04 Merged Data Pretraining

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0.1 Predicting Airline Delays

Notebook: Data Preparation Notebook

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/opt/conda/lib/python3.7/site-packages/secretstorage/dhcrypto.py:16:
CryptographyDeprecationWarning: int_from_bytes is deprecated, use int.from_bytes instead
 from cryptography.utils import int_from_bytes
/opt/conda/lib/python3.7/site-packages/secretstorage/util.py:25:
CryptographyDeprecationWarning: int_from_bytes is deprecated, use int.from_bytes instead
 from cryptography.utils import int_from_bytes
Requirement already satisfied: numpy in /opt/conda/lib/python3.7/site-packages
(1.21.5)
WARNING: Running pip as the 'root' user can result in broken permissions
and conflicting behaviour with the system package manager. It is recommended to use a virtual environment instead: https://pip.pypa.io/warnings/venv

/opt/conda/lib/python3.7/site-packages/secretstorage/dhcrypto.py:16: CryptographyDeprecationWarning: int_from_bytes is deprecated, use int.from_bytes

```
instead
```

from cryptography.utils import int_from_bytes /opt/conda/lib/python3.7/site-packages/secretstorage/util.py:25: CryptographyDeprecationWarning: int_from_bytes is deprecated, use int.from_bytes instead from cryptography.utils import int_from_bytes Requirement already satisfied: pandas in /opt/conda/lib/python3.7/site-packages (1.3.5)Requirement already satisfied: pytz>=2017.3 in /opt/conda/lib/python3.7/sitepackages (from pandas) (2021.3) Requirement already satisfied: python-dateutil>=2.7.3 in /opt/conda/lib/python3.7/site-packages (from pandas) (2.8.1) Requirement already satisfied: numpy>=1.17.3 in /opt/conda/lib/python3.7/sitepackages (from pandas) (1.21.5) Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.7/sitepackages (from python-dateutil>=2.7.3->pandas) (1.14.0) WARNING: Running pip as the 'root' user can result in broken permissions and conflicting behaviour with the system package manager. It is recommended to use a virtual environment instead: https://pip.pypa.io/warnings/venv

```
[2]: # INGEST Merged DATA

s3_client = boto3.client("s3")

BUCKET='ads-508-airline'
KEY='merged/Dec_modeling.csv'

response = s3_client.get_object(Bucket=BUCKET, Key=KEY)
dec_merged = pd.read_csv(response.get("Body"))
dec_merged.head()
```

[2]:		DAY_01	F_MONT	H DAY_OF_	WEEK OP_	UNIQUE	_CARRIER	TAIL	_NUN	origin	_AIRPORT_ID	\
	0			8	7		WN	N8	651	A	15016	
	1			8	7		WN	N9	39W1	J	15016	
	2			8	7		WN	N7	7410	C	15016	
	3			8	7		WN	N5	50W1	1	15016	
	4			8	7		WN	N8	319I	7	15016	
		ORIGIN	DEST	DEP_DEL15	DEP_TIM	E_BLK	ARR_TIME_	BLK		\		
	0	STL	SAN	0.0	1100	-1159	1300-1	1359	•••			
	1	STL	SAT	0.0	1200	-1259	1400-1	1459				
	2	STL	SAT	0.0	2100	-2159	0001-0)559	•••			
	3	STL	SEA	0.0	0900	-0959	1200-1	1259	•••			
	4	STL	SFO	1.0	1800	-1859	2000-2	2059	•••			

```
CARRIER_NAME PILOTS_COPILOTS PASSENGER_HANDLING \
O Southwest Airlines Co.
                                                        9668
                                     8989
1 Southwest Airlines Co.
                                     8989
                                                        9668
2 Southwest Airlines Co.
                                     8989
                                                        9668
3 Southwest Airlines Co.
                                     8989
                                                        9668
4 Southwest Airlines Co.
                                     8989
                                                        9668
  PASS_GEN_SVC_ADMIN MAINTENANCE PRCP
                                              SNWD TMAX AWND
                                        SNOW
0
                                               0.0 58.0 9.84
               15475
                             2482 0.02
                                         0.0
1
               15475
                             2482 0.02
                                         0.0
                                               0.0 58.0 9.84
2
                             2482 0.02
                                         0.0
                                               0.0 58.0 9.84
               15475
3
               15475
                             2482 0.02
                                          0.0
                                               0.0 58.0 9.84
               15475
                             2482 0.02
                                         0.0
                                               0.0 58.0 9.84
```

