

Python 3.11.7 | packaged by Anaconda, Inc. | (main, Dec 15 2023, 18:05:47) [MSC v.1916 64 bit (AMD64)]

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IPython 8.20.0 -- An enhanced Interactive Python.

```
In [1]: import numpy as np
...: import pandas as pd
```

```
In [2]: # Originating Pieces files - merge these two
```

```
In [3]: OriginatingPieces1 = pd.read_csv("C:/Users/andie/OneDrive/Documents/Capstone/
Mail_v2/Piece Output v2/Originating Pieces pt.1 v2.csv")
```

```
In [4]: OriginatingPieces2 = pd.read_csv("C:/Users/andie/OneDrive/Documents/Capstone/
Mail_v2/Piece Output v2/Originating Pieces pt.2 v2.csv")
```

```
In [5]: #----- Filter out data by ACTUAL_DLVRY_DATE that isn't between 01/08/24 to
01/21/24
```

```
In [6]: op_january = OriginatingPieces1[(OriginatingPieces1['ACTUAL_DLVRY_DATE'] >
"2024-01-08") & (OriginatingPieces1['ACTUAL_DLVRY_DATE'] < "2024-01-21")]
```

```
In [7]: op2_january = OriginatingPieces2[(OriginatingPieces2['ACTUAL_DLVRY_DATE'] >
"2024-01-08") & (OriginatingPieces2['ACTUAL_DLVRY_DATE'] < "2024-01-21")]
```

```
In [8]: # Clean the two data sets
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```
In [9]: noNull_op1 = op_january.dropna(how='any',axis=0)
```

```
In [10]: clean_op1 = noNull_op1.drop_duplicates()
```

```
In [11]: noNull_op2 = op2_january.dropna(how='any',axis=0)
```

```
In [12]: clean_op2 = noNull_op2.drop_duplicates()
```

```
In [13]: # Add columns that determine if the mail was On Time Exactly, Early, or Late
```

```
In [14]: clean_op1['OnTimeExactly'] = np.where(clean_op1['ACTUAL_DLVRY_DATE'] ==
clean_op1['EXPECTED_DELIVERY_DATE'], True, False)
```

```
In [15]: clean_op1['Early'] = np.where(clean_op1['ACTUAL_DLVRY_DATE'] <
clean_op1['EXPECTED_DELIVERY_DATE'], True, False)
```

```
In [16]: clean_op1['Late'] = np.where(clean_op1['ACTUAL_DLVRY_DATE'] >
clean_op1['EXPECTED_DELIVERY_DATE'], True, False)
```

```
In [17]: clean_op2['OnTimeExactly'] = np.where(clean_op2['ACTUAL_DLVRY_DATE'] ==
clean_op2['EXPECTED_DELIVERY_DATE'], True, False)
```

```
In [18]: clean_op2['Early'] = np.where(clean_op2['ACTUAL_DLVRY_DATE'] <
clean_op2['EXPECTED_DELIVERY_DATE'], True, False)
```

```
In [19]: clean_op2['Late'] = np.where(clean_op2['ACTUAL_DLVRY_DATE'] >
clean_op2['EXPECTED_DELIVERY_DATE'], True, False)
```

```
In [20]: # Combine the datasets with only data between Jan 8th and Jan 21st of 2024
```

```
In [21]: originatingPieces_stormPeriod = pd.concat([clean_op1, clean_op2])
```

```
In [22]: # Compare the delivery status of the mail (Late vs. Ontime / Early)
```

```
In [23]: originating_late = originatingPieces_stormPeriod['Late'].values.sum()
```

```

In [24]: originating_early = originatingPieces_stormPeriod['Early'].values.sum()

In [25]: originating_ontime = originatingPieces_stormPeriod['OnTimeExactly'].values.sum()

In [26]: print("Late mail: " + str(originating_late) + ", Early Mail: " +
str(originating_early) + ", On Time Mail: " + str(originating_ontime))
Late mail: 2755583, Early Mail: 17054971, On Time Mail: 8634603

In [27]: # ratio of late vs rest

In [28]: originating_late_ratio = originating_late / (originating_early +
originating_ontime)

In [29]: print("Ratio of Late Mail vs. Early or On Time mail: " + str("%f" %
originating_late_ratio))
Ratio of Late Mail vs. Early or On Time mail: 0.107265

In [30]: # distribution of mail shape by lateness

In [31]: originatingPieces_stormPeriod['MAIL_SHAPE'].value_counts()
Out[31]:
MAIL_SHAPE
Letter    25793548
Flat      2376958
Card       274651
Name: count, dtype: int64

In [32]: op_shape = originatingPieces_stormPeriod.groupby(by=["MAIL_SHAPE", "Late"]).size()
# True = Late, False = either Early or On Time Exactly

In [33]: print(op_shape)
MAIL_SHAPE  Late
Card        False    172241
            True     102410
Flat        False    2235735
            True     141223
Letter      False    23281598
            True     2511950
dtype: int64

In [34]: # Mail Class

In [35]: originatingPieces_stormPeriod['MAIL_CLASS'].value_counts()
Out[35]:
MAIL_CLASS
USPS Marketing Mail    17134116
First Class Presort    6163077
Single Piece First Class 4599518
Periodicals            548446
Name: count, dtype: int64

In [36]: op_class = originatingPieces_stormPeriod.groupby(by=["MAIL_CLASS", "Late"]).size()
# True = Late, False = either Early or On Time Exactly

In [37]: print(op_class)
MAIL_CLASS
First Class Presort  False    5190341
                   True     972736
Periodicals         False    521124
                   True     27322
Single Piece First Class False    3788133
                   True     811385

