
    # whether the category applies. This is especially common in feature_{\sqcup}
    →extraction, which is a topic in the data
    # mining course. Variables with a boolean value are typically called dummy_{\sqcup}
    \rightarrowvariables, and pandas has a built
    # in function called get\_dummies which will convert the values of a single_{\sqcup}
    →column into multiple columns of
    # zeros and ones indicating the presence of the dummy variable. I rarely use
    \rightarrow it, but when I do it's very
    # handy.
[9]: # Theres one more common scale-based operation Id like to talk about, and thats
    →on converting a scale from
    \# something that is on the interval or ratio scale, like a numeric grade, into
    →one which is categorical. Now,
    # this might seem a bit counter intuitive to you, since you are losing_
    \rightarrow information about the value. But its
    # commonly done in a couple of places. For instance, if you are visualizing the
    → frequencies of categories,
    →with converted interval or ratio
    # data. In addition, if youre using a machine learning classification approach
    →on data, you need to be using
```

poor

D+

```
# categorical data, so reducing dimensionality may be useful just to apply au
     ⇒ qiven technique. Pandas has a
     # function called cut which takes as an argument some array-like structure like_
     \rightarrowa column of a dataframe or a
     # series. It also takes a number of bins to be used, and all bins are kept at \Box
     \rightarrowequal spacing.
     # Lets go back to our census data for an example. We saw that we could group by
     ⇒state, then aggregate to get a
     # list of the average county size by state. If we further apply cut to this
     →with, say, ten bins, we can see
     # the states listed as categoricals using the average county size.
     # let's bring in numpy
     import numpy as np
     # Now we read in our dataset
     df=pd.read_csv("datasets/census.csv")
     # And we reduce this to country data
     df=df[df['SUMLEV']==50]
     # And for a few groups
     df=df.set_index('STNAME').groupby(level=0)['CENSUS2010POP'].agg(np.average)
     df.head()
 [9]: STNAME
    Alabama
                    71339.343284
     Alaska
                    24490.724138
     Arizona
                   426134.466667
    Arkansas
                    38878.906667
                   642309.586207
    California
    Name: CENSUS2010POP, dtype: float64
[10]: # Now if we just want to make "bins" of each of these, we can use cut()
     pd.cut(df,10)
[10]: STNAME
                               (11706.087, 75333.413]
    Alabama
                               (11706.087, 75333.413]
    Alaska
    Arizona
                             (390320.176, 453317.529]
                               (11706.087, 75333.413]
    Arkansas
    California
                             (579312.234, 642309.586]
    Colorado
                              (75333.413, 138330.766]
    Connecticut
                             (390320.176, 453317.529]
                             (264325.471, 327322.823]
    Delaware
    District of Columbia
                             (579312.234, 642309.586]
```

```
Florida
                         (264325.471, 327322.823]
                           (11706.087, 75333.413]
Georgia
Hawaii
                         (264325.471, 327322.823]
Idaho
                           (11706.087, 75333.413]
                          (75333.413, 138330.766]
Illinois
Indiana
                           (11706.087, 75333.413]
                           (11706.087, 75333.413]
Iowa
                           (11706.087, 75333.413]
Kansas
                           (11706.087, 75333.413]
Kentucky
                           (11706.087, 75333.413]
Louisiana
                          (75333.413, 138330.766]
Maine
                         (201328.118, 264325.471]
Maryland
Massachusetts
                         (453317.529, 516314.881]
                          (75333.413, 138330.766]
Michigan
Minnesota
                           (11706.087, 75333.413]
                           (11706.087, 75333.413]
Mississippi
Missouri
                           (11706.087, 75333.413]
                           (11706.087, 75333.413]
Montana
Nebraska
                           (11706.087, 75333.413]
Nevada
                         (138330.766, 201328.118]
New Hampshire
                          (75333.413, 138330.766]
                         (390320.176, 453317.529]
New Jersey
New Mexico
                           (11706.087, 75333.413]
New York
                         (264325.471, 327322.823]
North Carolina
                          (75333.413, 138330.766]
North Dakota
                           (11706.087, 75333.413]
Ohio
                          (75333.413, 138330.766]
                           (11706.087, 75333.413]
Oklahoma
                          (75333.413, 138330.766]
Oregon
                         (138330.766, 201328.118]
Pennsylvania
                         (201328.118, 264325.471]
Rhode Island
South Carolina
                          (75333.413, 138330.766]
                           (11706.087, 75333.413]
South Dakota
                           (11706.087, 75333.413]
Tennessee
                          (75333.413, 138330.766]
Texas
Utah
                          (75333.413, 138330.766]
                           (11706.087, 75333.413]
Vermont
Virginia
                           (11706.087, 75333.413]
                         (138330.766, 201328.118]
Washington
West Virginia
                           (11706.087, 75333.413]
Wisconsin
                          (75333.413, 138330.766]
                           (11706.087, 75333.413]
Wyoming
Name: CENSUS2010POP, dtype: category
Categories (10, interval[float64]): [(11706.087, 75333.413] < (75333.413,
138330.766] < (138330.766, 201328.118] < (201328.118, 264325.471] ...
(390320.176, 453317.529] < (453317.529, 516314.881] < (516314.881, 579312.234] <
(579312.234, 642309.586]]
```

Here we see that states like alabama and alaska fall into the same category, while california and the # disctrict of columbia fall in a very different category.

Now, cutting is just one way to build categories from your data, and thereware many other methods. For # instance, cut gives you interval data, where the spacing between each wategory is equal sized. But sometimes # you want to form categories based on frequency you want the number of items win each bin to the be the # same, instead of the spacing between bins. It really depends on what the want of your data is, and what

youre planning to do with it.

PivotTable_ed

November 15, 2021

A pivot table is a way of summarizing data in a DataFrame for a particular purpose. It makes heavy use of the aggregation function. A pivot table is itself a DataFrame, where the rows represent one variable that you're interested in, the columns another, and the cell's some aggregate value. A pivot table also tends to includes marginal values as well, which are the sums for each column and row. This allows you to be able to see the relationship between two variables at just a glance.

```
[1]: # Lets take a look at pivot tables in pandas
    import pandas as pd
    import numpy as np
[2]: # Here we have the Times Higher Education World University Ranking dataset,
     →which is one of the most
    # influential university measures. Let's import the dataset and see what it_{f \sqcup}
    df = pd.read_csv('datasets/cwurData.csv')
    df.head()
[2]:
       world_rank
                                                institution
                                                                      country
                                        Harvard University
                                                                          USA
    0
                 1
                                                                          USA
    1
                 2
                   Massachusetts Institute of Technology
                 3
    2
                                       Stanford University
                                                                          USA
    3
                 4
                                   University of Cambridge
                                                              United Kingdom
    4
                       California Institute of Technology
                                                                          USA
       national_rank
                       quality_of_education
                                               alumni_employment
                                                                   quality_of_faculty
    0
                                            7
                    2
                                            9
                                                               17
                                                                                      3
    1
    2
                    3
                                           17
                                                                                      5
                                                               11
    3
                    1
                                           10
                                                               24
                                                                                      4
                                                               29
    4
                                            2
       publications
                      influence
                                  citations
                                              broad_impact
                                                            patents
                                                                        score
                                                                               year
    0
                   1
                               1
                                           1
                                                       {\tt NaN}
                                                                   5
                                                                       100.00
                                                                               2012
    1
                  12
                               4
                                           4
                                                        NaN
                                                                   1
                                                                        91.67
                                                                               2012
    2
                   4
                               2
                                           2
                                                                   15
                                                                        89.50
                                                                               2012
                                                       NaN
    3
                              16
                                                                  50
                                                                        86.17
                                                                               2012
                  16
                                          11
                                                       NaN
                  37
                              22
                                          22
                                                       NaN
                                                                  18
                                                                        85.21 2012
```

```
[3]: # Here we can see each institution's rank, country, quality of education, other
     →metrics, and overall score.
    # Let's say we want to create a new column called Rank_Level, where_
     →institutions with world ranking 1-100 are
    # categorized as first tier and those with world ranking 101 - 200 are second
     →tier, ranking 201 - 300 are
    # third tier, after 301 is other top universities.
    # Now, you actually already have enough knowledge to do this, so why don't you
     →pause the video and give it a
    # try?
    # Here's my solution, I'm going to create a function called create_category_{\sqcup}
     \rightarrowwhich will operate on the first
    # column in the dataframe, world rank
    def create_category(ranking):
        # Since the rank is just an integer, I'll just do a bunch of if/elifu
     \rightarrowstatements
        if (ranking >= 1) & (ranking <= 100):
            return "First Tier Top Unversity"
        elif (ranking >= 101) & (ranking <= 200):
            return "Second Tier Top Unversity"
        elif (ranking >= 201) & (ranking <= 300):
            return "Third Tier Top Unversity"
        return "Other Top Unversity"
    # Now we can apply this to a single column of data to create a new series
    df['Rank_Level'] = df['world_rank'].apply(lambda x: create_category(x))
    # And lets look at the result
    df.head()
[3]:
       world rank
                                              institution
                                                                   country
    0
                                       Harvard University
                                                                       USA
                2 Massachusetts Institute of Technology
                                                                       USA
    1
                                      Stanford University
                3
                4
                                  University of Cambridge United Kingdom
    3
    4
                      California Institute of Technology
                                                                       USA
      national_rank quality_of_education alumni_employment
                                                                 quality_of_faculty
    0
                                          7
                                                             9
                   1
                                                                                   1
                   2
                                                                                   3
    1
                                          9
                                                             17
                                                                                  5
    2
                   3
                                         17
                                                             11
    3
                   1
                                         10
                                                             24
                                                                                   4
    4
                                          2
                                                             29
       publications influence citations broad_impact patents
    0
                  1
                                         1
                                                     \mathtt{NaN}
                                                                 5 100.00 2012
```

1	12	4	4	NaN	1	91.67	2012
2	4	2	2	NaN	15	89.50	2012
3	16	16	11	NaN	50	86.17	2012
4	37	22	22	NaN	18	85.21	2012

Rank_Level

- O First Tier Top Unversity
- 1 First Tier Top Unversity
- 2 First Tier Top Unversity
- 3 First Tier Top Unversity
- 4 First Tier Top Unversity

→aggfunc=[np.mean]).head()

```
[4]: # A pivot table allows us to pivot out one of these columns a new columnum headers and compare it against

# another column as row indices. Let's say we want to compare rank level versusum country of the universities

# and we want to compare in terms of overall score

# To do this, we tell Pandas we want the values to be Score, and index to be the country and the columns to be

# the rank levels. Then we specify that the aggregation function, and here we'll use the NumPy mean to get the

# average rating for universities in that country

df.pivot_table(values='score', index='country', columns='Rank_Level', use the level to the country', columns='Rank_Level', use the level to the country', columns='Rank_Level', use the level table to the country', columns='Rank_Level', use the level table table table to the country', columns='Rank_Level', use the level table table
```

[4]: mean \

Rank_Level First Tier Top Unversity Other Top Unversity country Argentina 44.672857 NaNAustralia 47.9425 44.645750 44.864286 Austria ${\tt NaN}$ Belgium 51.8750 45.081000 Brazil NaN44.499706

Rank_Level Second Tier Top Unversity Third Tier Top Unversity country

Argentina	NaN	NaN
Australia	49.2425	47.285000
Austria	NaN	47.066667
Belgium	49.0840	46.746667
Brazil	49.5650	NaN

[5]: # We can see a hierarchical dataframe where the index, or rows, are by country \rightarrow and the columns have two

```
# levels, the top level indicating that the mean value is being used and the
    ⇔second level being our ranks. In
   →we don't really need a
   # heirarchical index.
   # We notice that there are some NaN values, for example, the first row, __
    → Argentia. The NaN values indicate that
   # Argentia has only observations in the "Other Top Universities" category
[6]: # Now, pivot tables aren't limited to one function that you might want to apply.
    \hookrightarrow You can pass a named
   # parameter, aggfunc, which is a list of the different functions to apply, and
    →pandas will provide you with
   # the result using hierarchical column names. Let's try that same query, but \Box
    →pass in the max() function too
   df.pivot_table(values='score', index='country', columns='Rank_Level',_
    →aggfunc=[np.mean, np.max]).head()
[6]:
                                  mean
   Rank_Level First Tier Top Unversity Other Top Unversity
   country
   Argentina
                                   NaN
                                                 44.672857
   Australia
                               47.9425
                                                 44.645750
                                                 44.864286
   Austria
                                   NaN
                               51.8750
                                                 45.081000
   Belgium
   Brazil
                                   {\tt NaN}
                                                 44.499706
   Rank_Level Second Tier Top Unversity Third Tier Top Unversity
   country
   Argentina
                                    NaN
                                                             NaN
                                                       47.285000
   Australia
                                49.2425
   Austria
                                                       47.066667
                                    NaN
                                                       46.746667
   Belgium
                                49.0840
   Brazil
                                49.5650
                                                             NaN
                                  amax
   Rank_Level First Tier Top Unversity Other Top Unversity
   country
   Argentina
                                   {\tt NaN}
                                                     45.66
   Australia
                                 51.61
                                                     45.97
   Austria
                                   \mathtt{NaN}
                                                     46.29
   Belgium
                                 52.03
                                                     46.21
                                                     46.08
   Brazil
                                   \mathtt{NaN}
```

```
country
    Argentina
                                       \mathtt{NaN}
                                                                   NaN
                                     50.40
                                                                 47.47
    Australia
    Austria
                                       NaN
                                                                 47.78
    Belgium
                                     49.73
                                                                 47.14
    Brazil
                                     49.82
                                                                   NaN
[7]: # So now we see we have both the mean and the max. As mentioned earlier, we can
     \rightarrowalso summarize the values
    # within a given top level column. For instance, if we want to see an overall \Box
     →average for the country for the
    # mean and we want to see the max of the max, we can indicate that we want \sqcup
     → pandas to provide marginal values
    df.pivot_table(values='score', index='country', columns='Rank_Level', __
     →aggfunc=[np.mean, np.max],
                    margins=True).head()
[7]:
                                     mean
    Rank_Level First Tier Top Unversity Other Top Unversity
    country
    Argentina
                                      {\tt NaN}
                                                      44.672857
    Australia
                                  47.9425
                                                      44.645750
                                                      44.864286
    Austria
                                      {\tt NaN}
    Belgium
                                  51.8750
                                                      45.081000
                                                      44.499706
    Brazil
                                      {\tt NaN}
                                                                                    \
    Rank_Level Second Tier Top Unversity Third Tier Top Unversity
                                                                               All
    country
    Argentina
                                       NaN
                                                                   NaN 44.672857
                                   49.2425
                                                            47.285000 45.825517
    Australia
    Austria
                                       {\tt NaN}
                                                            47.066667 45.139583
    Belgium
                                   49.0840
                                                            46.746667 47.011000
                                   49.5650
                                                                   NaN 44.781111
    Brazil
                                     amax
    Rank_Level First Tier Top Unversity Other Top Unversity
    country
    Argentina
                                      NaN
                                                          45.66
    Australia
                                    51.61
                                                          45.97
    Austria
                                      \mathtt{NaN}
                                                          46.29
                                    52.03
    Belgium
                                                          46.21
    Brazil
                                      NaN
                                                          46.08
```

Rank_Level Second Tier Top Unversity Third Tier Top Unversity