[5 rows x 25 columns]

[3]: dec_merged.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 558026 entries, 0 to 558025
Data columns (total 25 columns):

#	Column	Non-Null Count	Dtype
0	DAY_OF_MONTH	558026 non-null	l int64
1	DAY_OF_WEEK	558026 non-null	l int64
2	OP_UNIQUE_CARRIER	558026 non-null	Lobject
3	TAIL_NUM	558026 non-null	Lobject
4	ORIGIN_AIRPORT_ID	558026 non-null	l int64
5	ORIGIN	558026 non-null	Lobject
6	DEST	558026 non-null	Lobject
7	DEP_DEL15	558026 non-null	l float64
8	DEP_TIME_BLK	558026 non-null	Lobject
9	ARR_TIME_BLK	558026 non-null	Lobject
10	CANCELLED	558026 non-null	l float64
11	CRS_ELAPSED_TIME	558026 non-null	l float64
12	DISTANCE	558026 non-null	l float64
13	DISTANCE_GROUP	558026 non-null	l int64
14	AIRLINE_ID	558026 non-null	l int64
15	CARRIER_NAME	558026 non-null	L object
16	PILOTS_COPILOTS	558026 non-null	l int64
17	PASSENGER_HANDLING	558026 non-null	l int64
18	PASS_GEN_SVC_ADMIN	558026 non-null	l int64
19	MAINTENANCE	558026 non-null	l int64
20	PRCP	558026 non-null	l float64
21	SNOW	558026 non-null	l float64
22	SNWD	558026 non-null	l float64
23	TMAX	558026 non-null	l float64

24 AWND 558026 non-null float64

dtypes: float64(9), int64(9), object(7)

memory usage: 106.4+ MB

```
[4]: dec_merged.shape
```

[4]: (558026, 25)

1 Data Cleaning

All missing values were imputed or dropped, as described above. Since our data has been validated by the Bureau of Transportation Statistics and Climate Data Online, outliers were investigated by reviewing the summary statistics of our data set.

[5]:	<pre>dec_merged.describe()</pre>								
[5]:		DAY_OF_MONTH	DAY_OF_WEEK	ORIGIN_AIRPORT_ID	DEP_DEL15	\			
	count	558026.000000	558026.000000	558026.000000	_	•			
	mean	15.830902	3.938745	12666.002996					
	std	8.957760	2.085336	1514.187330					
	min	1.000000	1.000000	10140.000000	0.000000				
	25%	8.000000	2.000000	11292.000000	0.000000				
	50%	16.000000	4.000000	12889.000000	0.000000				
	75%	23.000000	6.000000	13931.000000	0.000000				
	max	31.000000	7.000000	15919.000000	1.000000				
		CANCELLED CRS	S_ELAPSED_TIME	DISTANCE DIS	STANCE_GROUP \				
	count	558026.0			8026.000000				
	mean	0.0	148.552937	843.568687	3.844704				
	std	0.0	74.475448	604.827406	2.372199	2.372199			
	min	0.0	34.000000	66.000000	1.000000				
	25%	0.0	94.000000	400.000000	2.000000				
	50%	0.0	130.000000	680.000000	3.000000				
	75%	0.0	179.000000	1075.000000	5.000000				
	max	0.0	705.000000	5095.000000	11.000000				
		AIRLINE_ID	PILOTS_COPILOTS	S PASSENGER_HANDL	ING PASS_GEN_SV	C_ADMIN	\		
	count	558026.000000	558026.000000	558026.000	0000 558026	000000			
	mean	19954.738880	6132.51844	7 7380.776	3432 9991	.061352			
	std	368.971181	3163.78316	5 5905.764	240 6417	.203879			
	min	19393.000000	586.000000	0.000	0000 154	.000000			
	25%	19790.000000	2444.000000	1407.000	0000 3592	2.000000			
	50%	19930.000000	7637.000000	8586.000	0000 15237	.000000			
	75%	20314.000000	8989.000000	9668.000	0000 15502	2.000000			
	max	20436.000000	9293.00000	16888.000	15809	0.000000			
		MAINTENANCE	PRCP	SNOW	SNWD \				

