Data Cleaning and Preparation

```
In [49]: import pandas as pd
         import numpy as np
         import glob
         import re
         from bs4 import BeautifulSoup
         import requests
         import pickle
         import matplotlib.pyplot as plt
         import seaborn as sns
In [50]: # Load in T100 data from csv files
         files li = glob.glob('zippedData/T T100D*.2.csv')
         bts_df = pd.DataFrame(pd.read_csv(files_li[0]))
         for i in range(1,len(files li)):
             data = pd.read_csv(files_li[i])
             df1 = pd.DataFrame(data)
             bts_df = pd.concat([df1,bts_df],axis=0, ignore_index=True)
In [51]: # Keep only top 40 used aircraft models
         bts 40 = bts df['AIRCRAFT TYPE'].value counts().head(40).index.to list()
         indexNum = bts_df[~bts_df['AIRCRAFT_TYPE'].isin(bts_40)].index
         bts_df.drop(indexNum , inplace=True)
In [52]: # Get list of top 40 aircraft names for data cleaning
         bts_des = pd.DataFrame(pd.unique(bts_df['AIRCRAFT_TYPE']), columns =['AIRCRAFT_TYPE'])
In [53]: # BTS uses number code to identify aircraft type in it's flight data. The legend is in a seperate csv file.
         # Load in the legend for aircraft type.
         ac_df = pd.read_csv('zippedData/L_AIRCRAFT_TYPE.csv')
In [54]: # Merge flight data with aircfaft type. This will match BTS type names to the fligt data.
         bts_des = pd.merge(bts_des,ac_df,left_on='AIRCRAFT_TYPE',right_on='Code')
```

```
In [55]: # Drop piston aircraft
         i = bts_des[(bts_des['Code'] == 40)].index
         j = bts_des[(bts_des['Code'] == 35)].index
         k = bts des[(bts des['Code'] == 194)].index
         l = bts des[(bts des['Code'] == 79)].index
         m = bts des[(bts des['Code'] == 479)].index
         n = bts_des[(bts_des['Code'] == 415)].index
         o = bts des[(bts des['Code'] == 416)].index
         p = bts_des[(bts_des['Code'] == 412)].index
         bts des = bts des.drop(i)
         bts_des = bts_des.drop(j)
         bts des = bts des.drop(k)
         bts_des = bts_des.drop(l)
         bts_des = bts_des.drop(m)
         bts_des = bts_des.drop(n)
         bts des = bts des.drop(o)
         bts des = bts des.drop(p)
In [56]: common_words = ['BOEING','CHEROKEE', 'AIRBUS', 'INDUSTRIE', 'CANADAIR', 'EMBRAER', 'MCDONNELL', 'DOUGLAS',
                          'DE', 'HAVILLAND', 'BEAVER', 'SUPER', 'NAVAJO', 'EM', 'ER', 'BEECH']
         common_words_eng = ['ILYUSHIN', 'ANTONOV', 'JUNKERS', 'LOCKHEED', 'BOEING', 'AIRBUS', 'INDUSTRIE',
                              'CANADAIR', 'EMBRAER', 'STATIONAIR', 'MCDONNELL', 'DOUGLAS', 'DE', 'HAVILLAND', 'BEAVER',
                              'SUPER', 'NAVAJO', 'EM', 'ER', 'BEECH', 'CRAFT', 'BOMBARDIER', 'CANADA']
         common words emi = ['TRENT', 'CFMI', 'PRATT & WHITNEY', 'GENERAL ELECTRIC', 'CANADA',
                              'ROLLSROYCE', 'LYCOMING', 'ALLISON', 'GARRETT', 'CONTINENTAL', 'IAE']
In [57]: def short_name(name, x):
             if len(name) >= x:
                 short_name = name[:x]
                 return(short_name)
             return(name)
In [58]: def remove_dash(name):
             name = name.replace('-', '')
             return(name)
In [59]: def convert_upper(name):
             name = name.upper()
             return(name)
In [60]: def convert_date(date):
             date = date.astype('datetime64[ns]')
             return(date)
In [61]: def remove_slash(name):
             res = re.match(r''^{^{\prime\prime}}, name)
             return(str(res.group()))
```

```
In [62]: def remove common(name, common w):
             for common in common w:
                 name = name.replace(common, '')
             return(name)
In [63]: def remove_space(name):
             name = name.replace(' ', "")
             return(name)
In [64]:
         def convert boeing(name):
             if re.match(r''^{(7)}(.)(7)'', name):
                 name = name[:3]
             return name
In [65]: def convert cri(name):
             if re.match(r"^RJ", name):
                 name = name.replace('RJ', 'CRJ')
             return name
In [66]: # Change Descriptions for data cleaning and matching
         bts des.loc[bts des['Code'] == 629, 'Description'] = 'Canadair RJ-200'
         bts_des.loc[bts_des['Code'] == 631, 'Description'] = 'Canadair RJ-700'
         bts des.loc[bts des['Code'] == 638, 'Description'] = 'Canadair RJ-900'
         bts_des.loc[bts_des['Code'] == 838, 'Description'] = 'Boeing 737-800'
         bts_des.loc[bts_des['Code'] == 674, 'Description'] = 'Embraer ERJ-130'
         bts des.loc[bts des['Code'] == 676, 'Description'] = 'Embraer ERJ-140'
         bts des.loc[bts des['Code'] == 675, 'Description'] = 'Embraer ERJ-145'
         bts des.loc[bts des['Code'] == 677, 'Description'] = 'Embraer ERJ-170'
         bts_des.loc[bts_des['Code'] == 678, 'Description'] = 'Embraer ERJ-190'
In [67]: # Change Descriptions for data cleaning and matching
         bts_des['Description'] = [convert_upper(x) for x in bts_des['Description']]
         bts des['Description'] = [remove slash(x) for x in bts des['Description']]
         bts des['Description'] = [remove dash(x) for x in bts des['Description']]
         bts_des['Description'] = [remove_common(x, common_words) for x in bts_des['Description']]
         bts_des['Description'] = [remove_space(x) for x in bts_des['Description']]
         bts_des['Description'] = [short_name(x, 4) for x in bts_des['Description']]
         bts_des['Description'] = [convert_boeing(x) for x in bts_des['Description']]
         bts des['Description'] = [convert cri(x) for x in bts des['Description']]
In [68]: # Merge Flight data with cleaned aircraft names
         bts_df = pd.merge(bts_df,bts_des[['Code', 'Description']],left_on='AIRCRAFT_TYPE', right_on='Code')
In [69]: # Drop Code column as we don't need it anymore
         bts des.drop(['AIRCRAFT TYPE', 'Code'], axis=1, inplace=True)
```

Crash Data

```
In [70]: # Import Data Set from CSV
         crash df = pd.read csv('zippedData/AviationData.csv', encoding='latin-1', dtype={6: str, 7: str,
                                                                                          14: str, 15: str,
                                                                                          28: str})
In [71]: # Drop rows before 2012 so we can match data by year in bts df
         date mask = (crash df['Event.Date'] > '2011-12-31')
         # THIS RETURNS ALL ROWS GREATER THAN THE DATE PROVIDED ABOVE
         crash_df = crash_df.loc[date_mask]
         # Rows without an aircraft model type don't help use. Dropping these rows.
         crash df = crash df.dropna(subset=['Model'])
In [72]: # Convert Event.Date from string to datetime. Used to filter df on years
         crash_df['Event.Date'] = crash_df['Event.Date'].astype('datetime64[ns]')
In [73]: # Change Model name for data cleaning and matching
         crash_df['Model'] = [convert_upper(x) for x in crash_df['Model']]
         #crash_df['Event.Date'] = [convert_date(x) for x in crash_df['Event.Date']]
         crash df['Model'] = [remove slash(x) for x in crash df['Model']]
         crash df['Model'] = [remove dash(x) for x in crash df['Model']]
         crash df['Model'] = [remove common(x, common words) for x in crash df['Model']]
         crash_df['Model'] = [remove_space(x) for x in crash_df['Model']]
         crash_df['Model'] = [short_name(x, 4) for x in crash_df['Model']]
         crash_df['Model'] = [convert_boeing(x) for x in crash_df['Model']]
         #crash df['Model'].replace('PA-', 'PA', inplace=True, regex=True)
In [74]: # Takes in an aircraft model name and returns the cloest match in bts_des
         def match_types(model):
             for newmodel in bts des['Description']:
                 if newmodel.startswith(model):
                     #print(f'Processing: {model} \t:{newmodel}')
                     return newmodel
In [75]: # Create new column that matches model name form bts_des
         crash_df['NewModel'] = [match_types(model) for model in crash_df['Model']]
In [76]: # Drop Report.Status column as we don't need it and it conflicts with excel writer
         crash_df.drop(['Report.Status'], axis=1, inplace=True)
In [77]: | crash_lite_df = crash_df.dropna(subset=['NewModel']).copy()
```

Engine Data

```
In [78]: # In place of running this and making lots of web requests, import the pickle file from below
         # Assign URL |
         #url = "https://asn.flightsafety.org/database/engines/"
         # Make a GET request to fetch the raw HTML content
         #html_content = requests.get(url).text
         # Parse the response with html.parser
         #soup = BeautifulSoup(html_content,"html.parser")
         # The data we need are wrapped in <a> tags. Grab all the <a> tags here and we'll filter later.
         #datas = soup.find_all("a")
         #Make list of URLs and list oF Aircraft to create dictionary
         \# k = []
         # v = [1]
         # for item in datas:
               if 'engine' in item['href']:
                   k.append(item['href'])
                   v.append(item.text)
         # lookup_dict = {'URL': k, 'Engine': v}
         #url_df = pd.DataFrame.from_dict(lookup_dict)
         # with open('lookup_dict.pkl', 'wb') as f:
               pickle.dump(lookup_dict, f)
         with open('zippedData/lookup_dict.pkl', 'rb') as f:
             lookup_dict = pickle.load(f)
```

```
In [79]: # Take in the engine and retriev the matching data page with Aircraft data
         # Add Aircraft list from web a python list.
         # Return the python list
         # url p1 = 'https://asn.flightsafety.org'
         # def lookup_ac(url_p2):
               return list = []
               url = f'{url_p1}{url_p2}'
               sub_content = requests.get(url).text
               soup2 = BeautifulSoup(sub_content,"html.parser")
               u list=soup2.find all('ul')[1]
               i = 0
               while i < len(u_list.select('li')):</pre>
                   (u_list.select('li')[i].text)
                   return_list.append(u_list.select('li')[i].text)
                   i += 1
               return(return list)
         # Create Aircraft column to list aircraft for each engine.
         #url_df['Aircraft'] = url_df['URL'].map(lookup_ac)
         # Save df to pickle file.
         #url_df.to_pickle('zippedData/url_df.pkl')
         with open('zippedData/url_df.pkl', 'rb') as f:
             url df = pickle.load(f)
In [80]: # Create a row for each aircraft for each engine.
         # Iterate through url_df['Aircraft'] column. For each element
         # in list, create a new row.
         engines = []
         aircraft = []
         for idx, row in url_df.iterrows():
             for ac in row['Aircraft']:
                 engines.append(row['Engine'])
                 aircraft.append(ac)
In [81]: # Create dictionary from lists to load into dataFrame
         ac_eng_dict = {'Engine': engines, 'Aircraft': aircraft}
In [82]: # Create dataFrame from dictionary
         engines_df = pd.DataFrame.from_dict(ac_eng_dict)
```

```
In [83]: # Change Name for data cleaning and matching
         engines df.loc[engines df['Aircraft'] == 'Bombardier CRJ100 / 200 / 440'] = ['General Electric CF34','Canadair RJ-100']
         engines df.loc[engines df['Aircraft'] == 'Bombardier CRJ700'] = ['General Electric CF34', 'Canadair RJ-700']
         engines df.loc[engines df['Aircraft'] == 'Bombardier CRJ900'] = ['General Electric CF34', 'Canadair RJ-900']
         engines_df.loc[len(engines_df.index)] = ['General Electric CF34', 'Canadair RJ-200']
         engines df.loc[len(engines df.index)] = ['General Electric CF34', 'Canadair RJ-440']
         engines_df.loc[len(engines_df.index)] = ['General Electric CF34', 'Embraer ERJ-130']
         engines df.loc[len(engines df.index)] = ['General Electric CF34', 'Embraer ERJ-135']
         engines_df.loc[len(engines_df.index)] = ['General Electric CF34', 'Embraer ERJ-140']
         engines df.loc[len(engines df.index)] = ['General Electric CF34', 'Embraer ERJ-175']
         engines df.loc[len(engines df.index)] = ['General Electric CF34', 'Embraer ERJ-190']
         engines df.loc[len(engines df.index)] = ['Pratt & Whitney JT8D', 'McDonnell Douglas DC-9-80']
In [84]: # Change Aircraft name for data cleaning and matching
         engines_df['Aircraft'] = [convert_upper(x) for x in engines_df['Aircraft']]
         engines df['Aircraft'] = [remove slash(x) for x in engines df['Aircraft']]
         engines_df['Aircraft'] = [remove_dash(x) for x in engines_df['Aircraft']]
         engines df['Aircraft'] = [remove common(x, common words eng) for x in engines df['Aircraft']]
         engines df['Aircraft'] = [remove space(x) for x in engines df['Aircraft']]
         engines df['Aircraft'] = [short name(x, 4) for x in engines df['Aircraft']]
         engines df['Aircraft'] = [convert boeing(x) for x in engines df['Aircraft']]
         engines df['Aircraft'] = [convert cri(x) for x in engines df['Aircraft']]
In [85]: # Change Engine name for data cleaning and matching
         engines df['Engine Name'] = [convert upper(x) for x in engines df['Engine']]
         engines df['Engine Name'] = [remove slash(x) for x in engines df['Engine Name']]
         engines df['Engine Name'] = [remove dash(x) for x in engines df['Engine Name']]
         engines df['Engine Name'] = [remove common(x, common words emi) for x in engines df['Engine Name']]
         engines_df['Engine_Name'] = [remove_common(x, common_words_emi) for x in engines_df['Engine_Name']]
         engines_df['Engine_Name'] = [remove_space(x) for x in engines_df['Engine_Name']]
         engines df['Engine Name'] = [short name(x, 5) for x in engines <math>df['Engine Name']]
         \#engines\_df['Engine\_Name'] = [convert\_boeing(x) for x in engines\_df['Engine_Name']]
         #engines df['Engine Name'] = [convert cri(x) for x in engines <math>df['Engine Name']]
In [86]: # Add Engine data to bts des
         bts_des = pd.merge(bts_des, engines_df,left_on='Description', right_on='Aircraft', how='left')
         bts_des.drop_duplicates(inplace=True)
```

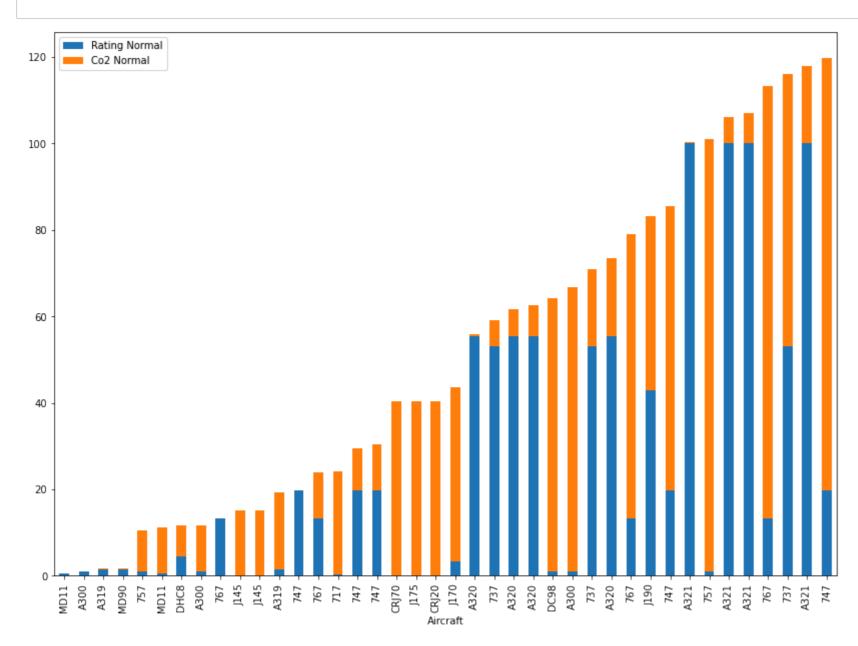
Emissions Data

```
In [88]: # Change Engine name for data cleaning and matching
         emissions_df['Engine Identification'] = [convert_upper(x) for x in emissions_df['Engine Identification']]
         emissions_df['Engine Identification'] = [remove_slash(x) for x in emissions df['Engine Identification']]
         emissions df['Engine Identification'] = [remove dash(x) for x in emissions df['Engine Identification']]
         emissions_df['Engine Identification'] = [remove_common(x, common_words_emi) for x in emissions_df['Engine Identification']]
         emissions df['Engine Identification'] = [remove space(x) for x in emissions <math>df['Engine Identification']]
         emissions df['Engine Identification'] = [short name(x, 5) for x in emissions <math>df['Engine Identification']]
In [89]: # Force name change for certain engines
         emissions_df.at[527,'Engine Identification'] = 'PW100'
         emissions_df.at[611, 'Engine Identification'] = 'PW200'
         emissions df.at[633, 'Engine Identification'] = 'PW400'
In [90]: # Takes in an Engine name and returns the CO2 value
         def match engine(str1):
             tmp df = pd.DataFrame()
             mask = tmp df['engine id starts with CFM'] = list(
                 map(lambda x: x.startswith(str1), emissions_df['Engine Identification']))
             filtered df = emissions df[mask]
             #return(filtered df.at[0,1]['CO Dp/Foo Avg (g/kN)']).head(1)
             if filtered df.shape[0] < 1:</pre>
                 return(0)
             return(filtered_df.iat[0,4])
In [91]: # Add Co2 data to bts des
         bts_des['Co2'] = [match_engine(x) for x in bts_des['Engine_Name']]
         Data Processing
```

```
In [92]: # Some rows have NaN for values. Replace these with 0 value.
         crash_df['Total.Fatal.Injuries'] = crash_df['Total.Fatal.Injuries'].fillna(0)
         crash_df['Total.Serious.Injuries'] = crash_df['Total.Serious.Injuries'].fillna(0)
         crash_df['Total.Minor.Injuries'] = crash_df['Total.Minor.Injuries'].fillna(0)
In [93]: # Our scoring system will use 3 points for fatality; 1 point for serious injury and .5 points for minor injury.
         crash_df['Injury_Score'] = crash_df['Total.Fatal.Injuries'].map(lambda x: x*3)
         crash df['Injury Score'] = crash df['Injury Score'] + crash df['Total.Serious.Injuries']
         crash_df['Injury_Score'] = crash_df['Injury_Score'] + (crash_df['Total.Minor.Injuries'] * .5)
```

```
In [94]: # Sum Injury Score and put into new temp df
         total_score = crash_df.groupby('NewModel')['Injury_Score'].sum().to_frame()
         total score['Count'] = bts df.groupby('Description').size()
         total_score['Rating'] = (total_score['Injury_Score'] / total_score['Count'])
         total_score = pd.merge(total_score, bts_des[['Co2', 'Aircraft', 'Engine']], left_on='NewModel',
                                right_on='Aircraft', how='left')
In [95]:
         final_score = total_score.sort_values(['Co2', 'Rating'], ascending=[True, True])
         # Drop engines for which data is unavailable
         final_score = final_score.loc[final_score['Co2'] > 0]
In [96]: # Normalize Injury Rating and Co2
         co2_max = final_score['Co2'].max()
         co2_min = final_score['Co2'].min()
         rating_max = final_score['Rating'].max()
         rating min = final score['Rating'].min()
         co2 range = co2 max - co2 min
         rating_range = rating_max - rating_min
         # normalized_value = (feature_value - feature_min) / (feature_max - feature_min)
         final score['Co2 Normal'] = [((x - co2 min) / (co2 range) * 100) for x in final score['Co2']]
         final_score['Rating Normal'] = [((x - rating_min) / (rating_range) * 100) for x in final_score['Rating']]
         final_score['Combined'] = final_score['Co2 Normal'] + final_score['Rating Normal']
         final_sorted = final_score.sort_values(by=['Combined'])
```

```
In [97]: plt.rcParams['figure.figsize'] = [14, 10]
final_sorted.plot(x='Aircraft', y=['Rating Normal', 'Co2 Normal'], kind='bar', stacked=True)
plt.show()
```



Airport Data Analysis

```
In [98]: # Load Airport data from csv file
ap_df = pd.read_csv('zippedData/us-airports.csv')
```

```
In [99]: # Get Population data from Wikipedia
          url = "https://en.wikipedia.org/wiki/List_of_North_American_metropolitan_areas_by_population"
          wiki df = pd.read html(url)[0]
          wiki_df = wiki_df[(wiki_df['Country'] == 'United States')]
In [100]: # Only keep columns we need. Drop other columns
          ap_df = ap_df[['id', 'ident', 'type', 'name', 'local_region', 'municipality','local_code',
                         'iata_code', 'keywords']]
          ap_df = ap_df[(ap_df['type'] == 'large_airport')]
In [101]: # Get count of flights originated at airport and convert to dict
          ap_counts = bts_df['ORIGIN'].value_counts()
          ap_counts_dict = ap_counts.to_dict()
In [102]: # This takes in an airport code and returns the total number of flights orginating from there
          def get_num_flt(code):
              try:
                val = ap_counts_dict[code]
              except:
                val = 0
              return val
In [103]: # Takes in city name and returns population
          def get_wiki_pop(city):
              try:
                  pop = wiki_df.loc[wiki_df['Metropolitan area'] == city, 'Population'].values[0]
                  return pop
              except:
                  return(0)
In [104]: # Create df to combine flight data and population
          apuse df = ap df.copy()
          apuse_df['NumFlights'] = [get_num_flt(x) for x in apuse_df['iata_code']]
          ap_mask = (apuse_df['NumFlights'] != 0)
          apuse_df = apuse_df.loc[ap_mask]
```

```
In [105]: # Group city and state airports and sum total flights from multiple airports
          result2 = apuse_df.groupby(['municipality', 'local_region'])
          # extract keys from groups
          keys = result2.groups.keys()
          # Add num of flights to city name
          city lst = []
          state lst = []
          num flt lst = []
          iata_lst = []
          for i in keys:
              a_group = result2.get_group(i)
              city_lst.append(a_group['municipality'][:1].to_string(index=False))
              state_lst.append(a_group['local_region'][:1].to_string(index=False))
              num_flt_lst.append(a_group['NumFlights'].sum())
              iata_lst.append(a_group['iata_code'][:1].to_string(index=False))
In [106]: # Add population data to result_df
          result2_dict = {'City': city_lst, 'State': state_lst, 'NumFlights': num_flt_lst, 'Code': iata_lst}
          result2 df = pd.DataFrame.from dict(result2 dict)
          pop lst = []
          for i, x in result2_df.iterrows():
              pop_lst.append(get_wiki_pop(x['City']))
          result2 df['Population'] = pop lst
          result2_df = result2_df.loc[result2_df['Population'] > 0]
In [107]: # Calculate flight per popuation ratio
          pfr_lst = []
          for i, x in result2_df.iterrows():
              ratio = x['Population']/x['NumFlights']
              pfr_lst.append(ratio)
          result2_df['FPR'] = pfr_lst
In [108]: result2_df = result2_df.sort_values('FPR', ascending=False)[:11]
```

In []:

```
In [109]: result2_df[['City', 'State', 'FPR']]
Out[109]:
                      City State
                                    FPR
                  Portland
                           ME 415.729577
           46
                           NY 259.439290
           35
                  New York
                           CA 200.556894
                Los Angeles
                            FL 171.359531
           30
                    Miami
           51
                Sacramento
                           CA 160.136403
                           TX 127.566186
                San Antonio
                           MA 127.011528
            5
                    Boston
                           OH 115.058592
                 Columbus
                Philadelphia
                            PA 114.615198
           21
                   Houston
                           TX 106.999753
                           CA 102.549896
           55 San Francisco
In [110]: # Group origin and destination pairs for Sankey diagram
          od_grp = bts_df.groupby(['ORIGIN', 'DEST']).size().sort_values(ascending=False)
          od_grp_df = od_grp.reset_index(name='count')
          # Limit origin to top 11 FPR airports
          od_grp_df = od_grp_df[od_grp_df['ORIGIN'].isin(result2_df['Code'][:11])]
          # Delete entries with less than 1000 trips
          od_grp_df.drop(od_grp_df[od_grp_df['count'] < 1000].index, inplace=True)
In [111]: # Write result2_df to excel for use in Tableau
          writer = pd.ExcelWriter('zippedData/tableau.xlsx')
           final_sorted.to_excel(writer, 'Sheet1')
           result2_df.to_excel(writer, 'Sheet2')
          crash_lite_df.to_excel(writer, 'Sheet3')
           od_grp_df.to_excel(writer, 'Sheet4')
          writer.close()
```