Viewpoint

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1 Viewpoint Assessment

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2.1 Question 1

Please reference the below tables, where entries shown are purely examples of what could be in these tables.

The first table, Students, has the following columns: - id, a unique integer for each student - name, a string containing the student's name, where the first name is capitalized.

The second table, Assignments, has the following columns: - id, a unique integer for each assignment a student has taken - student_id, a foreign key referring to the student table - grade, an integer giving a student's grade on an assignment

2.1.1 A. Write a SQL query that returns the id, name, and maximum assignment grade of the student(s) with the highest average grade across all assignments.

2.1.2 B. Assume that assignments that are not submitted are simply not entered into the Assignments table. Write a SQL query that returns the names of the students that have not submitted an assignment.

Now assume both tables are instead given to you as pandas DataFrame objects called students_df and assignments_df.

2.1.3 C. Write a python function, utilizing pandas, that takes in student_df and returns a new DataFrame of the same id's and name's, but where now each name containing the letter "e" is uppercased, and lowercased otherwise (e.g. "Edward" → "EDWARD", "Bob" → "bob").

This block imports panda and creates a small DataFrame for testing C and D.

```
[1]: import pandas as pd
     students_df = pd.DataFrame(
         [[1, 'Edward'],
          [2, 'Bob'],
          [3, 'Zec'],
          [4, 'Phil'],
          [5, 'Aarone']],
         index = [1,2,3,4,5],
         columns = ['id', 'name']
         )
     assignments_df = pd.DataFrame(
         [[1, 1, 100],
          [2, 1, 90],
          [3, 1, 95],
          [4, 2, 79],
          [5, 2, 2],
          [6, 3, 84],
          [7, 3, 87],
          [8, 3, 79],
          [9, 4, 95],
          [10, 4, 95],
          [11, 4, 95],
          [12, 5, 10],
          [13, 5, 87],
          [14, 5, 34]],
         index = [1,2,3,4,5,6,7,8,9,10,11,12,13,14],
         columns = ['id', 'student_id', 'grade'])
```

```
[2]: def check_e(df):
    """

Takes in a dataframe of students and puts any names with an 'e' to uppercase and any name without an 'e' to lowercase.

Parameters
-----
df: DataFrame
```

Students before:

```
[3]: students_df
```

- [3]: id name
 - 1 1 Edward
 - 2 2 Bob
 - 3 3 Zec
 - 4 4 Phil
 - 5 5 Aarone
- [4]: students_lower_df = check_e(students_df)

Student's after:

[5]: students_lower_df

- [5]: id name
 - 1 1 EDWARD
 - 2 2 bob
 - 3 3 ZEC
 - 4 4 phil
 - 5 5 AARONE

2.1.4 D. Now write a python function that takes in the output DataFrame of the function written above, along with assignments_df, and returns a data frame that summarizes the average grade of uppercase named students and lowercase named students.

```
[6]: def case_average(students_df, assignments_df):
         Takes in students of after running check e() and assignments of. Averages
         grades of students with uppercase and lowercase names.
         Parameters
         students\_df : DataFrame
             Holds information of students with all names either all uppercase or all
             lowercase based on if they contain an 'e'.
         assignments df : DataFrame
            Holds information of assignments turned in by students.
         Returns
        Average of grades with students grouped by the case of their name.
         11 11 11
         # combines the DataFrames using students.id and assignments.student id
         combined_df = students_df.merge(assignments_df, left_on='id',_
      # removes unnecessary columns
         combined_df.drop(['id_x', 'id_y', 'student_id'], inplace = True, axis = 1)
         # averages grades based off the case of students name
        upper_avg = combined_df[combined_df.name.str.isupper()].mean()
        lower_avg = combined_df[combined_df.name.str.islower()].mean()
         # creates new dataframe holding the averages
        rtn_df = pd.DataFrame({'Uppercase': upper_avg, 'Lowercase': lower_avg})
        rtn_df.rename(index = {'grade': 'average'}, inplace = True)
        return rtn_df
```

Finding averages of uppercase and lowercase Students:

```
[7]: averages = case_average(students_lower_df, assignments_df)
[8]: averages
[8]: Uppercase Lowercase average 74.0 73.2
```

2.2 Question 2

Consider the data set below. Write some python code that illustrates some common feature engineering and/or data preparation tasks.

https://raw.githubusercontent.com/ireapps/white-house-salaries-2017/master/white_house_2017_salaries.csv Importing data:

```
[9]: wh_df = pd.read_csv('https://raw.githubusercontent.com/ireapps/

→white-house-salaries-2017/master/white_house_2017_salaries.csv')
```

We will run a few commands to see the basics of the dataset.

```
[10]: wh_df.head()
```

```
[10]:
                            NAME
                                    STATUS
                                                           PAY BASIS
                                                   SALARY
           Alexander, Monica K.
                                  Employee
                                              $56,000.00
                                                           Per Annum
      0
          Ambrosini, Michael J.
                                  Employee
      1
                                              $95,000.00
                                                           Per Annum
      2
                 Amin, Stacy C.
                                  Employee
                                             $140,000.00
                                                           Per Annum
      3
           Andersen, Whitney N.
                                  Employee
                                              $94,000.00
                                                           Per Annum
        Anderson, Alexander J.
                                  Employee
                                              $77,000.00
                                                           Per Annum
```

POSITION TITLE

- O EXECUTIVE ASSISTANT
- 1 SPECIAL ASSISTANT TO THE PRESIDENT AND DIRECTO...
- 2 SPECIAL ASSISTANT TO THE PRESIDENT AND ASSOCIA...
- 3 DEPUTY DIRECTOR OF OPERATIONS FOR THE WHITE HO...
- 4 DIRECTOR OF DIGITAL ENGAGEMENT

[11]: wh_df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 377 entries, 0 to 376
Data columns (total 5 columns):

#	Column	Non-Null Count	Dtype
0	NAME	377 non-null	object
1	STATUS	377 non-null	object
2	SALARY	377 non-null	object
3	PAY BASIS	377 non-null	object
4	POSITION TITLE	377 non-null	object

dtypes: object(5)
memory usage: 14.9+ KB

[12]: wh_df.describe()

[12]:	NAME	STATUS	SALARY	PAY BASIS	\
count	377	377	377	377	
unique	377	3	74	1	

```
Walters, Lindsay E.
                               Employee
                                         $115,000.00
                                                        Per Annum
top
                                                               377
                                    358
                                                    46
freq
             POSITION TITLE
                         377
count
unique
                         215
top
        EXECUTIVE ASSISTANT
freq
                          20
```

- head() shows the first four rows of the dataset so we can see what it looks like.
- info() shows each column as well as its type and amount of non-null entries.
- describe() shows how many rows there are and how many unique values there are in each column.

Between info() and describe() we see that there are 377 rows and each column has 377 non-null values.

We also see that SALARY is stored as an object. If we want to perform any analysis on it, we will need it as a numeric value. In this case, storing it as an int would likely work as it seems like all of the salaries are whole numbers, but I am going to convert it to a float just in case there are any with decimal values.

```
[13]: wh_df.SALARY = wh_df.SALARY.astype(str).str.replace(',', '').str.strip('$').
       →astype(float)
[14]:
     wh_df.head()
[14]:
                            NAME
                                    STATUS
                                              SALARY
                                                      PAY BASIS
      0
           Alexander, Monica K.
                                  Employee
                                             56000.0
                                                      Per Annum
      1
          Ambrosini, Michael J.
                                  Employee
                                             95000.0
                                                      Per Annum
      2
                 Amin, Stacy C.
                                  Employee
                                                      Per Annum
                                            140000.0
      3
           Andersen, Whitney N.
                                  Employee
                                                      Per Annum
                                             94000.0
         Anderson, Alexander J.
                                  Employee
                                             77000.0 Per Annum
                                             POSITION TITLE
      0
                                        EXECUTIVE ASSISTANT
         SPECIAL ASSISTANT TO THE PRESIDENT AND DIRECTO...
      1
         SPECIAL ASSISTANT TO THE PRESIDENT AND ASSOCIA...
      3 DEPUTY DIRECTOR OF OPERATIONS FOR THE WHITE HO...
      4
                             DIRECTOR OF DIGITAL ENGAGEMENT
```

Another problem is 'PAY BASIS' and 'POSITION TITLE' contain a whitespace in the name, which I will replace with an underscore.

