

clean_spell

November 2, 2021

1 Data Intern Challenge

1.0.1 Written by Okiri Albert Cleaning Text Columns Based on Closeness of Spelling

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

1.0.2 Load excel spreadsheet

```
[2]: df = pd.read_excel('copy_data_intern_challenge.xlsx', sheet_name="data")
df[:10]
```

```
[2]:
```

	api_names	brand_sold_product	avg_price	count
0	Artemether / Lumefantrine	NaN	113.636364	11.0
1	Artemether / Lumefantrine	6T	200.000000	3.0
2	Artemether / Lumefantrine	96	0.000000	3.0
3	Artemether / Lumefantrine	ACT	67.000000	30.0
4	Artemether / Lumefantrine	ACT AL	50.000000	66.0
5	Artemether / Lumefantrine	ACTM	78.174557	2122.0
6	Artemether / Lumefantrine	AJANTA	124.545455	77.0
7	Artemether / Lumefantrine	AJANTA AL	60.000000	7.0
8	Artemether / Lumefantrine	AJANTA AL S	84.444444	18.0
9	Artemether / Lumefantrine	AJANTA _VOUTURE PROGRAM	50.000000	1.0

```
[3]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 141 entries, 0 to 140
Data columns (total 4 columns):
#   Column                Non-Null Count  Dtype
---  -
0   api_names             141 non-null   object
1   brand_sold_product    140 non-null   object
2   avg_price             141 non-null   float64
3   count                 141 non-null   float64
dtypes: float64(2), object(2)
memory usage: 4.5+ KB
```

1.1 List of correct brand_sold_products

```
[4]: brand_sold_product = df['brand_sold_product']
```

```
[5]: clean_list = {'UNNAMED', '6T', '96', 'ACT', 'ACTM', 'AJANTA', 'AJANTA VOUTURE',
    ↪ 'PROGRAM', 'A-L ',
    ↪ 'AL 25KG', 'ALFANTRINE', 'ALL BRANDS', 'LUFENART', 'LUMERAX DT', 'AL',
    ↪ 'ORIGINAL', 'PRO FANTRINE FORTE',
    ↪ 'AL S ', 'ALS FALCIZE', 'AL SYUP', 'ANTIMALARIAL', 'ARTEFAN', 'ARTEFAN',
    ↪ 'DISPERSIBLE', 'ARTEFENTRINE',
    ↪ 'ARTELAT', 'ARTEMETHER',
    ↪ 'LUMEFANTRINE', 'ARTEMETHER', 'BLISS', 'CARTER', 'CIPLA', 'COARINET', 'COARTEM',
    ↪ 'COARTESIANE', 'CO CORITHER', 'CO MALATHER', 'CO_MAX', 'COMAX',
    ↪ 'DPS', 'COMBIART', 'CROWN', 'DISPERSABLE',
    ↪ 'DOSE', 'ERITHER', 'FALCIZED', 'GAME', 'GENERIC', 'GENERIC APICA',
    ↪ 'LTD', 'GENERIC IPICA LABORATORIES',
    ↪ 'GENERIC S', 'GVITHER', 'IPCA', 'L',
    ↪ 'ARTEM', 'LEMERAX', 'LONART', 'LONEX', 'LONART DUSPERSIBLE', 'LORNAT',
    ↪ 'LOYALTY', 'LUFANATE', 'LUFENART', 'LUFENART', 'LUMARTEM', 'LUMARTEM',
    ↪ 'DT', 'LUMARTEM FORTE', 'LUMATEM',
    ↪ 'LUMATERM', 'LUMATERM DT', 'LUMEFAC', 'LUMERAX', 'LUMERAX',
    ↪ 'DT', 'LUMERAX DT S', 'LUMESoft PLUS', 'LUMET',
    ↪ 'LUMET ADULTS', 'LUMETHER DRY', 'LUMET',
    ↪ 'S', 'LUMIART', 'LUMITER', 'LUMITER DT', 'MACLEODS', 'MALARATE',
    ↪ 'MALAREM', 'MALBETA', 'MALODAR', 'NOVARTIS', 'NOVATIS', 'ORANGE',
    ↪ 'FLAVOUR', 'P ALAXIN', 'PRO FANTRINE FORTE',
    ↪ 'SHAL ARTEM', 'SHAL ARTEM FORTE', 'SHAL ARTEM',
    ↪ 'TEMRIN', 'TEMTRIN', 'VETENARY AL GLOVES', 'WINART',
    ↪ 'WINART FORTE'}
```

1.1.1 Fill in Missing Values in brand_sold_product Column

```
[6]: df['brand_sold_product'] = df['brand_sold_product'].fillna('UNNAMED')
```

```
[7]: df.head()
```

```
[7]:
```

	api_names	brand_sold_product	avg_price	count
0	Artemether / Lumefantrine	UNNAMED	113.636364	11.0
1	Artemether / Lumefantrine	6T	200.000000	3.0
2	Artemether / Lumefantrine	96	0.000000	3.0
3	Artemether / Lumefantrine	ACT	67.000000	30.0
4	Artemether / Lumefantrine	ACT AL	50.000000	66.0

1.1.2 Spelling Function to Match With Closeness of Spelling

```
[8]: # Spelling function
def clean_spell(df):
    return difflib.get_close_matches(df, clean_list, n=4, cutoff=0.48)[0]
```

```
[9]: df['corrected_brand_product'] = df['brand_sold_product'].apply(clean_spell)
```

```
[10]: df[:10]
```

```
[10]:
```

	api_names	brand_sold_product	avg_price	count	\
0	Artemether / Lumefantrine	UNNAMED	113.636364	11.0	
1	Artemether / Lumefantrine	6T	200.000000	3.0	
2	Artemether / Lumefantrine	96	0.000000	3.0	
3	Artemether / Lumefantrine	ACT	67.000000	30.0	
4	Artemether / Lumefantrine	ACT AL	50.000000	66.0	
5	Artemether / Lumefantrine	ACTM	78.174557	2122.0	
6	Artemether / Lumefantrine	AJANTA	124.545455	77.0	
7	Artemether / Lumefantrine	AJANTA AL	60.000000	7.0	
8	Artemether / Lumefantrine	AJANTA AL S	84.444444	18.0	
9	Artemether / Lumefantrine	AJANTA _VOUTURE PROGRAM	50.000000	1.0	


```

corrected_brand_product
0      UNNAMED
1          6T
2          96
3          ACT
4          ACT
5          ACTM
6          AJANTA
7          AJANTA
8          AJANTA
9  AJANTA VOUTURE PROGRAM

```

1.1.3 Get Similarity Score

```
[11]: # Define the function that Scores the spelling:
def spell_diff(row):
    return difflib.SequenceMatcher(None, row['brand_sold_product'],
    ↪row['corrected_brand_product']).ratio()
```

```
[12]: df['score'] = df.apply(spell_diff, axis=1)
```

```
[13]: df[:20]
```

```
[13]:
```

	api_names	brand_sold_product	avg_price	count	\
0	Artemether / Lumefantrine	UNNAMED	113.636364	11.0	

1	Artemether / Lumefantrine	6T	200.000000	3.0
2	Artemether / Lumefantrine	96	0.000000	3.0
3	Artemether / Lumefantrine	ACT	67.000000	30.0
4	Artemether / Lumefantrine	ACT AL	50.000000	66.0
5	Artemether / Lumefantrine	ACTM	78.174557	2122.0
6	Artemether / Lumefantrine	AJANTA	124.545455	77.0
7	Artemether / Lumefantrine	AJANTA AL	60.000000	7.0
8	Artemether / Lumefantrine	AJANTA AL S	84.444444	18.0
9	Artemether / Lumefantrine	AJANTA_VOUTURE PROGRAM	50.000000	1.0
10	Artemether / Lumefantrine	AJANTA VOUTURE PROGRAM	73.043478	23.0
11	Artemether / Lumefantrine	AL	32.248713	9960.0
12	Artemether / Lumefantrine	A L	35.810635	126.0
13	Artemether / Lumefantrine	AL 25KG	50.410959	73.0
14	Artemether / Lumefantrine	AL 3	36.551724	58.0
15	Artemether / Lumefantrine	AL 6	47.142857	35.0
16	Artemether / Lumefantrine	A L AL	96.666667	3.0
17	Artemether / Lumefantrine	AL AL	100.000000	3.0
18	Artemether / Lumefantrine	AL DISPERSABLE	55.000000	2.0
19	Artemether / Lumefantrine	ALFANTRINE	252.371795	78.0

	corrected_brand_product	score
0	UNNAMED	1.000000
1	6T	1.000000
2	96	1.000000
3	ACT	1.000000
4	ACT	0.666667
5	ACTM	1.000000
6	AJANTA	1.000000
7	AJANTA	0.800000
8	AJANTA	0.705882
9	AJANTA VOUTURE PROGRAM	0.977778
10	AJANTA VOUTURE PROGRAM	1.000000
11	A-L	0.666667
12	A-L	0.571429
13	AL 25KG	1.000000
14	A-L	0.750000
15	A-L	0.750000
16	A-L	0.600000
17	A-L	0.666667
18	DISPERSABLE	0.880000
19	ALFANTRINE	1.000000

1.1.4 Get Total Amount from Average Price and Counts of Brands Sold

```
[14]: df['total_amount'] = df['avg_price'] * df['count']
```

```
[15]: df_sum = pd.DataFrame()
```

1.1.5 Group the corrected_brand_name_product Column Entries and Sum their total_amount and count

```
[16]: df_sum['total_revenue_brand'] = df.groupby('corrected_brand_product').
      ↪total_amount.apply(lambda g: g.sum())
```

```
[17]: df_sum['total_count_brand'] = df.groupby('corrected_brand_product')['count'].
      ↪apply(lambda g: g.sum())
```

1.1.6 Get the Average Price

```
[18]: df_sum['average_price'] = df_sum['total_revenue_brand'] /
      ↪df_sum['total_count_brand']
```

1.1.7 Round the Decimal Values for Currency (2) and Total counts

```
[19]: df_sum = df_sum.round({'average_price': 2, 'total_count_brand': 0,
      ↪'total_revenue_brand': 2})
```

```
[20]: df_sum = df_sum.reset_index()
```

```
[21]: df_sum
```

```
[21]:   corrected_brand_product  total_revenue_brand  total_count_brand \
0                6T            600.00                3.0
1                96             0.00                3.0
2                A-L       330069.32       10185.0
3                ACT        5310.00          96.0
4                ACTM      165886.41       2122.0
..                ...                ...
83            TEMTRIN         770.04          52.0
84            UNNAMED        1250.00          11.0
85  VETENARY AL GLOVES         59.00           4.0
86            WINART        2280.00          19.0
87      WINART FORTE      31715.52       217.0

      average_price
0          200.00
1           0.00
2          32.41
3          55.31
4          78.17
..           ...
83          14.81
84         113.64
85          14.75
86         120.00
```

87 146.15

[88 rows x 4 columns]

1.1.8 Write New Excel File for Submission

```
[22]: df_sum.to_excel("cleaned.xlsx", index=False, sheet_name='data')
```

```
[ ]:
```