count	558026.000000	558026.000000	558026.000000	558026.000000
mean	3576.673642	0.116641	0.048059	0.121935
std	3092.215270	0.352309	0.347030	0.806783
min	34.000000	0.000000	0.000000	0.000000
25%	898.000000	0.000000	0.000000	0.000000
50%	2482.000000	0.000000	0.000000	0.000000
75%	6122.000000	0.040000	0.000000	0.000000
max	9677.000000	7.130000	13.300000	18.100000
	TMAX	AWND		
count	558026.000000	558026.000000		
mean	56.160668	8.137934		
std	14.612596	4.014022		
min	8.000000	0.000000		
25%	45.000000	4.920000		
50%	56.000000	7.610000		
75%	67.000000	10.510000		
max	87.000000	25.720000		

Evaluation of min and max values indicate that outliers do not exist. Also, thanks to the preprocessing of our datasets before merging, duplicates and formatting issues were already resolved.

1.1 1. Feature Selection

- We will disregard columns with IDs
- Also drop any redundant columns such as having only 1 distinct value
- \bullet Since we have both the distance and distance group column, it is redundant or DEP_TIME_BLK vs. ARR_TIME_BLK

[6]: dec_merged.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 558026 entries, 0 to 558025
Data columns (total 25 columns):

#	Column	Non-Null Count	Dtype
0	DAY_OF_MONTH	558026 non-null	int64
1	DAY_OF_WEEK	558026 non-null	int64
2	OP_UNIQUE_CARRIER	558026 non-null	object
3	TAIL_NUM	558026 non-null	object
4	ORIGIN_AIRPORT_ID	558026 non-null	int64
5	ORIGIN	558026 non-null	object
6	DEST	558026 non-null	object
7	DEP_DEL15	558026 non-null	float64
8	DEP_TIME_BLK	558026 non-null	object
9	ARR_TIME_BLK	558026 non-null	object
10	CANCELLED	558026 non-null	float64
11	CRS_ELAPSED_TIME	558026 non-null	float64

```
DISTANCE
                         558026 non-null
                                           float64
12
13
    DISTANCE_GROUP
                         558026 non-null
                                           int64
14
    AIRLINE_ID
                         558026 non-null
                                           int64
    CARRIER NAME
                                           object
15
                         558026 non-null
    PILOTS COPILOTS
                                           int64
16
                         558026 non-null
    PASSENGER HANDLING
                         558026 non-null
17
                                           int64
    PASS GEN SVC ADMIN
                         558026 non-null
                                           int64
19
    MAINTENANCE
                         558026 non-null
                                           int64
20
    PRCP
                         558026 non-null
                                           float64
21
    SNOW
                         558026 non-null
                                           float64
22
    SNWD
                         558026 non-null
                                           float64
                         558026 non-null
23
    TMAX
                                           float64
24
    AWND
                         558026 non-null
                                           float64
```

dtypes: float64(9), int64(9), object(7)

memory usage: 106.4+ MB

3

4

Of the 24 features remaining in our dataset, several were redundant upon inspection. ORI-GIN_AIRPORT_ID is redundant with ORIGIN. ARR_TIME_BLK, which represents the arrival time block, is redundant with the combination of DEP_TIME_BLK and CRS_ELAPSED_TIME, or the schedule length of the flight, and DISTANCE_GROUP, or the distance of the flight. AIRLINE_ID is redundant with CARRIER_NAME. DISTANCE is redundant with the DISTANCE_GROUP binned feature. Since CRS_ELAPSED_TIME would be a function of the distance traveled, it was also dropped. OP_UNIQUE_CARRIER is redundant with CARRIER_NAME, which is also easier to interpret in subsequent sections, therefore OP_UNIQUE_CARRIER will be dropped. Since all cancelled flights were removed, the CANCELLED feature is irrelevant. This left 17 features in our dataset. Some features, such as CARRIER NAME, ORIGIN, and DEST will not be used in modeling but are retained for later use.

```
[7]: dropped = ['TAIL NUM', 'ORIGIN AIRPORT ID', |
      →'ARR_TIME_BLK', 'CANCELLED', 'AIRLINE_ID', 'DISTANCE', 'CRS_ELAPSED_TIME', _
      →'OP UNIQUE CARRIER']
     dec_red = dec_merged.drop(dropped, axis=1)
     dec red.head()
[7]:
        DAY_OF_MONTH
                       DAY_OF_WEEK ORIGIN DEST
                                                  DEP_DEL15 DEP_TIME_BLK
     0
                                                        0.0
                    8
                                  7
                                       STL
                                            SAN
                                                                1100-1159
                    8
                                  7
     1
                                       STL
                                            SAT
                                                        0.0
                                                                1200-1259
                                  7
     2
                    8
                                       STL
                                            SAT
                                                        0.0
                                                                2100-2159
```