```

```
USPS Marketing Mail      False    16189976
                        True      944140
dtype: int64
```

```
In [38]: #ORIGIN_FACILITY
```

```
In [39]: originatingPieces_stormPeriod['ORIGIN_FACILITY'].value_counts()
```

```
Out[39]:
ORIGIN_FACILITY
NASHVILLE - 1441275      13046409
MEMPHIS NDC - 1372672     7163871
MEMPHIS - 1441274        6071953
MUSIC CITY ANNEX - 1532174 2162924
Name: count, dtype: int64
```

```
In [40]: op_originFacility = originatingPieces_stormPeriod.groupby(by=["ORIGIN_FACILITY",
"Late"]).size() # True = Late, False = either Early or On Time Exactly
```

```
In [41]: print(op_originFacility)
ORIGIN_FACILITY      Late
MEMPHIS - 1441274    False    4807851
                        True     1264102
MEMPHIS NDC - 1372672  False    6520111
                        True     643760
MUSIC CITY ANNEX - 1532174 False    2053387
                        True     109537
NASHVILLE - 1441275    False   12308225
                        True     738184
dtype: int64
```

```
In [42]: # Convert delivery date columns to the 'date' data type
```

```
In [43]: originatingPieces_stormPeriod['ACTUAL_DLVRY_DATE'] =
pd.to_datetime(originatingPieces_stormPeriod['ACTUAL_DLVRY_DATE'])
```

```
In [44]: print(originatingPieces_stormPeriod['ACTUAL_DLVRY_DATE'].head())
0    2024-01-16
1    2024-01-13
2    2024-01-16
3    2024-01-16
4    2024-01-16
Name: ACTUAL_DLVRY_DATE, dtype: datetime64[ns]
```

```
In [45]: originatingPieces_stormPeriod['EXPECTED_DELIVERY_DATE'] =
pd.to_datetime(originatingPieces_stormPeriod['EXPECTED_DELIVERY_DATE'])
```

```
In [46]: # Difference between EXPECTED and ACTUAL delivery dates - Positive values indicate
LATE deliveries
```

```
In [47]: originatingPieces_stormPeriod['Difference'] =
(originatingPieces_stormPeriod['ACTUAL_DLVRY_DATE'] -
originatingPieces_stormPeriod['EXPECTED_DELIVERY_DATE']).dt.days
```

```
In [48]: # Average days late a piece of mail arrives
```

```
In [49]: late_deliveries =
originatingPieces_stormPeriod.loc[originatingPieces_stormPeriod.Late]
```

```
In [50]: latemean = late_deliveries['Difference'].mean()
```

```
In [51]: print("The mean for mail delivered late is: " + str("%f" % latemean) + " days after
expected delivery date")
The mean for mail delivered late is: 1.955629 days after expected delivery date
```

```
In [52]: # Differences and their Count by Mail Class
```

```
In [53]: late_by_class = late_deliveries.groupby(by=["MAIL_CLASS", "Difference"]).size() #  
True = Late, False = either Early or On Time Exactly
```

```
In [54]: print(late_by_class)
```

MAIL_CLASS	Difference	
First Class Presort	1	457482
	2	105264
	3	220840
	4	140535
	5	39176
	6	5482
	7	2202
	8	965
	9	663
	10	113
	11	14
Periodicals	1	18051
	2	5506
	3	1653
	4	1110
	5	440
	6	238
	7	140
	8	118
	9	60
	10	6
Single Piece First Class	1	366897
	2	127668
	3	170358
	4	86932
	5	31786
	6	14106
	7	7623
	8	3876
	9	1619
	10	520
USPS Marketing Mail	1	694905
	2	156839
	3	52583
	4	21396
	5	7277
	6	5131
	7	4004
	8	1443
	9	562

```
dtype: int64
```

```
In [55]: # Differences and their Count by Mail Shape
```

```
In [56]: late_by_shape = late_deliveries.groupby(by=["MAIL_SHAPE", "Difference"]).size() #  
True = Late, False = either Early or On Time Exactly
```

```
In [57]: print(late_by_shape)
```

MAIL_SHAPE	Difference	
Card	1	29796
	2	13202
	3	26948
	4	27079
	5	3062
	6	1110

	7	1033
	8	155
	9	23
	10	2
Flat	1	76032
	2	28747
	3	19968
	4	9575
	5	3503
	6	1333
	7	1412
	8	493
	9	148
	10	12
Letter	1	1431507
	2	353328
	3	398518
	4	213319
	5	72114
	6	22514
	7	11524
	8	5754
	9	2733
	10	625
	11	14

dtype: int64

In [58]: