Project 1

October 16, 2021

[451]:
```
#Imports
from datetime import datetime
from datetime import timedelta
import requests, pandas as pd, numpy as np, time, datetime, matplotlib.pyplot␣
↪as plt, matplotlib.dates, math
from bs4 import BeautifulSoup
```

[ ]:
```

```

[452]:
```
#Part 1: Data Scraping and preparation
#Step 1: Scrape your competitor's data

r = requests.get('https://cmsc320.github.io/files/top-50-solar-flares.html')
root = BeautifulSoup(r.content)
root.prettify()
tables = pd.read_html('https://cmsc320.github.io/files/top-50-solar-flares.
↪html')
df_swl = tables[0]
df_swl = df_swl.rename(columns={"Unnamed: 0":"rank", "Unnamed: 1":
↪"x_classification",
                        "Unnamed: 2":"date", "Region":"region", "Start":
↪"start_time",
                       "Maximum":"maximum_time", "End":"end_time", "Unnamed: 7":
↪"movie"})
df_swl
```

[452]:

	rank	x_classification	date	region	start_time	maximum_time	\
0	1	X28+	2003/11/04	486	19:29	19:53	
1	2	X20+	2001/04/02	9393	21:32	21:51	
2	3	X17.2+	2003/10/28	486	09:51	11:10	
3	4	X17+	2005/09/07	808	17:17	17:40	
4	5	X14.4	2001/04/15	9415	13:19	13:50	
5	6	X10	2003/10/29	486	20:37	20:49	
6	7	X9.4	1997/11/06	8100	11:49	11:55	
7	8	X9.3	2017/09/06	2673	11:53	12:02	
8	9	X9	2006/12/05	930	10:18	10:35	
9	10	X8.3	2003/11/02	486	17:03	17:25	

10	11	X8.2	2017/09/10	2673	15:35	16:06
11	12	X7.1	2005/01/20	720	06:36	07:01
12	13	X6.9	2011/08/09	1263	07:48	08:05
13	14	X6.5	2006/12/06	930	18:29	18:47
14	15	X6.2	2005/09/09	808	19:13	20:04
15	16	X6.2	2001/12/13	9733	14:20	14:30
16	17	X5.7	2000/07/14	9077	10:03	10:24
17	18	X5.6	2001/04/06	9415	19:10	19:21
18	19	X5.4	2012/03/07	1429	00:02	00:24
19	20	X5.4	2005/09/08	808	20:52	21:06
20	21	X5.4	2003/10/23	486	08:19	08:35
21	22	X5.3	2001/08/25	9591	16:23	16:45
22	23	X4.9	2014/02/25	1990	00:39	00:49
23	24	X4.9	1998/08/18	8307	22:10	22:19
24	25	X4.8	2002/07/23	39	00:18	00:35
25	26	X4	2000/11/26	9236	16:34	16:48
26	27	X3.9	2003/11/03	488	09:43	09:55
27	28	X3.9	1998/08/19	8307	21:35	21:45
28	29	X3.8	2005/01/17	720	06:59	09:52
29	30	X3.7	1998/11/22	8384	06:30	06:42
30	31	X3.6	2005/09/09	808	09:42	09:59
31	32	X3.6	2004/07/16	649	13:49	13:55
32	33	X3.6	2003/05/28	365	00:17	00:27
33	34	X3.4	2006/12/13	930	02:14	02:40
34	35	X3.4	2001/12/28	9767	20:02	20:45
35	36	X3.3	2013/11/05	1890	22:07	22:12
36	37	X3.3	2002/07/20	39	21:04	21:30
37	38	X3.3	1998/11/28	8395	04:54	05:52
38	39	X3.2	2013/05/14	1748	00:00	01:11
39	40	X3.1	2014/10/24	2192	21:07	21:41
40	41	X3.1	2002/08/24	69	00:49	01:12
41	42	X3	2002/07/15	30	19:59	20:08
42	43	X2.8	2013/05/13	1748	15:48	16:05
43	44	X2.8	2001/12/11	9733	07:58	08:08
44	45	X2.8	1998/08/18	8307	08:14	08:24
45	46	X2.7	2015/05/05	2339	22:05	22:11
46	47	X2.7	2003/11/03	488	01:09	01:30
47	48	X2.7	1998/05/06	8210	07:58	08:09
48	49	X2.6	2005/01/15	720	22:25	23:02
49	50	X2.6	2001/09/24	9632	09:32	10:38