```
Australia
                                   50.40
                                                            47.47 51.61
   Austria
                                     NaN
                                                            47.78 47.78
                                   49.73
                                                            47.14 52.03
   Belgium
   Brazil
                                   49.82
                                                              NaN 49.82
[8]: # A pivot table is just a multi-level dataframe, and we can access series or
    →cells in the dataframe in a similar way
   # as we do so for a regular dataframe.
   # Let's create a new dataframe from our previous example
   new_df=df.pivot_table(values='score', index='country', columns='Rank_Level',u
    →aggfunc=[np.mean, np.max],
                  margins=True)
   # Now let's look at the index
   print(new_df.index)
   # And let's look at the columns
   print(new_df.columns)
   Index(['Argentina', 'Australia', 'Austria', 'Belgium', 'Brazil', 'Bulgaria',
          'Canada', 'Chile', 'China', 'Colombia', 'Croatia', 'Cyprus',
          'Czech Republic', 'Denmark', 'Egypt', 'Estonia', 'Finland', 'France',
          'Germany', 'Greece', 'Hong Kong', 'Hungary', 'Iceland', 'India', 'Iran',
          'Ireland', 'Israel', 'Italy', 'Japan', 'Lebanon', 'Lithuania',
          'Malaysia', 'Mexico', 'Netherlands', 'New Zealand', 'Norway', 'Poland',
          'Portugal', 'Puerto Rico', 'Romania', 'Russia', 'Saudi Arabia',
          'Serbia', 'Singapore', 'Slovak Republic', 'Slovenia', 'South Africa',
          'South Korea', 'Spain', 'Sweden', 'Switzerland', 'Taiwan', 'Thailand',
          'Turkey', 'USA', 'Uganda', 'United Arab Emirates', 'United Kingdom',
          'Uruguay', 'All'],
         dtype='object', name='country')
   MultiIndex([('mean', 'First Tier Top Unversity'),
               ('mean',
                              'Other Top Unversity'),
               ('mean', 'Second Tier Top Unversity'),
                        'Third Tier Top Unversity'),
               ('mean',
               ('mean',
                                              'All'),
               ('amax', 'First Tier Top Unversity'),
                              'Other Top Unversity'),
               ('amax',
               ('amax', 'Second Tier Top Unversity'),
               ('amax', 'Third Tier Top Unversity'),
               ('amax',
                                              'All')],
              names=[None, 'Rank_Level'])
```

 ${\tt NaN}$

NaN 45.66

country

Argentina

```
# the lower level column indices have four categories, which are the four rank_
     → levels. How would we query this
     # if we want to get the average scores of First Tier Top Unversity levels in_{\square}
     →each country? We would just need
     # to make two dataframe projections, the first for the mean, then the second_{\square}
     → for the top tier
     new_df['mean']['First Tier Top Unversity'].head()
 [9]: country
     Argentina
                      NaN
                  47.9425
    Australia
    Austria
                      NaN
                  51.8750
    Belgium
    Brazil
                      NaN
    Name: First Tier Top Unversity, dtype: float64
[10]: # We can see that the output is a series object which we can confirm by \Box
     →printing the type. Remember that when
     # you project a single column of values out of a DataFrame you get a series.
     type(new_df['mean']['First Tier Top Unversity'])
[10]: pandas.core.series.Series
[11]: # What if we want to find the country that has the maximum average score on
     →First Tier Top University level?
     # We can use the idxmax() function.
     new_df['mean']['First Tier Top Unversity'].idxmax()
[11]: 'United Kingdom'
[12]: \# Now, the idxmax() function isn't special for pivot tables, it's a built in
     → function to the Series object.
     # We don't have time to go over all pandas functions and attributes, and I wantu
     →to encourage you to explore
     # the API to learn more deeply what is available to you.
[13]: # If you want to achieve a different shape of your pivot table, you can do so \square
     →with the stack and unstack
     # functions. Stacking is pivoting the lowermost column index to become the
     →innermost row index. Unstacking is
     # the inverse of stacking, pivoting the innermost row index to become the _{f L}
     → lowermost column index. An example
     # will help make this clear
     # Let's look at our pivot table first to refresh what it looks like
     new_df.head()
[13]:
                                     mean
    Rank_Level First Tier Top Unversity Other Top Unversity
```

country

	Argentina	NaN	4	4.672857	
	Australia	47.9425	4	4.645750	
	Austria	NaN	4	4.864286	
	Belgium	51.8750	4	5.081000	
	Brazil	NaN	4	4.499706	
					\
	Rank Lovo	l Second Tier Top Unversity	Third Tion	Ton Unversity	All
	country	i become free top onversity	iniid iiei	Top onversity	MII
	•	NaN		N - N	44 670057
	Argentina			NaN	
	Australia			47.285000	
	Austria	NaN		47.066667	
	Belgium	49.0840		46.746667	
	Brazil	49.5650		NaN	44.781111
		amax		\	
	Rank_Level	l First Tier Top Unversity	Other Top U	nversity	
	country				
	Argentina	NaN		45.66	
	Australia	51.61		45.97	
	Austria	NaN		46.29	
	Belgium	52.03		46.21	
	Brazil	NaN		46.08	
	Rank Leve	l Second Tier Top Unversity	Third Tier	Top Unversity	All
	country	- contact of the contact of			
	Argentina	NaN		NaN	45.66
	Australia				51.61
	Austria	NaN			47.78
	Belgium	49.73		47.14	
	Brazil				49.82
	prazii	49.82		NaN	49.02
[14]:	# Now let	's try stacking, this shoul	d move the	lowermost colum	n, so the tiers
	→of the	university rankings, to			
	# the inn	er most row			
	new df=nev	w_df.stack()			
	new_df.hea	_			
[14]:			mean	amax	
	country	Rank_Level			
	Argentina	Other Top Unversity	44.672857	45.66	
		All	44.672857	45.66	
	Australia	First Tier Top Unversity	47.942500	51.61	
		Other Top Unversity	44.645750	45.97	
		Second Tier Top Unversity	49.242500	50.40	
		<u>-</u>			

```
⇔stacking, rank levels become the
     # innermost index, appearing to the right after country
     # Now let's try unstacking
     new_df.unstack().head()
[15]:
                                     mean
    Rank_Level First Tier Top Unversity Other Top Unversity
     country
     Argentina
                                      {\tt NaN}
                                                     44.672857
                                  47.9425
                                                     44.645750
    Australia
    Austria
                                      {\tt NaN}
                                                     44.864286
                                  51.8750
    Belgium
                                                     45.081000
    Brazil
                                      NaN
                                                     44.499706
    Rank_Level Second Tier Top Unversity Third Tier Top Unversity
                                                                             All
     country
                                                                 NaN 44.672857
     Argentina
                                       NaN
     Australia
                                   49.2425
                                                           47.285000 45.825517
                                                           47.066667
     Austria
                                       NaN
                                                                      45.139583
    Belgium
                                   49.0840
                                                           46.746667
                                                                      47.011000
                                   49.5650
                                                                 NaN 44.781111
    Brazil
                                     amax
    Rank_Level First Tier Top Unversity Other Top Unversity
     country
     Argentina
                                      NaN
                                                         45.66
     Australia
                                    51.61
                                                         45.97
     Austria
                                      \mathtt{NaN}
                                                         46.29
    Belgium
                                    52.03
                                                         46.21
    Brazil
                                      NaN
                                                         46.08
    Rank_Level Second Tier Top Unversity Third Tier Top Unversity
     country
     Argentina
                                       NaN
                                                                 NaN 45.66
     Australia
                                     50.40
                                                               47.47 51.61
     Austria
                                                               47.78 47.78
                                       {\tt NaN}
                                     49.73
                                                               47.14 52.03
    Belgium
    Brazil
                                     49.82
                                                                 NaN 49.82
[16]: # That seems to restore our dataframe to its original shape. What do you think
     →would happen if we unstacked twice in a row?
     new_df.unstack().unstack().head()
```

[15]: # In the original pivot table, rank levels are the lowermost column, after

```
[16]: Rank_Level country
mean First Tier Top Unversity Argentina NaN
Australia 47.9425
Austria NaN
Belgium 51.8750
Brazil NaN
```

dtype: float64

```
[17]: # We actually end up unstacking all the way to just a single column, so a

→ series object is returned. This

# column is just a "value", the meaning of which is denoted by the

→ heirarachical index of operation, rank, and

# country.
```

So that's pivot tables. This has been a pretty short description, but they're incredibly useful when dealing with numeric data, especially if you're trying to summarize the data in some form. You'll regularly be creating new pivot tables on slices of data, whether you're exploring the data yourself or preparing data for others to report on. And of course, you can pass any function you want to the aggregate function, including those that you define yourself.

DateFunctionality_ed

November 15, 2021

In today's lecture, where we'll be looking at the time series and date functionally in pandas. Manipulating dates and time is quite flexible in Pandas and thus allows us to conduct more analysis such as time series analysis, which we will talk about soon. Actually, pandas was originally created by Wed McKinney to handle date and time data when he worked as a consultant for hedge funds.

```
[1]: # Let's bring in pandas and numpy as usual import pandas as pd import numpy as np
```

0.0.1 Timestamp

```
[2]: # Pandas has four main time related classes. Timestamp, DatetimeIndex, Period, □ → and PeriodIndex. First, let's
# look at Timestamp. It represents a single timestamp and associates values □ → with points in time.

# For example, let's create a timestamp using a string 9/1/2019 10:05AM, and □ → here we have our timestamp.
# Timestamp is interchangeable with Python's datetime in most cases.
pd.Timestamp('9/1/2019 10:05AM')
```

[2]: Timestamp('2019-09-01 10:05:00')

```
[3]: # We can also create a timestamp by passing multiple parameters such as year, □ → month, date, hour, # minute, separately pd.Timestamp(2019, 12, 20, 0, 0)
```

[3]: Timestamp('2019-12-20 00:00:00')

```
[4]: # Timestamp also has some useful attributes, such as isoweekday(), which shows

→ the weekday of the timestamp

# note that 1 represents Monday and 7 represents Sunday

pd.Timestamp(2019, 12, 20, 0, 0).isoweekday()
```

[4]: 5

```
[5]: # You can find extract the specific year, month, day, hour, minute, second from \rightarrow a timestamp
```

```
pd.Timestamp(2019, 12, 20, 5, 2,23).second
```

[5]: 23

0.0.2 Period

```
[6]: # Suppose we weren't interested in a specific point in time and instead wanted

→ a span of time. This is where

# the Period class comes into play. Period represents a single time span, such

→ as a specific day or month.

# Here we are creating a period that is January 2016,
pd.Period('1/2016')
```

[6]: Period('2016-01', 'M')

```
[7]: # You'll notice when we print that out that the granularity of the period is M<sub>□</sub>

→ for month, since that was the

# finest grained piece we provided. Here's an example of a period that is March

→ 5th, 2016.

pd.Period('3/5/2016')
```

[7]: Period('2016-03-05', 'D')

```
[8]: # Period objects represent the full timespan that you specify. Arithmetic on → period is very easy and # intuitive, for instance, if we want to find out 5 months after January 2016, → we simply plus 5 pd.Period('1/2016') + 5
```

[8]: Period('2016-06', 'M')

```
[9]: # From the result, you can see we get June 2016. If we want to find out two⊔

→days before March 5th 2016, we

# simply subtract 2

pd.Period('3/5/2016') - 2
```

[9]: Period('2016-03-03', 'D')

[10]: # The key here is that the period object encapsulates the granularity for⊔

→ arithmetic

0.0.3 DatetimeIndex and PeriodIndex

```
[11]: # The index of a timestamp is DatetimeIndex. Let's look at a quick example. □ → First, let's create our example

# series t1, we'll use the Timestamp of September 1st, 2nd and 3rd of 2016. □ → When we look at the series, each

# Timestamp is the index and has a value associated with it, in this case, a, b□ → and c.
```

```
t1 = pd.Series(list('abc'), [pd.Timestamp('2016-09-01'), pd.
     →Timestamp('2016-09-02'),
                                   pd.Timestamp('2016-09-03')])
     t1
[11]: 2016-09-01
     2016-09-02
     2016-09-03
     dtype: object
[12]: # Looking at the type of our series index, we see that it's DatetimeIndex.
     type(t1.index)
[12]: pandas.core.indexes.datetimes.DatetimeIndex
[13]: # Similarly, we can create a period-based index as well.
     t2 = pd.Series(list('def'), [pd.Period('2016-09'), pd.Period('2016-10'),
                                  pd.Period('2016-11')])
     t2
[13]: 2016-09
    2016-10
     2016-11
                f
    Freq: M, dtype: object
[14]: # Looking at the type of the ts2.index, we can see that it's PeriodIndex.
     type(t2.index)
[14]: pandas.core.indexes.period.PeriodIndex
    0.0.4 Converting to Datetime
[15]: # Now, let's look into how to convert to Datetime. Suppose we have a list of
     \rightarrowdates as strings and we want to
     # create a new dataframe
     # I'm going to try a bunch of different date formats
     d1 = ['2 June 2013', 'Aug 29, 2014', '2015-06-26', '7/12/16']
     # And just some random data
     ts3 = pd.DataFrame(np.random.randint(10, 100, (4,2)), index=d1,
                        columns=list('ab'))
```

[15]: a b
2 June 2013 38 40
Aug 29, 2014 94 99
2015-06-26 75 32
7/12/16 68 62

ts3

```
[16]: # Using pandas to_datetime, pandas will try to convert these to Datetime and
     →put them in a standard format.
     ts3.index = pd.to_datetime(ts3.index)
[16]:
                    b
    2013-06-02 38 40
    2014-08-29 94
                    99
    2015-06-26 75 32
    2016-07-12 68 62
[17]: # to_datetime also() has options to change the date parse order. For example,
     -111e
     # can pass in the argument dayfirst = True to parse the date in European date.
    pd.to_datetime('4.7.12', dayfirst=True)
[17]: Timestamp('2012-07-04 00:00:00')
    0.0.5 Timedelta
[18]: # Timedeltas are differences in times. This is not the same as a a period, but \Box
     →conceptually similar. For
     # instance, if we want to take the difference between September 3rd and \Box
     →September 1st, we get a Timedelta of
     # two days.
     pd.Timestamp('9/3/2016')-pd.Timestamp('9/1/2016')
[18]: Timedelta('2 days 00:00:00')
[19]: # We can also do something like find what the date and time is for 12 days and
     → three hours past September 2nd,
     # at 8:10 AM.
    pd.Timestamp('9/2/2016 8:10AM') + pd.Timedelta('12D 3H')
[19]: Timestamp('2016-09-14 11:10:00')
    0.0.6 Offset
[20]: # Offset is similar to timedelta, but it follows specific calendar duration
     →rules. Offset allows flexibility
     # in terms of types of time intervals. Besides hour, day, week, month, etc it_
     →also has business day, end of
     # month, semi month begin etc
     # Let's create a timestamp, and see what day is that
     pd.Timestamp('9/4/2016').weekday()
```

```
[20]: 6
[21]: # Now we can now add the timestamp with a week ahead
    pd.Timestamp('9/4/2016') + pd.offsets.Week()
[21]: Timestamp('2016-09-11 00:00:00')
[22]: # Now let's try to do the month end, then we would have the last day of
    pd.Timestamp('9/4/2016') + pd.offsets.MonthEnd()
[22]: Timestamp('2016-09-30 00:00:00')
    0.0.7 Working with Dates in a Dataframe
[23]: # Next, let's look at a few tricks for working with dates in a DataFrame.
     →Suppose we want to look at nine
    # measurements, taken bi-weekly, every Sunday, starting in October 2016. Usingu
     → date_range, we can create this
    # DatetimeIndex. In data range, we have to either specify the start or end date.
     → If it is not explicitly
    ⇒specify number of periods, and
    # a frequency. Here, we set it to "2W-SUN", which means biweekly on Sunday
    dates = pd.date_range('10-01-2016', periods=9, freq='2W-SUN')
    dates
[23]: DatetimeIndex(['2016-10-02', '2016-10-16', '2016-10-30', '2016-11-13',
                   '2016-11-27', '2016-12-11', '2016-12-25', '2017-01-08',
                   '2017-01-22'],
                  dtype='datetime64[ns]', freq='2W-SUN')
[24]: # There are many other frequencies that you can specify. For example, you can
     →do business day
    pd.date_range('10-01-2016', periods=9, freq='B')
[24]: DatetimeIndex(['2016-10-03', '2016-10-04', '2016-10-05', '2016-10-06',
                   '2016-10-07', '2016-10-10', '2016-10-11', '2016-10-12',
                   '2016-10-13'],
                  dtype='datetime64[ns]', freq='B')
[25]: # Or you can do quarterly, with the quarter start in June
    pd.date_range('04-01-2016', periods=12, freq='QS-JUN')
[25]: DatetimeIndex(['2016-06-01', '2016-09-01', '2016-12-01', '2017-03-01',
                   '2017-06-01', '2017-09-01', '2017-12-01', '2018-03-01',
                   '2018-06-01', '2018-09-01', '2018-12-01', '2019-03-01'],
                  dtype='datetime64[ns]', freq='QS-JUN')
```

```
[26]: # Now, let's go back to our weekly on Sunday example and create a DataFrame_
      →using these dates, and some random
     # data, and see what we can do with it.
     dates = pd.date_range('10-01-2016', periods=9, freq='2W-SUN')
     df = pd.DataFrame({'Count 1': 100 + np.random.randint(-5, 10, 9).cumsum(),
                        'Count 2': 120 + np.random.randint(-5, 10, 9)}, index=dates)
     df
[26]:
                 Count 1
                          Count 2
     2016-10-02
                     109
                               119
     2016-10-16
                     109
                              127
    2016-10-30
                     117
                              118
    2016-11-13
                              129
                     115
    2016-11-27
                     113
                              122
    2016-12-11
                     122
                              129
    2016-12-25
                     127
                              129
     2017-01-08
                     135
                              117
     2017-01-22
                     138
                              128
[27]: \# First, we can check what day of the week a specific date is. For example,
      →here we can see that all the dates
     # in our index are on a Sunday. Which matches the frequency that we set
     df.index.weekday_name
[27]: Index(['Sunday', 'Sunday', 'Sunday', 'Sunday', 'Sunday', 'Sunday', 'Sunday',
            'Sunday', 'Sunday'],
           dtype='object')
[28]: # We can also use diff() to find the difference between each date's value.
     df.diff()
[28]:
                 Count 1 Count 2
     2016-10-02
                     {\tt NaN}
                              NaN
     2016-10-16
                     0.0
                              8.0
     2016-10-30
                     8.0
                             -9.0
    2016-11-13
                    -2.0
                             11.0
                    -2.0
                             -7.0
    2016-11-27
    2016-12-11
                     9.0
                              7.0
                     5.0
    2016-12-25
                              0.0
     2017-01-08
                     8.0
                            -12.0
     2017-01-22
                     3.0
                             11.0
[29]: # Suppose we want to know what the mean count is for each month in our
      → DataFrame. We can do this using
     # resample. Converting from a higher frequency from a lower frequency is called
      \rightarrow downsampling (we'll talk about
     # this in a moment)
     df.resample('M').mean()
```

```
[29]:
                    Count 1
                                Count 2
    2016-10-31 111.666667
                             121.333333
    2016-11-30 114.000000
                             125.500000
    2016-12-31 124.500000
                             129.000000
    2017-01-31 136.500000
                             122.500000
[30]: # Now let's talk about datetime indexing and slicing, which is a wonderfulu
     → feature of the pandas DataFrame.
     # For instance, we can use partial string indexing to find values from a_{\sqcup}
     →particular year,
     df['2017']
[30]:
                 Count 1
                          Count 2
     2017-01-08
                     135
                               117
                     138
    2017-01-22
                               128
[31]: # Or we can do it from a particular month
     df['2016-12']
[31]:
                 Count 1 Count 2
    2016-12-11
                     122
                               129
    2016-12-25
                     127
                               129
[32]: # Or we can even slice on a range of dates For example, here we only want the
     →values from December 2016
     # onwards.
     df['2016-12':]
[32]:
                 Count 1
                          Count 2
    2016-12-11
                     122
                               129
     2016-12-25
                     127
                               129
     2017-01-08
                     135
                               117
     2017-01-22
                     138
                               128
[33]: df['2016']
[33]:
                 Count 1 Count 2
    2016-10-02
                     109
                               119
    2016-10-16
                     109
                               127
    2016-10-30
                     117
                               118
     2016-11-13
                     115
                               129
    2016-11-27
                     113
                               122
    2016-12-11
                     122
                               129
    2016-12-25
                     127
                               129
```

assignment3

November 15, 2021

1 Assignment 3

All questions are weighted the same in this assignment. This assignment requires more individual learning then the last one did - you are encouraged to check out the pandas documentation to find functions or methods you might not have used yet, or ask questions on Stack Overflow and tag them as pandas and python related. All questions are worth the same number of points except question 1 which is worth 17% of the assignment grade.

Note: Questions 3-13 rely on your question 1 answer.

```
[27]: import pandas as pd
import numpy as np

# Filter all warnings. If you would like to see the warnings, please comment

the two lines below.
import warnings
warnings.filterwarnings('ignore')
```

1.0.1 **Question 1**

Load the energy data from the file assets/Energy Indicators.xls, which is a list of indicators of energy supply and renewable electricity production from the United Nations for the year 2013, and should be put into a DataFrame with the variable name of **Energy**.

Keep in mind that this is an Excel file, and not a comma separated values file. Also, make sure to exclude the footer and header information from the datafile. The first two columns are unneccessary, so you should get rid of them, and you should change the column labels so that the columns are:

['Country', 'Energy Supply', 'Energy Supply per Capita', '% Renewable]

Convert Energy Supply to gigajoules (**Note: there are 1,000,000 gigajoules in a petajoule**). For all countries which have missing data (e.g. data with "...") make sure this is reflected as np.NaN values.

Rename the following list of countries (for use in later questions):

"Republic of Korea": "South Korea", "United States of America": "United States", "United Kingdom of Great Britain and Northern Ireland": "United Kingdom", "China, Hong Kong Special Administrative Region": "Hong Kong"

There are also several countries with numbers and/or parenthesis in their name. Be sure to remove these, e.g. 'Bolivia (Plurinational State of)' should be 'Bolivia'. 'Switzerland17' should be 'Switzerland'.

Next, load the GDP data from the file assets/world_bank.csv, which is a csv containing countries' GDP from 1960 to 2015 from World Bank. Call this DataFrame GDP.

Make sure to skip the header, and rename the following list of countries:

```
"Korea, Rep.": "South Korea", "Iran, Islamic Rep.": "Iran", "Hong Kong SAR, China": "Hong Kong"
```

Finally, load the Sciamgo Journal and Country Rank data for Energy Engineering and Power Technology from the file assets/scimagojr-3.xlsx, which ranks countries based on their journal contributions in the aforementioned area. Call this DataFrame ScimEn.

Join the three datasets: GDP, Energy, and ScimEn into a new dataset (using the intersection of country names). Use only the last 10 years (2006-2015) of GDP data and only the top 15 countries by Scimagojr 'Rank' (Rank 1 through 15).

The index of this DataFrame should be the name of the country, and the columns should be ['Rank', 'Documents', 'Citable documents', 'Citations', 'Self-citations', 'Citations per document', 'H index', 'Energy Supply', 'Energy Supply per Capita', '% Renewable', '2006', '2007', '2008', '2009', '2010', '2011', '2012', '2013', '2014', '2015'].

This function should return a DataFrame with 20 columns and 15 entries, and the rows of the DataFrame should be sorted by "Rank".

```
[24]: def answer_one():
         # YOUR CODE HERE
         import pandas as pd
         import numpy as np
         import re
         Energy=pd.read_excel('assets/Energy Indicators.xls',header=1,skipfooter=1)
         Energy=Energy.drop(['Unnamed: 0','Unnamed: 1'],axis=1)
         for i in range(243,280):
             Energy=Energy.drop(i)
         for i in range (0,16):
             Energy=Energy.drop(i)
         Energy.columns=['Country', 'Energy Supply', 'Energy Supply per Capita', '%_
      →Renewable'l
         Energy['Energy Supply']=Energy['Energy Supply']*1000000
         Energy['Energy Supply'] = Energy['Energy Supply'] . replace('...',np.nan)
         Energy['Country'] = Energy['Country'].replace("Bolivia (Plurinational State_
      →of)","Bolivia")
         pattern2="""(?P<Country2>.*?)(?P<bidule>\s\(.*\)|)"""
         pattern1="""(?P<Country2>.*?)(?P<Num>\d*$)"""
         extract1=Energy['Country'].str.extract(pattern1)
         Energy['Country'] = extract1['Country2']
         Energy['Energy Supply per Capita']=Energy['Energy Supply per Capita'].
      →replace('...',np.nan)
         Energy['Country'] = Energy['Country'].replace("Republic of Korea", "South
      →Korea")
         Energy['Country'] = Energy['Country'].replace("United States of America", __
      →"United States")
```

```
Energy['Country']=Energy['Country'].replace("United Kingdom of Great⊔
             →Britain and Northern Ireland", "United Kingdom")
                    Energy['Country'] = Energy['Country'].replace("China, Hong Kong Special Louis Louis
             →Administrative Region", "Hong Kong")
                    Energy['Country'] = Energy['Country'].replace("Iran (Islamic Republic of)", __
             →"Iran")
                    #extract2=Energy['Country'].str.extract(pattern2)
                    #Energy['Country']=extract2['Country2']
                    GDP=pd.read_csv('assets/world_bank.csv',header=4)
                    GDP=GDP.rename({'Country Name':'Country'},axis='columns')
                    GDP['Country']=GDP['Country'].replace({"Korea, Rep.":"South Korea", "Iran, |
             →Islamic Rep.":"Iran", "Hong Kong SAR, China": "Hong Kong"})
             →GDPmerge=GDP[['Country','2006','2007','2008','2009','2010','2011','2012','2013|,'2014','201
                    ScimEn=pd.read_excel('assets/scimagojr-3.xlsx')
                    ScimEnmerge=ScimEn[ScimEn['Rank']<16]</pre>
                    #print(len(ScimEn)-len(ScimEnmerge))
                    merge1=pd.merge(ScimEnmerge,Energy,how='left',on='Country')
                    merge2=pd.merge(merge1,GDPmerge,how='left',on='Country')
                    merge2=merge2.set_index('Country')
                    return merge2
           answer_one()
[24]:
                                                        Rank Documents Citable documents Citations \
           Country
           China
                                                               1
                                                                             127050
                                                                                                                        126767
                                                                                                                                                 597237
                                                               2
           United States
                                                                                                                                                 792274
                                                                               96661
                                                                                                                          94747
           Japan
                                                               3
                                                                               30504
                                                                                                                          30287
                                                                                                                                                 223024
           United Kingdom
                                                               4
                                                                               20944
                                                                                                                          20357
                                                                                                                                                 206091
           Russian Federation
                                                                               18534
                                                                                                                          18301
                                                               5
                                                                                                                                                   34266
           Canada
                                                               6
                                                                                                                          17620
                                                                                                                                                 215003
                                                                               17899
                                                               7
           Germany
                                                                               17027
                                                                                                                          16831
                                                                                                                                                 140566
           India
                                                               8
                                                                               15005
                                                                                                                          14841
                                                                                                                                                 128763
           France
                                                               9
                                                                               13153
                                                                                                                          12973
                                                                                                                                                 130632
           South Korea
                                                             10
                                                                               11983
                                                                                                                          11923
                                                                                                                                                 114675
           Italy
                                                             11
                                                                               10964
                                                                                                                          10794
                                                                                                                                                 111850
           Spain
                                                             12
                                                                                 9428
                                                                                                                            9330
                                                                                                                                                 123336
           Iran
                                                             13
                                                                                 8896
                                                                                                                            8819
                                                                                                                                                   57470
           Australia
                                                             14
                                                                                 8831
                                                                                                                            8725
                                                                                                                                                   90765
           Brazil
                                                             15
                                                                                 8668
                                                                                                                            8596
                                                                                                                                                   60702
                                                        Self-citations Citations per document H index \
           Country
           China
                                                                          411683
                                                                                                                                     4.70
                                                                                                                                                            138
           United States
                                                                          265436
                                                                                                                                     8.20
                                                                                                                                                            230
```

Japan	6155	4	7.31	134	
United Kingdom	3787		9.84	139	
Russian Federation	12422		1.85	57	
Canada	40930		12.01	149	
Germany	27426		8.26	126	
India	37209		8.58	115	
France	28601		9.93	114	
South Korea	22595		9.57	104	
Italy	26661		10.20	106	
Spain	23964		13.08	115	
Iran	19125		6.46	72	
Australia	15606		10.28	107	
Brazil	14396		7.00	86	
	Energy Supply	Energy Supply	per Capita %	Renewable \	
Country					
China	127191000000		93.0	19.7549	
United States	90838000000		286.0	11.571	
Japan	18984000000		149.0	10.2328	
United Kingdom	7920000000		124.0	10.6005	
Russian Federation	30709000000		214.0	17.2887	
Canada	10431000000		296.0	61.9454	
Germany	13261000000		165.0	17.9015	
India	33195000000		26.0	14.9691	
France	10597000000		166.0	17.0203	
South Korea	11007000000		221.0	2.27935	
Italy	6530000000		109.0	33.6672	
Spain	4923000000		106.0	37.9686	
Iran	9172000000		119.0	5.70772	
Australia	5386000000		231.0	11.8108	
Brazil	12149000000		59.0	69.648	
	2006	2007	2008	2009	\
Country					
China	3.992331e+12	4.559041e+12	4.997775e+12	5.459247e+12	
United States	1.479230e+13	1.505540e+13	1.501149e+13	1.459484e+13	
Japan	5.496542e+12	5.617036e+12	5.558527e+12	5.251308e+12	
United Kingdom	2.419631e+12	2.482203e+12	2.470614e+12	2.367048e+12	
Russian Federation	1.385793e+12	1.504071e+12	1.583004e+12	1.459199e+12	
Canada	1.564469e+12	1.596740e+12	1.612713e+12	1.565145e+12	
Germany	3.332891e+12	3.441561e+12	3.478809e+12	3.283340e+12	
India	1.265894e+12	1.374865e+12	1.428361e+12	1.549483e+12	
France	2.607840e+12	2.669424e+12	2.674637e+12	2.595967e+12	
South Korea	9.410199e+11	9.924316e+11	1.020510e+12	1.027730e+12	
Italy	2.202170e+12	2.234627e+12	2.211154e+12	2.089938e+12	
Spain	1.414823e+12	1.468146e+12	1.484530e+12		
Iran	3.895523e+11		4.289909e+11	4.389208e+11	

```
Australia
                       1.021939e+12 1.060340e+12 1.099644e+12 1.119654e+12
   Brazil
                       1.845080e+12 1.957118e+12 2.056809e+12 2.054215e+12
                               2010
                                            2011
                                                          2012
                                                                        2013 \
   Country
   China
                       6.039659e+12 6.612490e+12 7.124978e+12 7.672448e+12
   United States
                       1.496437e+13 1.520402e+13 1.554216e+13 1.577367e+13
   Japan
                       5.498718e+12 5.473738e+12 5.569102e+12 5.644659e+12
                       2.403504e+12 2.450911e+12 2.479809e+12 2.533370e+12
   United Kingdom
   Russian Federation 1.524917e+12 1.589943e+12 1.645876e+12 1.666934e+12
   Canada
                       1.613406e+12 1.664087e+12 1.693133e+12 1.730688e+12
   Germany
                       3.417298e+12 3.542371e+12 3.556724e+12 3.567317e+12
   India
                       1.708459e+12 1.821872e+12 1.924235e+12 2.051982e+12
   France
                       2.646995e+12 2.702032e+12 2.706968e+12 2.722567e+12
   South Korea
                       1.094499e+12 1.134796e+12 1.160809e+12 1.194429e+12
   Italy
                       2.125185e+12 2.137439e+12 2.077184e+12 2.040871e+12
   Spain
                       1.431673e+12 1.417355e+12 1.380216e+12 1.357139e+12
                       4.677902e+11 4.853309e+11 4.532569e+11 4.445926e+11
   Iran
   Australia
                       1.142251e+12 1.169431e+12 1.211913e+12 1.241484e+12
   Brazil
                       2.208872e+12 2.295245e+12 2.339209e+12 2.409740e+12
                               2014
                                            2015
   Country
   China
                       8.230121e+12 8.797999e+12
   United States
                       1.615662e+13 1.654857e+13
   Japan
                       5.642884e+12 5.669563e+12
   United Kingdom
                       2.605643e+12 2.666333e+12
   Russian Federation 1.678709e+12 1.616149e+12
                       1.773486e+12 1.792609e+12
   Canada
                       3.624386e+12 3.685556e+12
   Germany
   India
                       2.200617e+12 2.367206e+12
                       2.729632e+12 2.761185e+12
   France
   South Korea
                       1.234340e+12 1.266580e+12
   Italy
                       2.033868e+12 2.049316e+12
   Spain
                       1.375605e+12 1.419821e+12
   Iran
                       4.639027e+11
                                             NaN
                       1.272520e+12 1.301251e+12
   Australia
   Brazil
                       2.412231e+12 2.319423e+12
[2]: assert type(answer_one()) == pd.DataFrame, "Q1: You should return a DataFrame!"
   assert answer_one().shape == (15,20), "Q1: Your DataFrame should have 20__
    ⇒columns and 15 entries!"
```

```
NameError
                                                 Traceback (most recent call_
→last)
       <ipython-input-2-e4f26ba1c257> in <module>
  ----> 1 assert type(answer_one()) == pd.DataFrame, "Q1: You should return a_
→DataFrame!"
        3 assert answer_one().shape == (15,20), "Q1: Your DataFrame should_
→have 20 columns and 15 entries!"
       <ipython-input-1-bab2565315dc> in answer_one()
         1 def answer one():
               # YOUR CODE HERE
  ---> 3
               df=pd.read_excel('assets/Energy Indicators.
→xls',header=1,skipfooter=1)
               df=df.drop(['Unnamed: 0','Unnamed: 1'],axis=1)
        5
               for i in range(243,280):
      NameError: name 'pd' is not defined
```

[]: # Cell for autograder.

1.0.2 **Question 2**

The previous question joined three datasets then reduced this to just the top 15 entries. When you joined the datasets, but before you reduced this to the top 15 items, how many entries did you lose?

This function should return a single number.

```
[25]: def answer_two():
    return int(176)
    answer_two()

[25]: 176

[]: assert type(answer_two()) == int, "Q2: You should return an int number!"
```

1.0.3 Question 3

What are the top 15 countries for average GDP over the last 10 years?

This function should return a Series named avgGDP with 15 countries and their average GDP sorted in descending order.

```
[38]: def answer_three():
         GDP=answer_one()
      →GDPmean=GDP[['2006','2007','2008','2009','2010','2011','2012','2013','2014','2015']]
         def meangdp(row):
      →cols=['2006','2007','2008','2009','2010','2011','2012','2013','2014','2015']
             data=row[cols]
             return pd.Series({'mean': np.mean(data)})
         GDPmean=GDPmean.apply(meangdp,axis=1)
         GDPmean=GDPmean.sort_values(ascending=False,by='mean')
         return GDPmean['mean']
     answer_three()
[38]: Country
    United States
                           1.536434e+13
     China
                           6.348609e+12
     Japan
                           5.542208e+12
     Germany
                           3.493025e+12
                           2.681725e+12
    France
                           2.487907e+12
    United Kingdom
    Brazil
                           2.189794e+12
     Italy
                           2.120175e+12
     India
                           1.769297e+12
     Canada
                           1.660647e+12
    Russian Federation
                           1.565459e+12
                           1.418078e+12
     Spain
     Australia
                           1.164043e+12
     South Korea
                           1.106715e+12
                           4.441558e+11
     Iran
     Name: mean, dtype: float64
 []: assert type(answer_three()) == pd.Series, "Q3: You should return a Series!"
```

1.0.4 Question 4

By how much had the GDP changed over the 10 year span for the country with the 6th largest average GDP?

This function should return a single number.

[46]: 246702696075.3999

```
[47]: # Cell for autograder.
```

1.0.5 **Question 5**

What is the mean energy supply per capita?

This function should return a single number.

```
[32]: def answer_five():
    # YOUR CODE HERE
    meanEnergy=answer_one()
    mean=np.mean(meanEnergy['Energy Supply per Capita'])
    return mean
answer_five()
```

[32]: 157.6

```
[]: # Cell for autograder.
```

1.0.6 Question 6

What country has the maximum % Renewable and what is the percentage? *This function should return a tuple with the name of the country and the percentage.*

```
[33]: def answer_six():
    import numpy as np# YOUR CODE HERE
    maxrenew=answer_one()
    renew=maxrenew['% Renewable']
    renew=renew.astype('float64')
    maxi=np.max(maxrenew['% Renewable'])
    idx=renew.idxmax(skipna=True)
    return (idx,maxi)
answer_six()
```

[33]: ('Brazil', 69.64803)

```
[66]: assert type(answer_six()) == tuple, "Q6: You should return a tuple!"
     assert type(answer_six()[0]) == str, "Q6: The first element in your result_
      ⇒should be the name of the country!"
    69.64803
            NotImplementedError
                                                      Traceback (most recent call
     →last)
            <ipython-input-66-daa95b3480e4> in <module>
        ---> 1 assert type(answer_six()) == tuple, "Q6: You should return a tuple!"
              3 assert type(answer_six()[0]) == str, "Q6: The first element in your_
     ⇒result should be the name of the country!"
            <ipython-input-65-1dd3f65ccb19> in answer_six()
                    answer=np.max(maxrenew['% Renewable'])
                    print(answer)
                    raise NotImplementedError()
        ---> 6
              7 answer_six()
```

NotImplementedError:

1.0.7 **Question** 7

Create a new column that is the ratio of Self-Citations to Total Citations. What is the maximum value for this new column, and what country has the highest ratio?

This function should return a tuple with the name of the country and the ratio.

```
[34]: def answer_seven():
         # YOUR CODE HERE
         import numpy as np
         ratio=answer_one()
         ratio['ratio']=ratio['Self-citations']/ratio['Citations']
         maxi=np.max(ratio['ratio'])
         cit=ratio['ratio']
         country=cit.idxmax()
         maxi=np.max(ratio['ratio'])
         return (country,maxi)
     answer_seven()
```

```
[34]: ('China', 0.6893126179389422)
[13]: assert type(answer_seven()) == tuple, "Q7: You should return a tuple!"
    assert type(answer_seven()[0]) == str, "Q7: The first element in your result_
     ⇒should be the name of the country!"
                          ______
           NotImplementedError
                                                   Traceback (most recent call_
     →last)
           <ipython-input-13-a29f0eae5d14> in <module>
        ----> 1 assert type(answer_seven()) == tuple, "Q7: You should return a tuple!
             3 assert type(answer_seven()[0]) == str, "Q7: The first element in ∪
     →your result should be the name of the country!"
           <ipython-input-12-f5f91fb135c5> in answer_seven()
                 cit=ratio['ratio']
             8
                 country=cit.idxmax()
       ---> 9
                 raise NotImplementedError()
                  return country
            11 answer_seven()
           NotImplementedError:
```

1.0.8 **Question 8**

Create a column that estimates the population using Energy Supply and Energy Supply per capita. What is the third most populous country according to this estimate?

This function should return the name of the country

```
[35]: def answer_eight():
    # YOUR CODE HERE
    pop=answer_one()
    pop=pop.reset_index()
    pop['population']=pop['Energy Supply']/pop['Energy Supply per Capita']
    pop=pop.sort_values(by='population',ascending=False)
    return pop.iloc[2]['Country']
answer_eight()
```

[35]: 'United States'

```
[]: assert type(answer_eight()) == str, "Q8: You should return the name of the →country!"
```

1.0.9 Question 9

Create a column that estimates the number of citable documents per person. What is the correlation between the number of citable documents per capita and the energy supply per capita? Use the .corr() method, (Pearson's correlation).

This function should return a single number.

(Optional: Use the built-in function plot9() to visualize the relationship between Energy Supply per Capita vs. Citable docs per Capita)

```
[42]: def answer_nine():
         # YOUR CODE HERE
         import scipy.stats as stats
        Top15=answer_one()
        Top15['PopEst'] = Top15['Energy Supply'] / Top15['Energy Supply per Capita']
        Top15['Citable docs per Capita'] = Top15['Citable documents'] /
      →Top15['PopEst']
        CORR, PVAL=stats.pearsonr(Top15['Citable docs per Capita'],Top15['PopEst'])
        def plot9():
             import matplotlib as plt
            %matplotlib inline
            Top15 = answer_one()
            Top15['PopEst'] = Top15['Energy Supply'] / Top15['Energy Supply per⊔
      Top15['Citable docs per Capita'] = Top15['Citable documents'] /
      →Top15['PopEst']
             Top15['Citable docs per Capita']=Top15['Citable docs per Capita'].
      →replace(np.nan,0)
            Top15.plot(x='Citable docs per Capita', y='Energy Supply per Capita',

→kind='scatter', xlim=[0, 0.0006])
        return CORR
    answer_nine()
```

[42]: -0.5296539178693623

```
[]: assert answer_nine() >= -1. and answer_nine() <= 1., "Q9: A valid correlation → should between -1 to 1!"
```

1.0.10 Question 10

Create a new column with a 1 if the country's % Renewable value is at or above the median for all countries in the top 15, and a 0 if the country's % Renewable value is below the median.

This function should return a series named HighRenew whose index is the country name sorted in ascending order of rank.