I am also going to make them lowercase to make it a little easier to work with.

```
[15]: wh_df.columns = wh_df.columns.str.replace("\s", "_", regex = True).str.lower()
[16]: wh_df.columns
```

Now we will look at the unique values in Status and Position Title to see if there is anything that could be used for models.

```
[17]: wh_df.status.unique()
```

```
[17]: array(['Employee', 'Detailee', 'Employee '], dtype=object)
```

Status only has Employees and Detailees, we just need to strip it real quick to fix the 'Employee'.

```
[18]: wh_df.status = wh_df.status.str.strip()
```

```
[19]: wh_df.status.unique()
```

```
[19]: array(['Employee', 'Detailee'], dtype=object)
```

We do not need to do anything further to Status as it only contains two unique values, which is pretty easy to work with.

Position Title on the other hand has 215 unique values. A lot of people hold multiple titles, like Michael J. Ambrosini (second entry), who is the SPECIAL ASSISTANT TO THE PRESIDENT AND DIRECTOR OF THE OFFICE OF THE CHIEF OF STAFF. It is hard to work with this many unique values, so it would be beneficial to creatue more features that can be used to group people with similar titles together.

```
[20]: # commented out because it takes up a lot of space #wh_df.position_title.unique()
```

```
[21]: def add_roles(df, role):

"""

Takes in the White House Salaries DataFrame and searches from a specified

→role

in Position Title. Adds a column to the DataFrame with a 1 if the person

→holds

the specified role and 0 if not.

Parameters

—————

df: DataFrame

Holds salary information about White House Staff.

role: str

Role being searched for.

"""

# checks if the role being searched for is in the Position Title

role_df = df.position_title.apply(lambda title: 1 if role in title else 0)

role_df.rename("{}".format(role.lower().replace(' ', '_')), inplace = True)

# adds it to the original DataFrame
```

```
df = pd.concat([df, role_df], axis = 1)
          return df
      wh_df = add_roles(wh_df, 'ADVISOR')
      wh_df = add_roles(wh_df, 'ASSISTANT')
      wh_df = add_roles(wh_df, 'ASSOCIATE')
      wh_df = add_roles(wh_df, 'CHIEF')
      wh_df = add_roles(wh_df, 'DEPUTY')
      wh df = add roles(wh df, 'DIRECTOR')
      wh_df = add_roles(wh_df, 'EXECUTIVE')
      wh_df = add_roles(wh_df, 'FIRST LADY')
      wh_df = add_roles(wh_df, 'PRESIDENT')
      wh_df = add_roles(wh_df, 'PRESS')
      wh_df = add_roles(wh_df, 'SENIOR')
      wh_df = add_roles(wh_df, 'SPECIAL')
[22]:
     wh_df.head()
[22]:
                            name
                                     status
                                               salary
                                                        pay_basis
      0
           Alexander, Monica K.
                                              56000.0
                                                       Per Annum
                                  Employee
          Ambrosini, Michael J.
      1
                                  Employee
                                              95000.0
                                                       Per Annum
      2
                  Amin, Stacy C.
                                  Employee
                                             140000.0
                                                       Per Annum
      3
           Andersen, Whitney N.
                                   Employee
                                              94000.0
                                                        Per Annum
         Anderson, Alexander J.
                                   Employee
                                              77000.0
                                                       Per Annum
                                              position_title
                                                               advisor
                                                                         assistant
      0
                                         EXECUTIVE ASSISTANT
                                                                      0
                                                                                 1
        SPECIAL ASSISTANT TO THE PRESIDENT AND DIRECTO...
                                                                    0
      1
                                                                               1
         SPECIAL ASSISTANT TO THE PRESIDENT AND ASSOCIA...
      2
                                                                    0
                                                                               1
      3 DEPUTY DIRECTOR OF OPERATIONS FOR THE WHITE HO...
                                                                    0
                                                                               0
                             DIRECTOR OF DIGITAL ENGAGEMENT
      4
                                                           first lady
         associate
                     chief
                            deputy
                                    director
                                               executive
                                                                        president
      0
                 0
                         0
                                 0
                                            0
                                                        1
                                                                     0
                                                                                0
      1
                 0
                         1
                                 0
                                            1
                                                        0
                                                                     0
                                                                                1
      2
                  1
                         0
                                 0
                                            0
                                                        0
                                                                     0
                                                                                1
      3
                  0
                         0
                                 1
                                            1
                                                        0
                                                                     0
                                                                                0
      4
                  0
                         0
                                                        0
                                                                                0
                senior
                         special
         press
      0
             0
                      0
                               0
                      0
      1
             0
                               1
      2
             0
                      0
                               1
      3
                      0
                               0
             0
      4
             0
                      0
                               0
```