SEA

SFO

0900-0959

1800-1859

0.0

1.0

	DISTANCE_GROUP		CARRIER_NAME	PILOTS_COPILOTS	\
0	7	Southwest	Airlines Co.	8989	
1	4	Southwest	Airlines Co.	8989	
2	4	Southwest	Airlines Co.	8989	
3	7	Southwest	Airlines Co.	8989	
4	7	Southwest	Airlines Co.	8989	

STL

STL

7

7

8

8

```
PASSENGER_HANDLING PASS_GEN_SVC_ADMIN MAINTENANCE PRCP
                                                                   SNOW
                                                                         SNWD \
     0
                      9668
                                         15475
                                                             0.02
                                                                    0.0
                                                                          0.0
                                                       2482
     1
                      9668
                                         15475
                                                       2482 0.02
                                                                    0.0
                                                                          0.0
     2
                      9668
                                         15475
                                                       2482
                                                            0.02
                                                                    0.0
                                                                          0.0
     3
                      9668
                                                       2482 0.02
                                                                    0.0
                                                                          0.0
                                         15475
                      9668
                                         15475
                                                       2482 0.02
                                                                    0.0
                                                                          0.0
       TMAX AWND
     0 58.0 9.84
     1 58.0 9.84
     2 58.0 9.84
     3 58.0 9.84
     4 58.0 9.84
[8]: dec red.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 558026 entries, 0 to 558025
    Data columns (total 17 columns):
     #
         Column
                             Non-Null Count
                                              Dtype
     0
         DAY_OF_MONTH
                             558026 non-null
                                              int64
     1
         DAY_OF_WEEK
                             558026 non-null
                                              int64
     2
         ORIGIN
                             558026 non-null
                                              object
     3
         DEST
                             558026 non-null
                                              object
                                              float64
     4
         DEP_DEL15
                             558026 non-null
     5
         DEP_TIME_BLK
                             558026 non-null
                                              object
     6
                             558026 non-null
         DISTANCE_GROUP
                                              int64
     7
         CARRIER_NAME
                             558026 non-null
                                              object
     8
                             558026 non-null
                                              int64
         PILOTS_COPILOTS
         PASSENGER_HANDLING
                             558026 non-null int64
     10 PASS_GEN_SVC_ADMIN
                             558026 non-null int64
     11
        MAINTENANCE
                             558026 non-null int64
     12
        PRCP
                             558026 non-null float64
                             558026 non-null float64
         SNOW
     13
     14
                             558026 non-null float64
        SNWD
     15
         TMAX
                             558026 non-null float64
     16 AWND
                             558026 non-null float64
    dtypes: float64(6), int64(7), object(4)
    memory usage: 72.4+ MB
[9]: # Correlation Matrix for multicollinearity
     plt.figure(figsize=(24, 12))
     mask = np.triu(np.ones_like(dec_red.corr(), dtype=np.bool))
     heatmap = sns.heatmap(dec_red.corr(), mask=mask, vmin=-1, vmax=1, annot=True)
     # Make the full heat map visible
```