	end_time	movie
0	20:06	MovieView archive
1	22:03	MovieView archive
2	11:24	MovieView archive
3	18:03	MovieView archive
4	13:55	MovieView archive

```
5     21:01  MovieView archive
6     12:01  MovieView archive
7     12:10  MovieView archive
8     10:45  MovieView archive
9     17:39  MovieView archive
10    16:31  MovieView archive
11    07:26  MovieView archive
12    08:08  MovieView archive
13    19:00  MovieView archive
14    20:36  MovieView archive
15    14:35  MovieView archive
16    10:43  MovieView archive
17    19:31  MovieView archive
18    00:40  MovieView archive
19    21:17  MovieView archive
20    08:49  MovieView archive
21    17:04  MovieView archive
22    01:03  MovieView archive
23    22:28       View archive
24    00:47  MovieView archive
25    16:56  MovieView archive
26    10:19  MovieView archive
27    21:50       View archive
28    10:07  MovieView archive
29    06:49  MovieView archive
30    10:08  MovieView archive
31    14:01  MovieView archive
32    00:39  MovieView archive
33    02:57  MovieView archive
34    21:32  MovieView archive
35    22:15  MovieView archive
36    21:54  MovieView archive
37    06:13  MovieView archive
38    01:20  MovieView archive
39    22:13  MovieView archive
40    01:31  MovieView archive
41    20:14  MovieView archive
42    16:16  MovieView archive
43    08:14  MovieView archive
44    08:32       View archive
45    22:15  MovieView archive
46    01:45  MovieView archive
47    08:20  MovieView archive
48    23:31  MovieView archive
49    11:09  MovieView archive
```

Part 1 Step 1: Using a BeautifulSoup object, I get the HTML content from the cmsc320 backup

of the SpaceWeatherLive site. I then scrape the table from the site and convert it into a pandas DataFrame. I rename the columns for clarity,

[453]:
```
#Step 2: Tidy the top 50 solar flare data
df_swl = df_swl.drop(columns=['movie'])
for index, row in df_swl.iterrows():
    startdate_str = str(row['date']) + " " + str(row['start_time']) + ":00"
    maxdate_str = str(row['date']) + " " + str(row['maximum_time']) + ":00"
    enddate_str = str(row['date']) + " " + str(row['end_time']) + ":00"
    startdate_str = startdate_str.replace('/', '-')
    maxdate_str = maxdate_str.replace('/', '-')
    enddate_str = enddate_str.replace('/', '-')
    startdate_obj = datetime.datetime.strptime(startdate_str, '%Y-%m-%d %H:%M:
 ↪%S')
    maxdate_obj = datetime.datetime.strptime(maxdate_str, '%Y-%m-%d %H:%M:%S')
    enddate_obj = datetime.datetime.strptime(enddate_str, '%Y-%m-%d %H:%M:%S')
    df_swl.at[index, 'date'] = startdate_obj
    df_swl.at[index, 'start_time'] = startdate_obj
    df_swl.at[index, 'maximum_time'] = maxdate_obj
    df_swl.at[index, 'end_time'] = enddate_obj
    if df_swl.at[index, 'region'] == '-':
        df_swl.replace(to_replace = 'region', value = np.nan)
df_swl = df_swl.drop(columns=['date'])
df_swl
```

[453]:

	rank	x_classification	region	start_time	maximum_time	\
0	1	X28+	486	2003-11-04 19:29:00	2003-11-04 19:53:00	
1	2	X20+	9393	2001-04-02 21:32:00	2001-04-02 21:51:00	
2	3	X17.2+	486	2003-10-28 09:51:00	2003-10-28 11:10:00	
3	4	X17+	808	2005-09-07 17:17:00	2005-09-07 17:40:00	
4	5	X14.4	9415	2001-04-15 13:19:00	2001-04-15 13:50:00	
5	6	X10	486	2003-10-29 20:37:00	2003-10-29 20:49:00	
6	7	X9.4	8100	1997-11-06 11:49:00	1997-11-06 11:55:00	
7	8	X9.3	2673	2017-09-06 11:53:00	2017-09-06 12:02:00	
8	9	X9	930	2006-12-05 10:18:00	2006-12-05 10:35:00	
9	10	X8.3	486	2003-11-02 17:03:00	2003-11-02 17:25:00	
10	11	X8.2	2673	2017-09-10 15:35:00	2017-09-10 16:06:00	
11	12	X7.1	720	2005-01-20 06:36:00	2005-01-20 07:01:00	
12	13	X6.9	1263	2011-08-09 07:48:00	2011-08-09 08:05:00	
13	14	X6.5	930	2006-12-06 18:29:00	2006-12-06 18:47:00	
14	15	X6.2	808	2005-09-09 19:13:00	2005-09-09 20:04:00	
15	16	X6.2	9733	2001-12-13 14:20:00	2001-12-13 14:30:00	
16	17	X5.7	9077	2000-07-14 10:03:00	2000-07-14 10:24:00	
17	18	X5.6	9415	2001-04-06 19:10:00	2001-04-06 19:21:00	
18	19	X5.4	1429	2012-03-07 00:02:00	2012-03-07 00:24:00	
19	20	X5.4	808	2005-09-08 20:52:00	2005-09-08 21:06:00	
20	21	X5.4	486	2003-10-23 08:19:00	2003-10-23 08:35:00	

21	22	X5.3	9591	2001-08-25 16:23:00	2001-08-25 16:45:00
22	23	X4.9	1990	2014-02-25 00:39:00	2014-02-25 00:49:00
23	24	X4.9	8307	1998-08-18 22:10:00	1998-08-18 22:19:00
24	25	X4.8	39	2002-07-23 00:18:00	2002-07-23 00:35:00
25	26	X4	9236	2000-11-26 16:34:00	2000-11-26 16:48:00
26	27	X3.9	488	2003-11-03 09:43:00	2003-11-03 09:55:00
27	28	X3.9	8307	1998-08-19 21:35:00	1998-08-19 21:45:00
28	29	X3.8	720	2005-01-17 06:59:00	2005-01-17 09:52:00
29	30	X3.7	8384	1998-11-22 06:30:00	1998-11-22 06:42:00
30	31	X3.6	808	2005-09-09 09:42:00	2005-09-09 09:59:00
31	32	X3.6	649	2004-07-16 13:49:00	2004-07-16 13:55:00
32	33	X3.6	365	2003-05-28 00:17:00	2003-05-28 00:27:00
33	34	X3.4	930	2006-12-13 02:14:00	2006-12-13 02:40:00
34	35	X3.4	9767	2001-12-28 20:02:00	2001-12-28 20:45:00
35	36	X3.3	1890	2013-11-05 22:07:00	2013-11-05 22:12:00
36	37	X3.3	39	2002-07-20 21:04:00	2002-07-20 21:30:00
37	38	X3.3	8395	1998-11-28 04:54:00	1998-11-28 05:52:00
38	39	X3.2	1748	2013-05-14 00:00:00	2013-05-14 01:11:00
39	40	X3.1	2192	2014-10-24 21:07:00	2014-10-24 21:41:00
40	41	X3.1	69	2002-08-24 00:49:00	2002-08-24 01:12:00
41	42	X3	30	2002-07-15 19:59:00	2002-07-15 20:08:00
42	43	X2.8	1748	2013-05-13 15:48:00	2013-05-13 16:05:00
43	44	X2.8	9733	2001-12-11 07:58:00	2001-12-11 08:08:00
44	45	X2.8	8307	1998-08-18 08:14:00	1998-08-18 08:24:00
45	46	X2.7	2339	2015-05-05 22:05:00	2015-05-05 22:11:00
46	47	X2.7	488	2003-11-03 01:09:00	2003-11-03 01:30:00
47	48	X2.7	8210	1998-05-06 07:58:00	1998-05-06 08:09:00
48	49	X2.6	720	2005-01-15 22:25:00	2005-01-15 23:02:00
49	50	X2.6	9632	2001-09-24 09:32:00	2001-09-24 10:38:00

	end_time
0	2003-11-04 20:06:00
1	2001-04-02 22:03:00
2	2003-10-28 11:24:00
3	2005-09-07 18:03:00
4	2001-04-15 13:55:00
5	2003-10-29 21:01:00
6	1997-11-06 12:01:00
7	2017-09-06 12:10:00
8	2006-12-05 10:45:00
9	2003-11-02 17:39:00
10	2017-09-10 16:31:00
11	2005-01-20 07:26:00
12	2011-08-09 08:08:00
13	2006-12-06 19:00:00
14	2005-09-09 20:36:00
15	2001-12-13 14:35:00

```
16  2000-07-14 10:43:00
17  2001-04-06 19:31:00
18  2012-03-07 00:40:00
19  2005-09-08 21:17:00
20  2003-10-23 08:49:00
21  2001-08-25 17:04:00
22  2014-02-25 01:03:00
23  1998-08-18 22:28:00
24  2002-07-23 00:47:00
25  2000-11-26 16:56:00
26  2003-11-03 10:19:00
27  1998-08-19 21:50:00
28  2005-01-17 10:07:00
29  1998-11-22 06:49:00
30  2005-09-09 10:08:00
31  2004-07-16 14:01:00
32  2003-05-28 00:39:00
33  2006-12-13 02:57:00
34  2001-12-28 21:32:00
35  2013-11-05 22:15:00
36  2002-07-20 21:54:00
37  1998-11-28 06:13:00
38  2013-05-14 01:20:00
39  2014-10-24 22:13:00
40  2002-08-24 01:31:00
41  2002-07-15 20:14:00
42  2013-05-13 16:16:00
43  2001-12-11 08:14:00
44  1998-08-18 08:32:00
45  2015-05-05 22:15:00
46  2003-11-03 01:45:00
47  1998-05-06 08:20:00
48  2005-01-15 23:31:00
49  2001-09-24 11:09:00
```

Part 1 Step 2: I begin cleaning the data by dropping the movies column from the DataFrame. I then create datetime objects using the 'date' column and the 'start_time', 'maximum_time', and 'end_time' columns. This datetime object replaces whatever is at the 'index' of the 'date', 'start_time', 'maximum_time', and 'end_time' columns. Additionally, if the cell at 'index' in column 'Region' contains a '-' character I replace the cell with NaN.

[454]:
```
#Step 3: Scrape the NASA data

r = requests.get('http://www.hcbravo.org/IntroDataSci/misc/waves_type2.html')
root = BeautifulSoup(r.content)
rawtext = root.find("pre").get_text()
rawtext = rawtext.split('\n')
```

```
table = []

for item in rawtext[12:-3]:
    item = item.split()
    table.append(item)
df_nasa =  pd.DataFrame(table)
df_nasa = df_nasa.iloc[: , : -9]
df_nasa = df_nasa.set_axis(['start_date', 'start_time', 'end_date', 'end_time',␣
 ↪'start_freq', 'end_freq', 'Loc', 'NOAA', 'Imp', 'cme_date', 'cme_time',␣
 ↪'CPA', 'cme_width', 'cme_speed', 'plots'], axis=1, inplace=False)
#df_nasa
```

Part 1 Step 3: I begin by using 'requests' and BeautifulSoup to get the HTML content from the hcbravo site. Next, I find the text of the page and 'split' it into a list of strings where each entry is a row of the table. From there, I iterate over the rows that make up the table and 'split' again to get each value by itself. Each value gets appended to a separate list and the whole list is converted into a DataFrame. I drop extra columns picked up from extraneous text from the site and set the column names.

[455]:
```
#Step 4: Tidy the NASA table
df_nasa['start_freq'] = df_nasa.start_freq.apply(lambda x: x if x != "????"␣
 ↪else np.nan)
df_nasa['end_freq'] = df_nasa.end_freq.apply(lambda x: x if x != "????" else np.
 ↪nan)
df_nasa['NOAA'] = df_nasa.NOAA.apply(lambda x: x if x != "-----" else np.nan)
df_nasa['Imp'] = df_nasa.Imp.apply(lambda x: x if x != "----" else np.nan)
df_nasa['is_halo'] = df_nasa.CPA.apply(lambda x: True if x == "Halo" else False)
df_nasa['CPA'] = df_nasa.CPA.apply(lambda x: x if x != "----" else np.nan)
df_nasa['CPA'] = df_nasa.CPA.apply(lambda x: x if x != "Halo" else np.nan)
df_nasa['width_lower_bound'] = df_nasa.cme_width.apply(lambda x: True if '>' in␣
 ↪str(x) else False)
df_nasa['cme_width'] = df_nasa.cme_width.apply(lambda x: str(x)[1:] if '>' in␣
 ↪str(x) else str(x))
df_nasa['cme_width'] = df_nasa.cme_width.apply(lambda x: x if x != "----" else␣
 ↪np.nan)
df_nasa['cme_speed'] = df_nasa.cme_speed.apply(lambda x: x if x != "----" else␣
 ↪np.nan)
df_nasa['Loc'] = df_nasa.Loc.apply(lambda x: np.nan if "back" in str(x).lower()␣
 ↪else str(x))
df_nasa['cme_date'] = df_nasa.cme_date.apply(lambda x: "01/01" if str(x) == "--/
 ↪--" else str(x))
df_nasa['cme_time'] = df_nasa.cme_time.apply(lambda x: "00:00" if str(x) == "--:
 ↪--" else str(x))

startdate_list = []
for index, row in df_nasa.iterrows():
    if "24:00" in str(row['start_time']):
```

```
        row['start_time'] = "00:00"
    if "24:00" in str(row['cme_time']):
        row['cme_time'] = "00:00"
    if "24:00" in str(row['end_time']):
        row['end_time'] = "00:00"
    startdate_str = str(row['start_date']) + " " + str(row['start_time'])
    startdate_list.append(startdate_str)
    cmedate_str = startdate_str[0:5] + str(row['cme_date']) + " " +␣
↪str(row['cme_time'])
    enddate_str = startdate_str[0:5] + str(row['end_date']) + " " +␣
↪str(row['end_time'])
    startdate_str = startdate_str.replace('/', '-')
    cmedate_str = cmedate_str.replace('/', '-')
    enddate_str = enddate_str.replace('/', '-')

    startdate_obj = datetime.datetime.strptime(startdate_str, '%Y-%m-%d %H:%M')
    cmedate_obj = datetime.datetime.strptime(cmedate_str, '%Y-%m-%d %H:%M')
    enddate_obj = datetime.datetime.strptime(enddate_str, '%Y-%m-%d %H:%M')
    df_nasa.at[index, 'start_datetime'] = startdate_obj
    df_nasa.at[index, 'end_datetime'] = cmedate_obj
    df_nasa.at[index, 'cme_datetime'] = enddate_obj

df_nasa = df_nasa.drop(columns = ['start_time', 'start_date', 'end_date',␣
↪'end_time', 'cme_date', 'cme_time'])
df_nasa
```

[455]:

	start_freq	end_freq	Loc	NOAA	Imp	CPA	cme_width	cme_speed	plots	\
0	8000	4000	S25E16	8026	M1.3	74	79	312	PHTX	
1	11000	1000	S28E19	8027	C6.8	NaN	360	878	PHTX	
2	12000	80	N21W08	8038	C1.3	NaN	360	464	PHTX	
3	5000	500	N05W12	8040	M1.3	263	165	296	PHTX	
4	6000	2000	S29E25	8088	C1.4	133	155	712	PHTX	
..	…	…	…	…	…	…	…	…	…	
477	14000	3900	W90b	NaN	NaN	NaN	360	2222	PHTX	
478	2900	2100	S11E33	12241	M1.1	107	108	869	PHTX	
479	14000	11500	S20E09	12242	M8.7	NaN	360	587	PHTX	
480	5100	1300	S11E15	12241	M6.9	NaN	360	1195	PHTX	
481	14000	7400	S14W25	12241	M1.0	NaN	360	669	PHTX	

