ProcessingV2

April 28, 2024

[26]:
```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
import scipy
import sklearn
import os
```

[27]:
```
def find_csv_files(folder_path, suffix=".csv"):
    """
    Find all CSV files in the given folder path.

    Args:
    - folder_path (str): Path to the folder where CSV files are located.
    - suffix (str): Suffix to filter files (default is '.csv').

    Returns:
    - list: List of CSV files with the given suffix.
    """
    csv_files = []
    for root, dirs, files in os.walk(folder_path):
        for file in files:
            if file.endswith(suffix):
                csv_files.append(os.path.join(root, file))
    return csv_files
```

[28]:
```
def plot_top_n_categorical(data, column, n=None, figsize=(15, 5)):
    """
    Plot the distribution of the top N categories of a categorical variable.

    Args:
    - data (pandas DataFrame): DataFrame containing the categorical variable.
    - column (str): Name of the categorical variable column.
    - n (int or None): Number of top categories to plot. If None, plot all␣
 ↪categories (default is None).
    - figsize (tuple): Width and height of the figure in inches (default is␣
 ↪(10, 6)).
```

```
    Returns:
    - None (displays the plot).
    """
    if n is None:
        categories = data[column].value_counts().index
    else:
        categories = data[column].value_counts().nlargest(n).index

    data_filtered = data[data[column].isin(categories)]

    plt.figure(figsize=figsize)
    sns.countplot(data=data_filtered, x=column, order=categories)
    plt.title(f'Top {len(categories)} Categories of {column}')
    plt.xlabel(column)
    plt.ylabel('Count')
    plt.xticks(rotation=90)
    plt.show()
```

[29]:
```
data = pd.read_csv("./INITIAL_PROCESSED_DATA.csv")
```

[30]:
```
data = data.sort_values(by=['Date (MM/DD/YYYY)', 'Scheduled Arrival Time'])
```

[31]:
```
to_drop = [
    'Arrival Delay (Minutes)',
    'Flight Number',
    'Tail Number',
]
```

[32]:
```
data = data.drop(columns=to_drop)
```

[33]:
```
weather_data_path = './imputed_weather/{}_weather_data.csv'
```

[34]:
```
data.head()
```

[34]:

	Carrier Code	Date (MM/DD/YYYY)	Origin Airport	Scheduled Arrival Time	Scheduled Elapsed Time (Minutes)	FLIGHT_STATUS	month	day	season	WeekDay
0	B6	2010-01-01	JFK	00:01	76	LATE	1	1	winter	Friday
1	B6	2010-01-01	JFK	08:55	75	LATE	1	1	winter	Friday
2	MQ	2010-01-01	ORD	11:20	100	ONTIME	1	1	winter	Friday
3	9E	2010-01-01	DTW	11:44	84	LATE	1	1	winter	Friday
4	B6	2010-01-01	JFK	11:52						

```
4                                71          LATE      1    1  winter  Friday
```

[35]:
```
data['Scheduled Arrival Time'] = data['Scheduled Arrival Time'].replace('24:
↪00', '23:59')
```

[36]:
```
data['UNIX_DATE'] = pd.to_datetime(data['Date (MM/DD/YYYY)'] + ' ' +␣
↪data['Scheduled Arrival Time'])
```

[37]:
```
data['UNIX_TIMESTAMP'] = data['UNIX_DATE'].apply(lambda x: int(x.timestamp()))
```

[38]:
```
data.head()
```

[38]:
```
  Carrier Code Date (MM/DD/YYYY) Origin Airport Scheduled Arrival Time  \
0           B6        2010-01-01            JFK                  00:01
1           B6        2010-01-01            JFK                  08:55
2           MQ        2010-01-01            ORD                  11:20
3           9E        2010-01-01            DTW                  11:44
4           B6        2010-01-01            JFK                  11:52

   Scheduled Elapsed Time (Minutes) FLIGHT_STATUS  month  day  season WeekDay  \
0                                76          LATE      1    1  winter  Friday
1                                75          LATE      1    1  winter  Friday
2                               100        ONTIME      1    1  winter  Friday
3                                84          LATE      1    1  winter  Friday
4                                71          LATE      1    1  winter  Friday

            UNIX_DATE  UNIX_TIMESTAMP
0 2010-01-01 00:01:00      1262304060
1 2010-01-01 08:55:00      1262336100
2 2010-01-01 11:20:00      1262344800
3 2010-01-01 11:44:00      1262346240
4 2010-01-01 11:52:00      1262346720
```

1 Weather

[39]:
```
def mode_imputer(data):
    """
    Perform mode imputation on a DataFrame.

    Parameters:
        data (DataFrame): Input DataFrame with missing values.

    Returns:
        DataFrame: DataFrame with missing values replaced by mode.
    """
    # Fill missing values with mode
```

```
    data_imputed = data.fillna(data.mode().iloc[0])

    return data_imputed
```

[40]:
```
all_weather_data = find_csv_files('./imputed_weather/')
```

[ ]:
```
```

[41]:
```
# for d in all_weather_data:
#     dd = mode_imputer(pd.read_csv(d))
#     print(f"============{d}===============")
#     dd = dd.fillna(1013)
#     print(dd.isna().sum())
#     print("==========================")
```

[42]:
```
weather1 = pd.read_csv(all_weather_data[0], parse_dates=['date'])
```

[43]:
```
weather1['UNIX_TIMESTAMP'] = weather1['date'].apply(lambda x: int(x.
↪timestamp()))
```

[44]:
```
weather1.head()
```

[44]:

	station	date	latitude	longitude	elevation	wind_direction	wind_type	wind_speed	ceiling_height	ceiling_det_code	celing_CAVOK	visibility_dist	visibility_variability	air_temparature	dew_point_temparature	sea_level_pressure	UNIX_TIMESTAMP
0	MSP	2009-01-01 00:00:00	44.8831	-93.2289	265.8	140.0	N	36.0	3309.428571	9	N	16000.0	N	-156.0	-211.0	10274.0	1230768000
1	MSP	2009-01-01 00:53:00	44.8831	-93.2289	265.8	140.0	N	36.0	3658.000000	M	N	16093.0	N	-156.0	-206.0	10255.0	1230771180
2	MSP	2009-01-01 01:53:00	44.8831	-93.2289	265.8	150.0	V	31.0	3048.000000	M	N	16093.0	N	-133.0	-194.0	10237.0	1230774780
3	MSP	2009-01-01 02:29:00	44.8831	-93.2289	265.8	140.0	N	62.0	3353.000000	M	N	16093.0	N	-130.0			
4	MSP	2009-01-01 02:53:00	44.8831	-93.2289	265.8	160.0	N	46.0	3048.000000	M	N	16093.0	N	-128.0			

```
3                -190.0             10204.8       1230776940
4                -189.0             10217.0       1230778380
```

[45]:
```python
def binary_search_nearest_rows(data, target_timestamp, num_neighbors=3):
    """
    Perform binary search to find the nearest rows in a DataFrame for a given␣
 ↪timestamp.

    Parameters:
        data (DataFrame): Input DataFrame with timestamps.
        target_timestamp (int): Target timestamp for which the nearest rows␣
 ↪need to be found.
        num_neighbors (int): Number of nearest rows to find on each side of the␣
 ↪target timestamp.

    Returns:
        DataFrame: DataFrame containing the rows corresponding to the nearest␣
 ↪timestamps.
    """
    # Sort DataFrame by the timestamp column
    data_sorted = data.sort_values(by='UNIX_TIMESTAMP')

    # Convert the sorted timestamps column to a list
    timestamps = data_sorted['UNIX_TIMESTAMP'].tolist()

    # Binary search to find the nearest timestamp
    low = 0
    high = len(timestamps) - 1

    while low <= high:
        mid = (low + high) // 2
        mid_timestamp = timestamps[mid]

        if mid_timestamp == target_timestamp:
            nearest_indices = [mid]
            break

        elif mid_timestamp < target_timestamp:
            low = mid + 1

        else:
            high = mid - 1

    else:
        # Find the nearest timestamp
        nearest_indices = []
        if high < 0:
```

```
            nearest_indices.append(low)
        elif low >= len(timestamps):
            nearest_indices.append(high)
        else:
            if abs(timestamps[low] - target_timestamp) < abs(timestamps[high] -␣
↪target_timestamp):
                nearest_indices.append(low)
            else:
                nearest_indices.append(high)

    # Ensure we have enough neighbors
    while len(nearest_indices) < num_neighbors:
        if nearest_indices[0] - 1 >= 0:
            nearest_indices.insert(0, nearest_indices[0] - 1)
        else:
            nearest_indices.append(nearest_indices[-1] + 1)

    # Extract the rows corresponding to the nearest timestamps
    nearest_rows = data_sorted.iloc[nearest_indices]

    return nearest_rows
```

[46]:
```
binary_search_nearest_rows(weather1, 1262346720)[['latitude', 'longitude',␣
↪'elevation',
      'wind_direction', 'wind_type', 'wind_speed', 'ceiling_height',
      'ceiling_det_code', 'celing_CAVOK', 'visibility_dist',
      'visibility_variability', 'air_temparature', 'dew_point_temparature',
      'sea_level_pressure', 'UNIX_TIMESTAMP']]
```

[46]:

	latitude	longitude	elevation	wind_direction	wind_type	wind_speed
12994	44.8831	-93.2289	265.8	300.0	N	31.0
12995	44.8831	-93.2289	265.8	320.0	N	36.0
12996	44.8831	-93.2289	265.8	310.0	N	21.0

	ceiling_height	ceiling_det_code	celing_CAVOK	visibility_dist
12994	366.0	M	N	16093.0
12995	610.0	M	N	16093.0
12996	701.0	M	N	16093.0

	visibility_variability	air_temparature	dew_point_temparature
12994	N	-170.0	-210.0
12995	N	-170.0	-210.0
12996	N	-170.0	-210.0

	sea_level_pressure	UNIX_TIMESTAMP
12994	10302.932552	1262345460
12995	10306.167846	1262345880