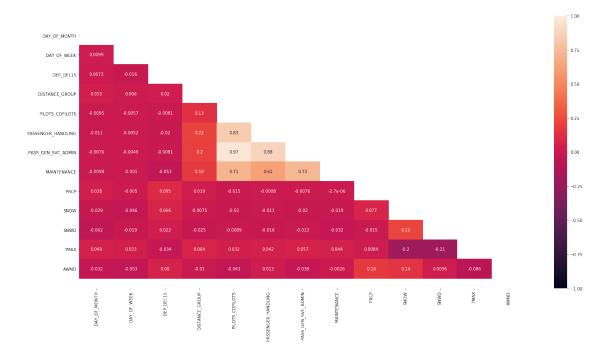
```
b, t = plt.ylim() # discover the values for bottom and top
b += 0.5 # Add 0.5 to the bottom
t -= 0.5 # Subtract 0.5 from the top
plt.ylim(b, t) # update the ylim(bottom, top) values
plt.show() # ta-da!
```

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:3:

DeprecationWarning: `np.bool` is a deprecated alias for the builtin `bool`. To silence this warning, use `bool` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.bool_` here.

Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations

This is separate from the ipykernel package so we can avoid doing imports until



1.2 2. Feature Creation

Since the day of the month did not show an easily identifiable pattern for departure delays, a new feature WEEK_OF_MONTH was created. This feature consists of 4 full (7-day) weeks and 1 partial (3-day) week. WEEK_OF_MONTH distributions for departure delay were evaluated to determine the optimal feature to use.

WEEK_OF_MONTH

```
1-7 = \text{Week } 1 = 1
8-14 = \text{Week } 2 = 2
```

```
15-21 = Week 3 = 3
       22-28 = \text{week } 4 = 4
       29-31 = partial week 5 = 5
     While DEP TIME BLK showed- REDUCE CARDINALITY OF DEPARTURE TIME BLOCKS
     (PREVIOUSLY 19 levels with inconsistent bucket size to 4 bins)
       Redeve/Early Departure = 12:01 a.m. - 5:59 a.m = 1
       Morning Departure = 6:00 \text{ a.m} - 11:59 \text{ a.m.} = 2
       Daytime Departure = 12:00 \text{ p.m.} - 5:59 \text{ p.m} = 3
       Late Departure = 6:00 p.m. - 11:59 p.m. = 4
     1.2.1 Create WEEK_OF_MONTH
[10]: def month_weeks_range(x):
          if x <= 7:
              return 1
          elif x \ll 14:
              return 2
          elif x <= 21:
              return 3
          elif x <= 28:
              return 4
          else:
              return 5
      dec_red['WEEK_OF_MONTH'] = dec_red['DAY_OF_MONTH'].apply(lambda x:_
       →month_weeks_range(x))
      dec_red['WEEK_OF_MONTH'].value_counts()
「10]: 3
           129144
      1
           128108
      2
           125664
      4
           122390
            52720
      Name: WEEK_OF_MONTH, dtype: int64
[11]: # Explore DAY OF MONTH with DEP_DEL15
      Week = pd.crosstab(dec_red['WEEK_OF_MONTH'], dec_red['DEP_DEL15'])
      Week['Total'] = Week.sum(axis=1)
      Week.loc['Total'] = Week.sum()
      Week['Percent_Delayed'] = ((Week.iloc[:,1])/((Week.iloc[:,0])+(Week.iloc[:,1])))
      Week = Week.sort_values('Percent_Delayed')
      Week
```

1.0

0.0

```
Total
                     441734 116292 558026
                                                    0.208399
      3
                              27508 129144
                     101636
                                                    0.213003
      1
                     100715
                              27393 128108
                                                    0.213827
      4
                      95691
                              26699 122390
                                                    0.218147
      5
                      40889
                              11831
                                      52720
                                                    0.224412
[12]: # Drop DAY_OF_MONTH in place of WEEK_OF_MONTH
      dec_red.drop(columns=['DAY_OF_MONTH'], inplace = True)
     1.2.2 Transform DEP_TIME_BLK
[13]: dep_blk = {'0600-0659':2, '0700-0759':2, '0800-0859':2,
                 '0900-0959':2,'1000-1059':2, '1100-1159':2,
                 '1200-1259':3, '1300-1359':3, '1400-1459':3,
                 '1500-1559':3, '1600-1659':3, '1700-1759':3,
                 '1800-1859':4, '1900-1959':4, '2000-2059':4,
                 '2100-2159':4, '2200-2259':4,
                 '2200-2259':4, '2300-2359':4, '0001-0559':1}
      dec_red['DEP_TIME_BLK'] = dec_red['DEP_TIME_BLK'].replace(dep_blk)
      dec_red['DEP_TIME_BLK'].value_counts()
[13]: 2
          213792
      3
           197393
      4
           131938
      1
            14903
      Name: DEP_TIME_BLK, dtype: int64
[14]: # Explore DEP TIME BLK with DEP DEL15
      DEP = pd.crosstab(dec_red['DEP_TIME_BLK'], dec_red['DEP_DEL15'])
      DEP['Total'] = DEP.sum(axis=1)
      DEP.loc['Total'] = DEP.sum()
      DEP['Percent_Delayed'] = ((DEP.iloc[:,1])/((DEP.iloc[:,0])+(DEP.iloc[:,1])))
      DEP = DEP.sort_values('Percent_Delayed')
      DEP
[14]: DEP_DEL15
                       0.0
                                     Total Percent_Delayed
                               1.0
     DEP_TIME_BLK
      1
                     13563
                              1340
                                    14903
                                                   0.089915
      2
                    183722
                             30070 213792
                                                   0.140651
      Total
                    441734 116292 558026
                                                   0.208399
      3
                    150278
                            47115 197393
                                                   0.238686
      4
                     94171
                             37767 131938
                                                   0.286248
```