	is_halo	width_lower_bound	start_datetime	end_datetime	\
0	False	False	1997-04-01 14:00:00	1997-04-01 15:18:00	
1	True	False	1997-04-07 14:30:00	1997-04-07 14:27:00	
2	True	False	1997-05-12 05:15:00	1997-05-12 05:30:00	
3	False	False	1997-05-21 20:20:00	1997-05-21 21:00:00	
4	False	False	1997-09-23 21:53:00	1997-09-23 22:02:00	
..	…	…	…	…	
477	True	False	2014-12-13 14:27:00	2014-12-13 14:24:00	

```
478    False              False 2014-12-17 04:09:00 2014-12-17 02:00:00
479     True              False 2014-12-17 05:00:00 2014-12-17 05:00:00
480     True              False 2014-12-18 22:31:00 2014-12-19 01:04:00
481     True              False 2014-12-21 12:05:00 2014-12-21 12:12:00

           cme_datetime
0   1997-04-01 14:15:00
1   1997-04-07 17:30:00
2   1997-05-14 16:00:00
3   1997-05-21 22:00:00
4   1997-09-23 22:16:00
..                  …
477 2014-12-13 14:51:00
478 2014-12-17 04:19:00
479 2014-12-17 05:09:00
480 2014-12-18 22:54:00
481 2014-12-21 12:28:00

[482 rows x 14 columns]
```

Part 1 Step 4: I complete steps 1, 2, and 3 listed on the github page using a series of lambda functions involving the columns of the DataFrame. For step 4, I iterate through the rows of the DataFrame and convert datetime objects for the start, maximum, and CME date and time columns. Then, I drop said columns as I stored the datetime objects into new columns.

[456]:
```
#Part 2: Analysis
#-----------------------------------------------------
#Question 1: Replication
def classify(x):
    if x == x and x != "FILA" and str(x)[0] == 'X':
        return str(x)[1:]
    else:
        return "NaN"

df_tmp1 = df_nasa.copy(deep=True)
df_tmp1['Imp_Classify'] = df_tmp1.Imp.apply(classify)
df_tmp1['Imp_Classify'] = df_tmp1['Imp_Classify'].astype(float)
df_tmp1.sort_values('Imp_Classify',inplace=True, ascending=False)
imps = df_tmp1.head(50)
print(type(imps))
#df_tmp1['Imp_Classify'].head(50)
```

```
<class 'pandas.core.frame.DataFrame'>
```

Question 1: Replication

The data coming from NASA is incomplete in some instances (e.g. 10/28/2003 NASA has X17. whereas SWL has X17.2) or incorrect (e.g. 09/07/2005 NASA has X1.7 whereas SWL has X17.0). Therefore you cannot get the table from SWL exactly without manipulating the data extracted

from NASA. Given the relatively small size of the SWL data (n = 50), this could be a meaningful investment.

[ ]:

[473]:
```
#Question 2: Integration
swl_datetimes = []
for index, row in df_swl.iterrows():
    swl_datetimes.append((df_swl.at[index,'start_time']))
nasa_datetimes = []
for index, row in df_nasa.iterrows():
    nasa_datetimes.append((df_nasa.at[index, 'start_datetime']))

df_match = pd.DataFrame(columns=['swl_date', "nasa_date", "diff", "swl_rank"])
count = 0
for i in range(len(swl_datetimes)):
    diff = timedelta(weeks=39)
    swl_dt = swl_datetimes[i]
    df_match.at[count, "swl_date"] = swl_dt
    for j in range(len(nasa_datetimes)):
        nasa_dt = nasa_datetimes[j]
        dt_diff = abs(swl_dt - nasa_dt)
        if dt_diff < diff:
            diff = dt_diff
            df_match.at[count, "nasa_date"] = nasa_dt
            df_match.at[count, "diff"] = diff
    df_match.at[count, "swl_rank"] = count + 1
    count += 1

df_match = df_match.sort_values(by=['nasa_date'])
df_match = df_match.reset_index(drop=True)
temp = [np.nan for i in range(482)]
df_nasa = df_nasa.assign(approx_SWLRank = temp)
df_nasa = df_nasa.sort_values(by=['start_datetime'])

match_list = []
nasa_list = df_match['nasa_date'].tolist()
swl_list = df_match['swl_rank'].tolist()
for i in range(50):
    match_list.append(tuple((nasa_list[i], swl_list[i])))

for index in range(len(df_nasa['start_datetime'])):
    t1 = pd.Timestamp(df_nasa.at[index, "start_datetime"])
    d1 = t1.to_pydatetime()
    s1 = d1.strftime("%m/%d/%Y")
    for j in range(50):
        temp = match_list[j]
        t2 = pd.Timestamp(temp[0])
```

```
        d2 = t2.to_pydatetime()
        s2 = d1.strftime("%m/%d/%Y")
        if s2 == s1:
            #print(temp[1])
            df_nasa.at[index, "approx_SWLRank"] = temp[1]

df_nasa.head(25)
```

[473]:

	start_freq	end_freq	Loc	NOAA	Imp	CPA	cme_width	cme_speed	plots
0	8000	4000	S25E16	8026	M1.3	74	79	312	PHTX
1	11000	1000	S28E19	8027	C6.8	NaN	360	878	PHTX
2	12000	80	N21W08	8038	C1.3	NaN	360	464	PHTX
3	5000	500	N05W12	8040	M1.3	263	165	296	PHTX
4	6000	2000	S29E25	8088	C1.4	133	155	712	PHTX
5	14000	250	S20W13	8100	C8.6	240	109	227	PHTX
6	14000	5000	S16W21	8100	M4.2	233	122	352	PHTX
7	14000	100	S14W33	8100	X2.1	NaN	360	785	PHTX
8	14000	100	S18W63	8100	X9.4	NaN	360	1556	PHTX
9	14000	7000	N17E63	8113	X2.6	98	91	441	PHTX
10	14000	8000	N25W52	8116	B9.4	278	73	191	PHTX
11	14000	10000	N21E25	8141	C1.1	NaN	360	693	PHTX
12	14000	7000	SW90	NaN	NaN	NaN	360	1397	PHTX
13	10000	35	S22W90	8194	M1.4	284	165	1863	PHTX
14	14000	200	S17E90	8210	X1.2	NaN	360	1691	PHTX
15	4700	2600	S10E90	8210	C8.9	100	84	1184	PHTX
16	10000	1000	S16E50	8210	X1.0	NaN	360	1385	PHTX
17	10000	2000	S18E20	8210	M6.8	NaN	360	1374	PHTX
18	5000	3000	S15W15	8210	X1.1	NaN	360	938	PHTX
19	14000	5000	S11W65	8210	X2.7	309	190	1099	PHTX
20	9000	400	S14W89	8210	M7.7	262	178	2331	PHTX
21	10000	1000	N32W90	8214	B6.6	208	301	830	PHTX
22	14000	3000	N29W46	8222	B7.9	268	139	801	PHTX
23	4000	1000	N19W62	8226	C7.5	175	268	878	PHTX
24	8000	4000	N16E86	8243	M1.4	123	177	1223	PHTX

	is_halo	width_lower_bound	start_datetime	end_datetime
0	False	False	1997-04-01 14:00:00	1997-04-01 15:18:00
1	True	False	1997-04-07 14:30:00	1997-04-07 14:27:00
2	True	False	1997-05-12 05:15:00	1997-05-12 05:30:00
3	False	False	1997-05-21 20:20:00	1997-05-21 21:00:00
4	False	False	1997-09-23 21:53:00	1997-09-23 22:02:00
5	False	False	1997-11-03 05:15:00	1997-11-03 05:28:00
6	False	False	1997-11-03 10:30:00	1997-11-03 11:11:00
7	True	False	1997-11-04 06:00:00	1997-11-04 06:10:00
8	True	False	1997-11-06 12:20:00	1997-11-06 12:10:00
9	False	False	1997-11-27 13:30:00	1997-11-27 13:56:00
10	False	False	1997-12-12 22:45:00	1997-12-13 00:26:00

