

Assignment for Data QA & QC Internship @ Datahut

Data Cleaning Task Instructions

I am provided with a dataset named messy_Data.csv. My task is to clean this dataset and ensure it is ready for analysis.

Following Steps were taken for the data cleaning process.

1. Load the Data:

- Load the dataset into a jupyter notebook.
- pd: read_csv used for this method

2. Inspect the Data:

- Understood the dimension
- understood the repetition in some columns to study the data patterns
- I feel there is some commonality between Unnamed 0 and ID
- Renamed Unnamed 0 to Serial No.

Handling the date format and Department Correction before handling the NULL values

3. Standardise Date Formats:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11000 entries, 0 to 10999
Data columns (total 8 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Serial No   11000 non-null  int64
1   ID          11000 non-null  object
2   Name        8667 non-null   object
3   Age         9253 non-null   float64
4   Email       9731 non-null   object
5   Join Date   8808 non-null   object
6   Salary      8761 non-null   float64
7   Department  8745 non-null   object
dtypes: float64(2), int64(1), object(5)
memory usage: 687.6+ KB
```

- After checking the df info() converted the data format datetime format (YYYY-MM-DD).

4. Correct Department Names:

- Corrected the Department name before handling the missing values by using check the unique department names that are mentioned.

5. Handle Missing Values:

- Almost all column have null values

```
Serial No    0
ID           0
Name        2333
Age         1747
Email       1269
Join Date   2192
Salary      2239
Department  2255
dtype: int64
```

- Handled: Name and age together being NULL in many places.
- Age and Salary are normally distributed. Filled the NULL with **Mean**
- Department NULL was filled with **Most frequent** value
- Join date was filled with **Median** Value

6. Remove Duplicates:

- There are some rows with same Serial No and ID. I have kept only 1 row of such data
- Same email id was repeated in more than 1 entry. Removed the duplicate and kept the first occurrence only

7. Correct Email Formats:

- Regular expression to make all email addresses follow a standard format (e.g., username@domain.com).
- Regular expression used: `r'^[w\.-]+@[a-zA-Z\d\.-]+\.(com|info|net|org|biz)$'`

8. Clean Name Fields:

- NULL names are dropped
- Found some name is having 3rd name. Trimmed out the 3rd name
- Found some name with title. Dropped the title and just kept the first name and Last name
- Found some extraneous words added towards the end of the Surname code added to remove those. Keeping all the remaining as same.
- Logic Followed to trim out the extraneous word is as follows:
 - List1 : Found all the Last name from `df['Name']`
 - List2: Found repeating Last names (more than 2) from `df['Name']`
 - List3 : keeping the List1 as it is but just if the last name matched the string in surname then replace the last name with the **valid surname**

This will remove the extraneous word from the **List1**.
- Some code to verify the correction:


```
df[df['Name'].str.contains('Taylordaughter', case=False)]
df[df['Name'].str.contains('Lamb', case=False)]
```

We can see difference between surnames in `df['Name']` and `df['Name New']`

9. Handle Salary Noise:

- Salary column was verified

```
df.describe()
```

	Serial No	Age	Join Date	Salary
count	5963.000000	5963.000000	5963	5963.000000
mean	5009.591648	54.122727	2022-03-11 22:24:07.727653632	90134.626250
min	0.000000	18.000000	2020-01-01 00:00:00	26233.921419
25%	2494.000000	37.000000	2021-07-18 12:00:00	63744.500000
50%	5046.000000	54.162650	2022-03-14 00:00:00	89903.243632
75%	7515.000000	71.000000	2022-11-05 00:00:00	116642.443828
max	9999.000000	90.000000	2024-06-12 00:00:00	176156.206747
std	2887.244181	20.443719	NaN	33127.155457

- Mean and Median(50%) is approximately same. Giving a normally distributed effect
- Minimum value is 26233.921419 and Maximum value is 176156.206747 which does not show any visible noise presence.

10. Save the dataset

- Data save to cleaned_dataset.csv