1.3 3. Feature Transformation

Since the non-numeric features remaining in our dataset will not be ingested in the model, but used later to enhance findings, interpretations and recommendations, and we are using the XGBoost Classification algorithm for modeling purposes, further transformations are unnecessary. XGBoost is not sensitive to transformations or scaling of features in the same way that decision trees and random forest are not. By not scaling our features, we remove the need to scale subsequent data ingested by the model and facilitate easier interpretability of our model with real-world data.

```
[15]: # List remaining features and types dec_red.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 558026 entries, 0 to 558025
Data columns (total 17 columns):

#	Column	Non-Null Count	Dtype
0	DAY_OF_WEEK	558026 non-null	int64
1	ORIGIN	558026 non-null	object
2	DEST	558026 non-null	object
3	DEP_DEL15	558026 non-null	float64
4	DEP_TIME_BLK	558026 non-null	int64
5	DISTANCE_GROUP	558026 non-null	int64
6	CARRIER_NAME	558026 non-null	object
7	PILOTS_COPILOTS	558026 non-null	int64
8	PASSENGER_HANDLING	558026 non-null	int64
9	PASS_GEN_SVC_ADMIN	558026 non-null	int64
10	MAINTENANCE	558026 non-null	int64
11	PRCP	558026 non-null	float64
12	SNOW	558026 non-null	float64
13	SNWD	558026 non-null	float64
14	TMAX	558026 non-null	float64
15	AWND	558026 non-null	float64
16	WEEK_OF_MONTH	558026 non-null	int64
dtyp	es: float64(6), int6	4(8), object(3)	
	70 / I MD		

memory usage: 72.4+ MB

1.4 4. Balance the data set

```
df_balanced.reset_index(drop=True, inplace =True)
      df balanced
                                        DEP DEL15 DEP TIME BLK DISTANCE GROUP
[17]:
              DAY OF WEEK ORIGIN DEST
                         2
                              ATL
                                   DEN
                                               0.0
      0
                                                                                 5
      1
                         6
                              MSP
                                   FLL
                                               0.0
                                                                2
                                                                                 6
      2
                         5
                              OGG
                                   SFO
                                               0.0
                                                                4
                                                                                10
                         5
                              IAD
                                                                3
      3
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                                                                         9668
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                Delta Air Lines Inc.
                                                    9293
                                                                        15331
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               United Air Lines Inc.
                                                   7637
                                                                        16888
      3
                Delta Air Lines Inc.
                                                    9293
                                                                        15331
                Alaska Airlines Inc.
                                                    2893
                                                                         1062
      232579
                Alaska Airlines Inc.
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                  Mesa Airlines Inc.
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                Delta Air Lines Inc.
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              Southwest Airlines Co.
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              PASS_GEN_SVC_ADMIN
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                                                               1.2 29.0
                                                                           19.24
              WEEK_OF_MONTH
      0
                           5
      1
                           1
```

df_balanced = df_grouped_by.apply(lambda x: x.sample(df_grouped_by.size().

→min()))

```
2 1
3 2
4 2
... ...
232579 4
232580 3
232581 1
232582 4
232583 4
```

[232584 rows x 17 columns]