11	True	False	1998-01-25 15:03:00	1998-01-25 15:26:00
12	True	False	1998-03-29 03:40:00	1998-03-29 03:48:00
13	False	False	1998-04-20 10:25:00	1998-04-20 10:07:00
14	True	False	1998-04-23 06:00:00	1998-04-23 05:55:00
15	False	False	1998-04-24 09:17:00	1998-04-24 08:55:00
16	True	False	1998-04-27 09:20:00	1998-04-27 08:56:00
17	True	False	1998-04-29 16:30:00	1998-04-29 16:58:00
18	True	False	1998-05-02 14:25:00	1998-05-02 14:06:00
19	False	False	1998-05-06 08:25:00	1998-05-06 08:29:00
20	False	False	1998-05-09 03:35:00	1998-05-09 03:35:00
21	False	True	1998-05-11 21:40:00	1998-05-11 21:55:00
22	False	False	1998-05-19 10:00:00	1998-05-19 10:27:00
23	False	False	1998-05-27 13:30:00	1998-05-27 13:45:00
24	False	False	1998-06-11 10:15:00	1998-06-11 10:28:00

	cme_datetime	approx_SWLRank
0	1997-04-01 14:15:00	11.0
1	1997-04-07 17:30:00	11.0
2	1997-05-14 16:00:00	11.0
3	1997-05-21 22:00:00	11.0
4	1997-09-23 22:16:00	11.0
5	1997-11-03 12:00:00	11.0
6	1997-11-03 11:30:00	11.0
7	1997-11-05 04:30:00	11.0
8	1997-11-07 08:30:00	11.0
9	1997-11-27 14:00:00	11.0
10	1997-12-12 23:20:00	11.0
11	1998-01-25 15:18:00	11.0
12	1998-03-29 03:52:00	11.0
13	1998-04-22 06:00:00	11.0
14	1998-04-23 15:30:00	11.0
15	1998-04-24 09:25:00	11.0
16	1998-04-27 10:00:00	11.0
17	1998-04-29 17:00:00	11.0
18	1998-05-02 14:50:00	11.0
19	1998-05-06 08:35:00	11.0
20	1998-05-09 10:00:00	11.0
21	1998-05-11 22:00:00	11.0
22	1998-05-19 11:30:00	11.0
23	1998-05-27 14:20:00	11.0
24	1998-06-11 10:20:00	11.0

Question 2: Integration

I chose to use the starting date of the type II burst (first column NASA data, second column SWL data) as a the way to replicate the SWL data using the NASA data. However, I only used MM-DD-YYYY format because anything more specific (e.g. hour, minute, etc.) wouldn't match between the two datasets exactly. Furthermore, since neither the X classification (column nine NASA data) nor

the region (column eight NASA data) were consistent between the datasets, I couldn't use either of those indicators. Thus, the only consistent way to compare datapoints between the two sites is the date. However, sometimes there were multiple entries in the NASA dataset on the same day. For example, on 04/02/2001 there were two recordings. The first recording had an x classification of X1.1 whereas the second had an x classification of X20. Thus, whichever recording had the higher x classification is the one I'd use.

[ ]:
```
#Question 3: Analysis
datetime_list = []
for item in startdate_list:
    item = item.replace('/', '-')
    item = datetime.datetime.strptime(item, '%Y-%m-%d %H:%M')
    datetime_list.append(item)
X_plot = matplotlib.dates.date2num(datetime_list)
Y_plot = df_nasa['NOAA'].values
Y_plot = Y_plot.astype(str)
plt.plot_date(X_plot, Y_plot)
plt.show()
```

Question 3:

I plotted location over time. This shows how quickly new solar flares got added to the database.

[ ]:

[ ]: