## 01-SF Salaries Exercise

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# 1 Assignment - 3

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### 2 SF Salaries Exercise

Welcome to a quick exercise for you to practice your pandas skills! We will be using the SF Salaries Dataset from Kaggle! Just follow along and complete the tasks outlined in bold below. The tasks will get harder and harder as you go along.

\*\* Import pandas as pd.\*\*

```
[1]: import pandas as pd
```

\*\* Read Salaries.csv as a dataframe called sal.\*\*

```
[2]: sal = pd.read_csv('Salaries.csv')
```

\*\* Check the head of the DataFrame. \*\*

```
[3]: sal.head()
```

[3]:	: Id Emplo		mployeeName				JobTitle \			
	0	1	NAT	HANIEL FORD	GENERAL MAN	AGER-METRO	POLITAN TRA	NSIT AUTHORITY		
	1	2	G	ARY JIMENEZ		CAPTAI	N III (POLI	CE DEPARTMENT)		
	2	3	ALB	ERT PARDINI	I CAPTAIN III (POLICE DEPARTMENT)					
	3	3 4 CHRISTOPHER CHONG			WIRE ROPE CABLE MAINTENANCE MECHANIC					
	4	5 PATRICK GARDNER		DEPUTY CHIEF OF DEPARTMENT, (FIRE DEPARTMENT)						
		В	asePay	OvertimePay	OtherPay	Benefits	TotalPay	TotalPayBenefits	\	
	0	167	411.18	0.00	400184.25	NaN	567595.43	567595.43		
	1	155	966.02	245131.88	137811.38	NaN	538909.28	538909.28		
	2	212	739.13	106088.18	16452.60	NaN	335279.91	335279.91		
	3	77	916.00	56120.71	198306.90	NaN	332343.61	332343.61		
	4	134	401.60	9737.00	182234.59	NaN	326373.19	326373.19		

```
Year
        Notes
                        Agency
                                 Status
   2011
0
           NaN
                 San Francisco
                                    NaN
1 2011
           NaN
                 San Francisco
                                    NaN
2 2011
                 San Francisco
           NaN
                                    NaN
3 2011
           NaN
                San Francisco
                                    NaN
4 2011
                San Francisco
           {\tt NaN}
                                    NaN
```

#### [4]: sal.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 148654 entries, 0 to 148653
Data columns (total 13 columns):

#	Column	Non-Null Count	Dtype
0	Id	148654 non-null	int64
1	EmployeeName	148654 non-null	object
2	JobTitle	148654 non-null	object
3	BasePay	148045 non-null	float64
4	OvertimePay	148650 non-null	float64
5	OtherPay	148650 non-null	float64
6	Benefits	112491 non-null	float64
7	TotalPay	148654 non-null	float64
8	${\tt TotalPayBenefits}$	148654 non-null	float64
9	Year	148654 non-null	int64
10	Notes	0 non-null	float64
11	Agency	148654 non-null	object
12	Status	0 non-null	float64
	47+ (1(0) :	+ (1 (0) - 1 + (2)	

dtypes: float64(8), int64(2), object(3)

memory usage: 14.7+ MB

#### What is the average BasePay?

- [5]: sal['BasePay'].mean()
- [5]: np.float64(66325.44884048769)
  - \*\* What is the highest amount of OvertimePay in the dataset? \*\*
- [6]: sal['OvertimePay'].max()
- [6]: np.float64(245131.88)
  - \*\* What is the job title of JOSEPH DRISCOLL? Note: Use all caps, otherwise you may get an answer that doesn't match up (there is also a lowercase Joseph Driscoll). \*\*
- [7]: sal[sal['EmployeeName'] == 'JOSEPH DRISCOLL']['JobTitle']

<sup>\*\*</sup> Use the .info() method to find out how many entries there are.\*\*

```
CAPTAIN, FIRE SUPPRESSION
 [7]: 24
      Name: JobTitle, dtype: object
     ** How much does JOSEPH DRISCOLL make (including benefits)? **
 [8]: sal[sal['EmployeeName'] == 'JOSEPH DRISCOLL']['TotalPayBenefits']
 [8]: 24
            270324.91
      Name: TotalPayBenefits, dtype: float64
     ** What is the name of highest paid person (including benefits)?**
 [9]: sal[sal['TotalPayBenefits'] == sal['TotalPayBenefits'].max()]['EmployeeName']
 [9]: 0
           NATHANIEL FORD
      Name: EmployeeName, dtype: object
     ** What is the name of lowest paid person (including benefits)? Do you notice something strange
     about how much he or she is paid?**
[10]: sal[sal['TotalPayBenefits'] == sal['TotalPayBenefits'].min()]['EmployeeName']
[10]: 148653
                 Joe Lopez
      Name: EmployeeName, dtype: object
     ** What was the average (mean) BasePay of all employees per year? (2011-2014)? **
[11]: sal.groupby('Year')['BasePay'].mean()
[11]: Year
      2011
              63595.956517
      2012
              65436.406857
      2013
              69630.030216
      2014
              66564.421924
      Name: BasePay, dtype: float64
     ** How many unique job titles are there? **
[12]: sal['JobTitle'].nunique()
[12]: 2159
     ** What are the top 5 most common jobs? **
[13]: sal['JobTitle'].value_counts().head()
[13]: JobTitle
      Transit Operator
                                        7036
      Special Nurse
                                        4389
      Registered Nurse
                                        3736
      Public Svc Aide-Public Works
                                        2518
```

```
Police Officer 3
                                        2421
      Name: count, dtype: int64
     ** How many Job Titles were represented by only one person in 2013? (e.g. Job Titles with only
     one occurence in 2013?) **
[14]: sum(sal[sal['Year'] == 2013]['JobTitle'].value_counts() == 1)
[14]: 202
     ** How many people have the word Chief in their job title? (This is pretty tricky) **
[15]: | sum(sal['JobTitle'].str.contains('chief', case=False, na=False))
[15]: 627
     ** Bonus: Is there a correlation between length of the Job Title string and Salary? **
[16]: # Create a new column for job title length
      sal['title_len'] = sal['JobTitle'].apply(len)
      # Calculate correlation between title length and TotalPay
      correlation = sal[['title_len', 'TotalPay']].corr()
      print("Correlation Matrix:")
      print(correlation)
      print("\nInterpretation:")
      print(f"The correlation coefficient between job title length and total pay is ⊔
       \hookrightarrow{correlation.iloc[0,1]:.6f}")
      print("This indicates a very weak negative correlation, meaning there is ⊔
       ⇔virtually no relationship")
      print("between the length of a job title and the salary amount. Job title,
       ⇔length does not")
      print("predict or influence salary in any meaningful way.")
     Correlation Matrix:
                 title len TotalPay
                  1.000000 -0.015356
     title len
     TotalPay
                 -0.015356 1.000000
     Interpretation:
     The correlation coefficient between job title length and total pay is -0.015356
     This indicates a very weak negative correlation, meaning there is virtually no
```

#### 3 Great Job!

relationship

between the length of a job title and the salary amount. Job title length does

predict or influence salary in any meaningful way.