12996 10309.400485 1262346660

2 FINAL DATA PREP

[48]:
```
import tqdm
all_weather = {}
all_rows = []
SYR_DF = pd.read_csv(weather_data_path.format('SYR'), parse_dates=['date'])
SYR_DF['UNIX_TIMESTAMP'] = SYR_DF['date'].apply(lambda x: int(x.timestamp()))
for index, row in tqdm.tqdm_notebook(data.iterrows(), total=len(data)):
    station = row['Origin Airport']
    if station not in all_weather:
        all_weather[station] = pd.read_csv(weather_data_path.format(station),␣
 ↪parse_dates=['date'])
        all_weather[station]['UNIX_TIMESTAMP'] = all_weather[station]['date'].
 ↪apply(lambda x: int(x.timestamp()))
    weatherDF = all_weather[station]
    row = row.to_dict()
    row_unix_ts = row['UNIX_TIMESTAMP']

    closest_rows = binary_search_nearest_rows(weatherDF, row_unix_ts,␣
 ↪num_neighbors=5)[['latitude', 'longitude', 'elevation',
       'wind_direction', 'wind_type', 'wind_speed', 'ceiling_height',
       'ceiling_det_code', 'celing_CAVOK', 'visibility_dist',
       'visibility_variability', 'air_temparature', 'dew_point_temparature',
       'sea_level_pressure']]

    for nc in ['latitude', 'longitude', 'elevation',
       'wind_direction', 'wind_type', 'wind_speed', 'ceiling_height',
       'ceiling_det_code', 'celing_CAVOK', 'visibility_dist',
       'visibility_variability', 'air_temparature', 'dew_point_temparature',
       'sea_level_pressure']:
        if nc in ['wind_type',  'ceiling_det_code', 'celing_CAVOK',␣
 ↪'visibility_variability']:
            row[nc] = closest_rows[nc].mode().iloc[0]
        else:
            row[nc] = closest_rows[nc].mean()

    closest_rows = binary_search_nearest_rows(SYR_DF, row_unix_ts,␣
 ↪num_neighbors=5)[['latitude', 'longitude', 'elevation',
       'wind_direction', 'wind_type', 'wind_speed', 'ceiling_height',
       'ceiling_det_code', 'celing_CAVOK', 'visibility_dist',
       'visibility_variability', 'air_temparature', 'dew_point_temparature',
       'sea_level_pressure']]

    for nc in ['latitude', 'longitude', 'elevation',
```

```
       'wind_direction', 'wind_type', 'wind_speed', 'ceiling_height',
       'ceiling_det_code', 'celing_CAVOK', 'visibility_dist',
       'visibility_variability', 'air_temparature', 'dew_point_temparature',
       'sea_level_pressure']:
        if nc in ['wind_type',  'ceiling_det_code', 'celing_CAVOK',␣
 ↪'visibility_variability']:
            row['SYR_'+nc] = closest_rows[nc].mode().iloc[0]
        else:
            row['SYR_'+nc] = closest_rows[nc].mean()
    all_rows.append(row)
new_data_df = pd.DataFrame(all_rows, index=data.index)
```

```
/tmp/ipykernel_32272/3525575574.py:6: TqdmDeprecationWarning: This function will
be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
  for index, row in tqdm.tqdm_notebook(data.iterrows(), total=len(data)):

  0%|          | 0/113671 [00:00<?, ?it/s]
```

[49]:
```
new_data_df.head()
```

[49]:

	Carrier Code	Date (MM/DD/YYYY)	Origin Airport	Scheduled Arrival Time	\
0	B6	2010-01-01	JFK	00:01	
1	B6	2010-01-01	JFK	08:55	
2	MQ	2010-01-01	ORD	11:20	
3	9E	2010-01-01	DTW	11:44	
4	B6	2010-01-01	JFK	11:52	

	Scheduled Elapsed Time (Minutes)	FLIGHT_STATUS	month	day	season	WeekDay	\
0	76	LATE	1	1	winter	Friday	
1	75	LATE	1	1	winter	Friday	
2	100	ONTIME	1	1	winter	Friday	
3	84	LATE	1	1	winter	Friday	
4	71	LATE	1	1	winter	Friday	

	…	SYR_wind_type	SYR_wind_speed	SYR_ceiling_height	\
0	…	N	17.4	741.370199	
1	…	C	0.0	548.800000	
2	…	C	0.0	841.400000	
3	…	C	0.0	1006.000000	
4	…	C	0.0	1006.000000	

	SYR_ceiling_det_code	SYR_celing_CAVOK	SYR_visibility_dist	\
0	M	N	6110.0	
1	M	N	2414.0	
2	M	N	3138.4	
3	M	N	3138.4	
4	M	N	3138.4	

```
  SYR_visibility_variability  SYR_air_temparature  SYR_dew_point_temparature  \
0                          N                 -4.8                      -24.0
1                          N                -17.0                      -28.8
2                          N                -19.4                      -29.6
3                          N                -19.8                      -29.2
4                          N                -19.8                      -29.2

  SYR_sea_level_pressure
0           10154.572817
1           10132.352069
2           10129.796558
3           10129.482076
4           10129.482076

[5 rows x 40 columns]
```