```
[11]: def answer_ten():
         # YOUR CODE HERE
         Top15=answer_one()
         Top15=Top15.reset_index()
         med=Top15['% Renewable'].median(skipna=True)
         def fonction(row, m=med):
             data=row['% Renewable']
             if data>=m:
                 row['HighRenew']=1
             else:
                 row['HighRenew']=0
             return row
         serie=Top15.apply(fonction,axis=1)
         serie=serie.set_index('Country')
         return serie['HighRenew']
     answer_ten()
```

```
[11]: Country
     China
                            1
     United States
                            0
                            0
     Japan
     United Kingdom
                            0
     Russian Federation
     Canada
     Germany
                            1
     India
                            0
    France
                            0
     South Korea
                            0
     Italy
                            1
     Spain
     Iran
                            0
     Australia
                            0
     Brazil
     Name: HighRenew, dtype: int64
 []: assert type(answer_ten()) == pd.Series, "Q10: You should return a Series!"
```

1.0.11 Question 11

Use the following dictionary to group the Countries by Continent, then create a DataFrame that displays the sample size (the number of countries in each continent bin), and the sum, mean, and std deviation for the estimated population of each country.

This function should return a DataFrame with index named Continent ['Asia', 'Australia', 'Europe', 'North America', 'South America'] and columns ['size', 'sum', 'mean', 'std']

```
[41]: def answer_eleven():
         # YOUR CODE HERE
         Top15=answer_one()
         ContinentDict = {'China':'Asia',
                        'United States': 'North America',
                        'Japan':'Asia',
                        'United Kingdom': 'Europe',
                        'Russian Federation': 'Europe',
                        'Canada':'North America',
                        'Germany': 'Europe',
                        'India':'Asia',
                        'France': 'Europe',
                        'South Korea': 'Asia',
                        'Italy': 'Europe',
                        'Spain': 'Europe',
                        'Iran': 'Asia',
                        'Australia': 'Australia',
                        'Brazil':'South America'}
         Top15['PopEst'] = Top15['Energy Supply'] / Top15['Energy Supply per Capita']
         Top15['PopEst']=Top15['PopEst'].astype('float64')
         #print(Top15['PopEst'])
         size=Top15.groupby(by=ContinentDict).agg({'PopEst':('count',np.sum,np.
      →mean,np.std)})
```

```
size.columns=['size','sum','mean','std']
         size.index.names = ['Continent']
        return size
    answer_eleven()
[41]:
                   size
                                   sum
                                               mean
                                                               std
    Continent
                      4 2.821591e+09 7.053977e+08 7.138779e+08
    Asia
    Australia
                      1 2.331602e+07 2.331602e+07
    Europe
                      5 3.940587e+08 7.881174e+07 3.813228e+07
    North America
                      1 3.523986e+07 3.523986e+07
                                                              NaN
    South America
                      1 2.059153e+08 2.059153e+08
                                                              NaN
 | : assert type(answer_eleven()) == pd.DataFrame, "Q11: You should return au
      →DataFrame!"
    assert answer_eleven().shape[0] == 5, "Q11: Wrong row numbers!"
```

1.0.12 **Question 12**

Cut % Renewable into 5 bins. Group Top15 by the Continent, as well as these new % Renewable bins. How many countries are in each of these groups?

assert answer_eleven().shape[1] == 4, "Q11: Wrong column numbers!"

This function should return a Series with a MultiIndex of Continent, then the bins for % Renewable. Do not include groups with no countries.

```
[48]: def answer_twelve():
         # YOUR CODE HERE
         Top15=answer_one()
         Top15=Top15.reset_index()
         ContinentDict = {'China':'Asia',
                        'United States': 'North America',
                        'Japan':'Asia',
                        'United Kingdom': 'Europe',
                        'Russian Federation': 'Europe',
                        'Canada':'North America',
                        'Germany': 'Europe',
                        'India': 'Asia',
                        'France': 'Europe',
                        'South Korea': 'Asia',
                        'Italy': 'Europe',
                        'Spain': 'Europe',
                        'Iran':'Asia',
                        'Australia': 'Australia',
                        'Brazil':'South America'}
         result=pd.cut(Top15['% Renewable'],5)
         result.name='Renewable bins'
```

```
[48]: Continent
     Asia
                     (2.212, 15.753]
                     (15.753, 29.227]
                     (2.212, 15.753]
     Australia
                                         1
                     (2.212, 15.753]
     Europe
                                         1
                     (15.753, 29.227]
                                         3
                     (29.227, 42.701]
                                         2
     North America (2.212, 15.753]
                                         1
                     (56.174, 69.648]
                                         1
     South America (56.174, 69.648]
     Name: Renewable bin, dtype: int64
```

```
[]: assert type(answer_twelve()) == pd.Series, "Q12: You should return a Series!"
assert len(answer_twelve()) == 9, "Q12: Wrong result numbers!"
```

1.0.13 Question 13

Convert the Population Estimate series to a string with thousands separator (using commas). Use all significant digits (do not round the results).

```
e.g. 12345678.90 -> 12,345,678.90
```

This function should return a series *PopEst* whose index is the country name and whose values are the population estimate string

```
[43]: def answer_thirteen():
    # YOUR CODE HERE
    Top15=answer_one()
    Top15['PopEst'] = Top15['Energy Supply'] / Top15['Energy Supply per Capita']
    Top15['PopEst']=Top15['PopEst'].astype('float64')
    Top15['PopEst']=Top15['PopEst'].apply('{:,}'.format)
    return Top15['PopEst']
    raise NotImplementedError()
answer_thirteen()
```

```
[43]: Country
China 1,367,645,161.2903225
United States nan
Japan 127,409,395.97315437
United Kingdom nan
```

```
143,500,000.0
                            35,239,864.86486486
   Canada
   Germany
                           80,369,696.96969697
   India
                         1,276,730,769.2307692
   France
                           63,837,349.39759036
   South Korea
                          49,805,429.864253394
                          59,908,256.880733944
   Italy
   Spain
                            46,443,396.2264151
   Iran
   Australia
                          23,316,017.316017315
   Brazil
                          205,915,254.23728815
   Name: PopEst, dtype: object
]: assert type(answer thirteen()) == pd.Series, "Q13: You should return a Series!"
   assert len(answer_thirteen()) == 15, "Q13: Wrong result numbers!"
```

1.0.14 Optional

Russian Federation

Use the built in function plot_optional() to see an example visualization.

```
[]: def plot_optional():
       import matplotlib as plt
       %matplotlib inline
       Top15 = answer one()
       ax = Top15.plot(x='Rank', y='% Renewable', kind='scatter',
    →c=['#e41a1c','#377eb8','#e41a1c','#4daf4a','#4daf4a','#377eb8','#4daf4a','#e41a1c',
    _{\hookrightarrow}'#4daf4a','#e41a1c','#4daf4a','#4daf4a','#e41a1c','#dede00','#ff7f00'],
                       xticks=range(1,16), s=6*Top15['2014']/10**10, alpha=.75,
    \rightarrowfigsize=[16,6]);
       for i, txt in enumerate(Top15.index):
           ax.annotate(txt, [Top15['Rank'][i], Top15['% Renewable'][i]],
    →ha='center')
       print("This is an example of a visualization that can be created to help,
    →understand the data. \
   This is a bubble chart showing \% Renewable vs. Rank. The size of the bubble \sqcup
    2014 GDP, and the color corresponds to the continent.")
```