This creates an easy way to identify and group employees with similar roles or seniority. It could be refined to where it creates better groupings, but this is a good start.

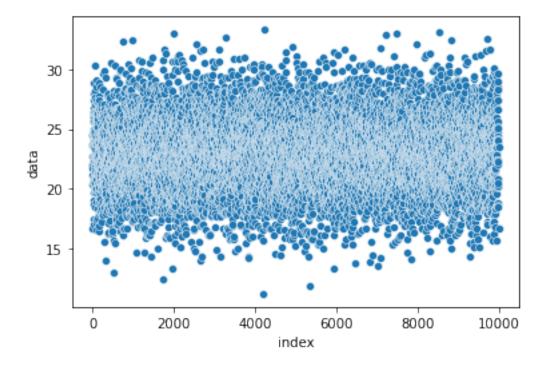
2.3 Question 3

Consider the file "data.csv" in the following GitHub repository. What are some descriptive statistics about this set? What can you say about the distribution of this data? https://github.com/fractalbass/data_engineer

```
[23]: df = pd.read_csv('https://raw.githubusercontent.com/fractalbass/data_engineer/
       →master/data.csv')
[24]: df.rename(columns = {'0': 'index', '23.82729036706873': 'data'}, inplace = True)
      df.set_index('index', inplace = True)
[25]: df.head()
[25]:
                  data
      index
      1
             23.589108
      2
             24.529556
      3
             16.684580
      4
             22.075507
      5
             21.892569
     df.describe()
[26]:
[26]:
                    data
             9999.000000
      count
      mean
               23.035996
      std
                2.995436
               11.248686
      min
      25%
               21.007170
               23.024031
      50%
      75%
               25.069370
      max
               33.286533
[27]: df.data.median()
[27]: 23.024031115748198
[28]: df.data.skew()
[28]: 0.00013644111255966445
     Coefficient of variance:
[29]: df.data.std() / df.data.mean()
[29]: 0.1300328245260554
```

```
[30]: import seaborn as sns sns.scatterplot(x = 'index', y = 'data', data = df)
```

[30]: <AxesSubplot:xlabel='index', ylabel='data'>



There are 9999 datapoints that are uniformly distributed. Looking at the scatter plot, most values are between 15 and 30.

The data has a small coefficient of variance and is very, very slightly positively skewed.

2.4 Question 4

2.4.1 A. If you were asked to impute null values in a column of a file that was 365 Gigabytes, what would you do? What tools would you use? What tools would you NOT use?

I do not have any experience working with files of this size, but I know I would need something beyond just Python or R, and will likely need something that uses SQL. A tool like Apache Spark seems like it would be a good tool for working with such a large file. It can also be used with a variety of languages.

2.4.2 B. What would you do if you were asked to do the above task every Thursday morning at 2:00am?

I would create either a BASH or Python script that would automate the task.

2.4.3 Question 5 Who is your favorite mathematician, statistician or computer scientist and why?

He is not that famous, but my favorite computer scientist is Mike Pound who is a professor at Nottingham University. The reason I like him is because of the videos he has done for the YouTube channel Computerphile. His videos are always very informative but still understanable and entertaining.