```
[18]: # Confirm the majority class was undersampled df_balanced['DEP_DEL15'].value_counts()
```

[18]: 0.0 116292 1.0 116292

Name: DEP_DEL15, dtype: int64

1.5 5. Split the data set

```
[19]: # partition model used features
model = df_balanced.drop(['ORIGIN','DEST', 'CARRIER_NAME'], axis = 1)

# Move DEP_DEL15 to first position for modeling
Dep = model['DEP_DEL15']
model.drop(labels=['DEP_DEL15'], axis=1,inplace = True)
model.insert(0, 'DEP_DEL15', Dep)
model
```

	model										
[19]:		DEP_DEL15	DAY_OF_W	EEK	DEP_TIME_BLK	DISTANCE_G	ROUP	PILO	TS_COP	ILOTS	\
	0	0.0		2	4		5			8989	
	1	0.0		6	2		6			9293	
	2	0.0		5	4		10			7637	
	3	0.0		5	3		10			9293	
	4	0.0		3	3		2			2893	
	•••	•••	•••		•••	•••		•••			
	232579	1.0		2	2		4			2893	
	232580	1.0		1	4		1			1312	
	232581	1.0		4	3		5			9293	
	232582	1.0		6	3		2			9293	
	232583	1.0		6	4		4			8989	
		PASSENGER_	HANDLING	PAS	S_GEN_SVC_ADMI	N MAINTENA	NCE	PRCP	SNOW	SNWD	\
	0		9668		1547	75 2	482	0.00	0.0	0.0	
	1		15331		1580	9 6	122	0.00	0.0	0.0	
	2		16888		1523	37 4	991	0.39	0.0	0.0	

```
0.0
      3
                           15331
                                                15809
                                                              6122 0.25
                                                                            0.0
      4
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                                                 5737
                                                               898
                                                                    0.54
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      232579
                            1062
                                                                    0.00
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                                                               898
      232580
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                                                                                  0.0
                           15331
                                                15809
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      232582
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              49.0
                     8.05
                                       2
      232579 40.0
                     4.70
                                       4
      232580 62.0
                     6.49
                                        3
                     2.91
      232581 69.0
                                        1
      232582 80.0 10.07
      232583 29.0 19.24
      [232584 rows x 14 columns]
[20]: # Split all data into 90% train and 10% holdout
      df_train, df_holdout = train_test_split(model, test_size=0.10,__

stratify=model['DEP_DEL15'])
      # Split holdout to 50% validation and 50% test
      df_val, df_test = train_test_split(df_holdout, test_size=0.50,__
       ⇔stratify=df_holdout['DEP_DEL15'])
     add code to upload to bucket
[21]: df_train.shape
[21]: (209325, 14)
[22]: df_val.shape
[22]: (11629, 14)
[23]: df_test.shape
[23]: (11630, 14)
```

```
[30]: # Save train data set for modeling to model data folder in bucket
      csv_buffer=StringIO()
      df_train.to_csv(csv_buffer, index=False)
      BUCKET_NAME = 'ads-508-airline'
      FileName= 'model_data/df_train.csv'
      s3csv = boto3.client('s3')
      response=s3csv.put_object(Body=csv_buffer.getvalue(),
                                 Bucket=BUCKET NAME,
                                 Key=FileName)
[31]: | # Save validation data set for modeling to model_data folder in bucket
      csv_buffer=StringIO()
      df_val.to_csv(csv_buffer, index=False)
      BUCKET NAME = 'ads-508-airline'
      FileName= 'model_data/df_val.csv'
      s3csv = boto3.client('s3')
      response=s3csv.put_object(Body=csv_buffer.getvalue(),
                                 Bucket=BUCKET_NAME,
                                 Key=FileName)
[32]: # Save test data set for modeling to model_data folder in bucket
      csv_buffer=StringIO()
      df_test.to_csv(csv_buffer, index=False)
      BUCKET NAME = 'ads-508-airline'
      FileName= 'model_data/df_test.csv'
      s3csv = boto3.client('s3')
      response=s3csv.put_object(Body=csv_buffer.getvalue(),
                                 Bucket=BUCKET_NAME,
                                 Key=FileName)
 []:
```