[50]:
```
new_data_df.isna().sum()
```

[50]:
```
Carrier Code                          0
Date (MM/DD/YYYY)                     0
Origin Airport                        0
Scheduled Arrival Time                0
Scheduled Elapsed Time (Minutes)      0
FLIGHT_STATUS                         0
month                                 0
day                                   0
season                                0
WeekDay                               0
UNIX_DATE                             0
UNIX_TIMESTAMP                        0
latitude                              0
longitude                             0
elevation                             0
wind_direction                        0
wind_type                             0
wind_speed                            0
ceiling_height                        0
ceiling_det_code                      0
celing_CAVOK                          0
visibility_dist                       0
visibility_variability                0
air_temparature                       0
dew_point_temparature                 0
sea_level_pressure                  728
SYR_latitude                          0
SYR_longitude                         0
```

```
SYR_elevation                       0
SYR_wind_direction                  0
SYR_wind_type                       0
SYR_wind_speed                      0
SYR_ceiling_height                  0
SYR_ceiling_det_code                0
SYR_celing_CAVOK                    0
SYR_visibility_dist                 0
SYR_visibility_variability          0
SYR_air_temparature                 0
SYR_dew_point_temparature           0
SYR_sea_level_pressure              0
dtype: int64
```

[51]:
```
new_data_df['sea_level_pressure'] = new_data_df['sea_level_pressure'].
↪fillna(new_data_df['sea_level_pressure'].mean())
```

[52]:
```
new_data_df.isna().sum()
```

[52]:
```
Carrier Code                        0
Date (MM/DD/YYYY)                   0
Origin Airport                      0
Scheduled Arrival Time              0
Scheduled Elapsed Time (Minutes)    0
FLIGHT_STATUS                       0
month                               0
day                                 0
season                              0
WeekDay                             0
UNIX_DATE                           0
UNIX_TIMESTAMP                      0
latitude                            0
longitude                           0
elevation                           0
wind_direction                      0
wind_type                           0
wind_speed                          0
ceiling_height                      0
ceiling_det_code                    0
celing_CAVOK                        0
visibility_dist                     0
visibility_variability              0
air_temparature                     0
dew_point_temparature               0
sea_level_pressure                  0
SYR_latitude                        0
SYR_longitude                       0
```

```
SYR_elevation                       0
SYR_wind_direction                  0
SYR_wind_type                       0
SYR_wind_speed                      0
SYR_ceiling_height                  0
SYR_ceiling_det_code                0
SYR_celing_CAVOK                    0
SYR_visibility_dist                 0
SYR_visibility_variability          0
SYR_air_temparature                 0
SYR_dew_point_temparature           0
SYR_sea_level_pressure              0
dtype: int64
```

[60]:
```
# new_data_df.to_csv("SYR_ORIGIN_WEATHER_IMPUTED.csv", index=False)
```

[61]:
```
def create_data_2nd_model_data_ext(df, num_prev=3):
    extended_data = []
    for index, row in df.iterrows():
        for i in range(num_prev):
            if index - i - 1 >= 0:
                extended_row = row.copy()  # Create a copy of the current row
                extended_row['PREV_STAT'] = df.iloc[index - i -␣
 ↪1]['FLIGHT_STATUS']
                extended_data.append(extended_row)
            else:
                extended_row = row.copy()  # Create a copy of the current row
                extended_row['PREV_STAT'] = 'ONTIME'
                extended_data.append(extended_row)
    df_extended = pd.DataFrame(extended_data)
    return df_extended.reset_index(drop=True)
```

[62]:
```
create_data_2nd_model_data_ext(new_data_df).to_csv("SYR_ORIGIN_3HOP.csv",␣
 ↪index=False)
```

[63]:
```
create_data_2nd_model_data_ext(new_data_df, 1).to_csv("SYR_ORIGIN_1HOP.csv",␣
 ↪index=False)
```

[64]:
```
create_data_2nd_model_data_ext(new_data_df, 2).to_csv("SYR_ORIGIN_2HOP.csv",␣
 ↪index=False)
```

[ ]:
```

```