

```

#Part 1
#Step 1
import json
import requests
import pandas as pd
from bs4 import BeautifulSoup
r = requests.get("https://cmssc320.github.io/files/top-50-solar-
flares.html")

rSoup = BeautifulSoup(r.content, 'html.parser')
#rSoup.prettify()

rTable = rSoup.find('table')

from IPython.display import display
# df = pd.DataFrame(columns=['rank', 'x_class', 'date', 'region',
'start', 'max_time', 'end', 'movie'])
tables = pd.read_html(r.content)
df = tables[0]
display(df)

```

	Unnamed: 0	Unnamed: 1	Unnamed: 2	Region	Start	Maximum	End	\
0	1	X28+	2003/11/04	486	19:29	19:53	20:06	
1	2	X20+	2001/04/02	9393	21:32	21:51	22:03	
2	3	X17.2+	2003/10/28	486	09:51	11:10	11:24	
3	4	X17+	2005/09/07	808	17:17	17:40	18:03	
4	5	X14.4	2001/04/15	9415	13:19	13:50	13:55	
5	6	X10	2003/10/29	486	20:37	20:49	21:01	
6	7	X9.4	1997/11/06	8100	11:49	11:55	12:01	
7	8	X9.3	2017/09/06	2673	11:53	12:02	12:10	
8	9	X9	2006/12/05	930	10:18	10:35	10:45	
9	10	X8.3	2003/11/02	486	17:03	17:25	17:39	
10	11	X8.2	2017/09/10	2673	15:35	16:06	16:31	
11	12	X7.1	2005/01/20	720	06:36	07:01	07:26	
12	13	X6.9	2011/08/09	1263	07:48	08:05	08:08	
13	14	X6.5	2006/12/06	930	18:29	18:47	19:00	
14	15	X6.2	2005/09/09	808	19:13	20:04	20:36	
15	16	X6.2	2001/12/13	9733	14:20	14:30	14:35	
16	17	X5.7	2000/07/14	9077	10:03	10:24	10:43	
17	18	X5.6	2001/04/06	9415	19:10	19:21	19:31	
18	19	X5.4	2012/03/07	1429	00:02	00:24	00:40	
19	20	X5.4	2005/09/08	808	20:52	21:06	21:17	
20	21	X5.4	2003/10/23	486	08:19	08:35	08:49	
21	22	X5.3	2001/08/25	9591	16:23	16:45	17:04	
22	23	X4.9	2014/02/25	1990	00:39	00:49	01:03	
23	24	X4.9	1998/08/18	8307	22:10	22:19	22:28	
24	25	X4.8	2002/07/23	39	00:18	00:35	00:47	
25	26	X4	2000/11/26	9236	16:34	16:48	16:56	
26	27	X3.9	2003/11/03	488	09:43	09:55	10:19	
27	28	X3.9	1998/08/19	8307	21:35	21:45	21:50	

28	29	X3.8	2005/01/17	720	06:59	09:52	10:07
29	30	X3.7	1998/11/22	8384	06:30	06:42	06:49
30	31	X3.6	2005/09/09	808	09:42	09:59	10:08
31	32	X3.6	2004/07/16	649	13:49	13:55	14:01
32	33	X3.6	2003/05/28	365	00:17	00:27	00:39
33	34	X3.4	2006/12/13	930	02:14	02:40	02:57
34	35	X3.4	2001/12/28	9767	20:02	20:45	21:32
35	36	X3.3	2013/11/05	1890	22:07	22:12	22:15
36	37	X3.3	2002/07/20	39	21:04	21:30	21:54
37	38	X3.3	1998/11/28	8395	04:54	05:52	06:13
38	39	X3.2	2013/05/14	1748	00:00	01:11	01:20
39	40	X3.1	2014/10/24	2192	21:07	21:41	22:13
40	41	X3.1	2002/08/24	69	00:49	01:12	01:31
41	42	X3	2002/07/15	30	19:59	20:08	20:14
42	43	X2.8	2013/05/13	1748	15:48	16:05	16:16
43	44	X2.8	2001/12/11	9733	07:58	08:08	08:14
44	45	X2.8	1998/08/18	8307	08:14	08:24	08:32
45	46	X2.7	2015/05/05	2339	22:05	22:11	22:15
46	47	X2.7	2003/11/03	488	01:09	01:30	01:45
47	48	X2.7	1998/05/06	8210	07:58	08:09	08:20
48	49	X2.6	2005/01/15	720	22:25	23:02	23:31
49	50	X2.6	2001/09/24	9632	09:32	10:38	11:09

Unnamed: 7

0	MovieView archive
1	MovieView archive
2	MovieView archive
3	MovieView archive
4	MovieView archive
5	MovieView archive
6	MovieView archive
7	MovieView archive
8	MovieView archive
9	MovieView archive
10	MovieView archive
11	MovieView archive
12	MovieView archive
13	MovieView archive
14	MovieView archive
15	MovieView archive
16	MovieView archive
17	MovieView archive
18	MovieView archive
19	MovieView archive
20	MovieView archive
21	MovieView archive
22	MovieView archive
23	View archive
24	MovieView archive
25	MovieView archive

```

26 MovieView archive
27     View archive
28 MovieView archive
29 MovieView archive
30 MovieView archive
31 MovieView archive
32 MovieView archive
33 MovieView archive
34 MovieView archive
35 MovieView archive
36 MovieView archive
37 MovieView archive
38 MovieView archive
39 MovieView archive
40 MovieView archive
41 MovieView archive
42 MovieView archive
43 MovieView archive
44     View archive
45 MovieView archive
46 MovieView archive
47 MovieView archive
48 MovieView archive
49 MovieView archive

```

#Step 2

```

#df = df.drop('Unnamed: 7', 1)
tidydf = df.rename({'Unnamed: 0': 'Rank', 'Unnamed: 1': 'X Class',
'Unnamed: 2': 'Date'}, axis=1)
display(tidydf)

```

	Rank	X Class	Date	Region	Start	Maximum	End
0	1	X28+	2003/11/04	486	19:29	19:53	20:06
1	2	X20+	2001/04/02	9393	21:32	21:51	22:03
2	3	X17.2+	2003/10/28	486	09:51	11:10	11:24
3	4	X17+	2005/09/07	808	17:17	17:40	18:03
4	5	X14.4	2001/04/15	9415	13:19	13:50	13:55
5	6	X10	2003/10/29	486	20:37	20:49	21:01
6	7	X9.4	1997/11/06	8100	11:49	11:55	12:01
7	8	X9.3	2017/09/06	2673	11:53	12:02	12:10
8	9	X9	2006/12/05	930	10:18	10:35	10:45
9	10	X8.3	2003/11/02	486	17:03	17:25	17:39
10	11	X8.2	2017/09/10	2673	15:35	16:06	16:31
11	12	X7.1	2005/01/20	720	06:36	07:01	07:26
12	13	X6.9	2011/08/09	1263	07:48	08:05	08:08
13	14	X6.5	2006/12/06	930	18:29	18:47	19:00
14	15	X6.2	2005/09/09	808	19:13	20:04	20:36
15	16	X6.2	2001/12/13	9733	14:20	14:30	14:35
16	17	X5.7	2000/07/14	9077	10:03	10:24	10:43
17	18	X5.6	2001/04/06	9415	19:10	19:21	19:31
18	19	X5.4	2012/03/07	1429	00:02	00:24	00:40

19	20	X5.4	2005/09/08	808	20:52	21:06	21:17
20	21	X5.4	2003/10/23	486	08:19	08:35	08:49
21	22	X5.3	2001/08/25	9591	16:23	16:45	17:04
22	23	X4.9	2014/02/25	1990	00:39	00:49	01:03
23	24	X4.9	1998/08/18	8307	22:10	22:19	22:28
24	25	X4.8	2002/07/23	39	00:18	00:35	00:47
25	26	X4	2000/11/26	9236	16:34	16:48	16:56
26	27	X3.9	2003/11/03	488	09:43	09:55	10:19
27	28	X3.9	1998/08/19	8307	21:35	21:45	21:50
28	29	X3.8	2005/01/17	720	06:59	09:52	10:07
29	30	X3.7	1998/11/22	8384	06:30	06:42	06:49
30	31	X3.6	2005/09/09	808	09:42	09:59	10:08
31	32	X3.6	2004/07/16	649	13:49	13:55	14:01
32	33	X3.6	2003/05/28	365	00:17	00:27	00:39
33	34	X3.4	2006/12/13	930	02:14	02:40	02:57
34	35	X3.4	2001/12/28	9767	20:02	20:45	21:32
35	36	X3.3	2013/11/05	1890	22:07	22:12	22:15
36	37	X3.3	2002/07/20	39	21:04	21:30	21:54
37	38	X3.3	1998/11/28	8395	04:54	05:52	06:13
38	39	X3.2	2013/05/14	1748	00:00	01:11	01:20
39	40	X3.1	2014/10/24	2192	21:07	21:41	22:13
40	41	X3.1	2002/08/24	69	00:49	01:12	01:31
41	42	X3	2002/07/15	30	19:59	20:08	20:14
42	43	X2.8	2013/05/13	1748	15:48	16:05	16:16
43	44	X2.8	2001/12/11	9733	07:58	08:08	08:14
44	45	X2.8	1998/08/18	8307	08:14	08:24	08:32
45	46	X2.7	2015/05/05	2339	22:05	22:11	22:15
46	47	X2.7	2003/11/03	488	01:09	01:30	01:45
47	48	X2.7	1998/05/06	8210	07:58	08:09	08:20
48	49	X2.6	2005/01/15	720	22:25	23:02	23:31
49	50	X2.6	2001/09/24	9632	09:32	10:38	11:09

```
st_value = pd.to_datetime(tidydf['Date'] + ' ' + tidydf['Start'])
mt_value = pd.to_datetime(tidydf['Date'] + ' ' + tidydf['Maximum'])
et_value = pd.to_datetime(tidydf['Date'] + ' ' + tidydf['End'])
```

```
tidydf['Start DateTime'] = st_value
tidydf['Max DateTime'] = mt_value
tidydf['End DateTime'] = et_value
```

```
tidydf = tidydf.drop('Date', 1)
tidydf = tidydf.drop('Start', 1)
tidydf = tidydf.drop('Maximum', 1)
tidydf = tidydf.drop('End', 1)
```

```
/var/folders/yl/0_18jst15nb9gbl2n_j_z5tw0000gn/T/
ipykernel_3763/3178710839.py:9: FutureWarning: In a future version of
pandas all arguments of DataFrame.drop except for the argument
'labels' will be keyword-only.
    tidydf = tidydf.drop('Date', 1)
```

```
/var/folders/yl/0_18jst15nb9gbl2n_j_z5tw0000gn/T/ipykernel_3763/317871
0839.py:10: FutureWarning: In a future version of pandas all arguments
of DataFrame.drop except for the argument 'labels' will be keyword-
only.
```

```
tidydf = tidydf.drop('Start', 1)
```

```
/var/folders/yl/0_18jst15nb9gbl2n_j_z5tw0000gn/T/ipykernel_3763/317871
0839.py:11: FutureWarning: In a future version of pandas all arguments
of DataFrame.drop except for the argument 'labels' will be keyword-
only.
```

```
tidydf = tidydf.drop('Maximum', 1)
```

```
/var/folders/yl/0_18jst15nb9gbl2n_j_z5tw0000gn/T/ipykernel_3763/317871
0839.py:12: FutureWarning: In a future version of pandas all arguments
of DataFrame.drop except for the argument 'labels' will be keyword-
only.
```

```
tidydf = tidydf.drop('End', 1)
```

```
tidydf = tidydf[['Rank', 'X Class', 'Start DateTime', 'Max DateTime',
'End DateTime', 'Region']]
display(tidydf)
```

	Rank	X Class	Start DateTime	Max DateTime	End
0	1	X28+	2003-11-04 19:29:00	2003-11-04 19:53:00	2003-11-04 20:06:00
1	2	X20+	2001-04-02 21:32:00	2001-04-02 21:51:00	2001-04-02 22:03:00
2	3	X17.2+	2003-10-28 09:51:00	2003-10-28 11:10:00	2003-10-28 11:24:00
3	4	X17+	2005-09-07 17:17:00	2005-09-07 17:40:00	2005-09-07 18:03:00
4	5	X14.4	2001-04-15 13:19:00	2001-04-15 13:50:00	2001-04-15 13:55:00
5	6	X10	2003-10-29 20:37:00	2003-10-29 20:49:00	2003-10-29 21:01:00
6	7	X9.4	1997-11-06 11:49:00	1997-11-06 11:55:00	1997-11-06 12:01:00
7	8	X9.3	2017-09-06 11:53:00	2017-09-06 12:02:00	2017-09-06 12:10:00
8	9	X9	2006-12-05 10:18:00	2006-12-05 10:35:00	2006-12-05 10:45:00
9	10	X8.3	2003-11-02 17:03:00	2003-11-02 17:25:00	2003-11-02 17:39:00
10	11	X8.2	2017-09-10 15:35:00	2017-09-10 16:06:00	2017-09-10 16:31:00
11	12	X7.1	2005-01-20 06:36:00	2005-01-20 07:01:00	2005-01-20 07:26:00
12	13	X6.9	2011-08-09 07:48:00	2011-08-09 08:05:00	2011-08-09 08:08:00
13	14	X6.5	2006-12-06 18:29:00	2006-12-06 18:47:00	2006-12-06 19:00:00
14	15	X6.2	2005-09-09 19:13:00	2005-09-09 20:04:00	2005-09-09

20:36:00						
15	16	X6.2	2001-12-13	14:20:00	2001-12-13	14:30:00 2001-12-13
14:35:00						
16	17	X5.7	2000-07-14	10:03:00	2000-07-14	10:24:00 2000-07-14
10:43:00						
17	18	X5.6	2001-04-06	19:10:00	2001-04-06	19:21:00 2001-04-06
19:31:00						
18	19	X5.4	2012-03-07	00:02:00	2012-03-07	00:24:00 2012-03-07
00:40:00						
19	20	X5.4	2005-09-08	20:52:00	2005-09-08	21:06:00 2005-09-08
21:17:00						
20	21	X5.4	2003-10-23	08:19:00	2003-10-23	08:35:00 2003-10-23
08:49:00						
21	22	X5.3	2001-08-25	16:23:00	2001-08-25	16:45:00 2001-08-25
17:04:00						
22	23	X4.9	2014-02-25	00:39:00	2014-02-25	00:49:00 2014-02-25
01:03:00						
23	24	X4.9	1998-08-18	22:10:00	1998-08-18	22:19:00 1998-08-18
22:28:00						
24	25	X4.8	2002-07-23	00:18:00	2002-07-23	00:35:00 2002-07-23
00:47:00						
25	26	X4	2000-11-26	16:34:00	2000-11-26	16:48:00 2000-11-26
16:56:00						
26	27	X3.9	2003-11-03	09:43:00	2003-11-03	09:55:00 2003-11-03
10:19:00						
27	28	X3.9	1998-08-19	21:35:00	1998-08-19	21:45:00 1998-08-19
21:50:00						
28	29	X3.8	2005-01-17	06:59:00	2005-01-17	09:52:00 2005-01-17
10:07:00						
29	30	X3.7	1998-11-22	06:30:00	1998-11-22	06:42:00 1998-11-22
06:49:00						
30	31	X3.6	2005-09-09	09:42:00	2005-09-09	09:59:00 2005-09-09
10:08:00						
31	32	X3.6	2004-07-16	13:49:00	2004-07-16	13:55:00 2004-07-16
14:01:00						
32	33	X3.6	2003-05-28	00:17:00	2003-05-28	00:27:00 2003-05-28
00:39:00						
33	34	X3.4	2006-12-13	02:14:00	2006-12-13	02:40:00 2006-12-13
02:57:00						
34	35	X3.4	2001-12-28	20:02:00	2001-12-28	20:45:00 2001-12-28
21:32:00						
35	36	X3.3	2013-11-05	22:07:00	2013-11-05	22:12:00 2013-11-05
22:15:00						
36	37	X3.3	2002-07-20	21:04:00	2002-07-20	21:30:00 2002-07-20
21:54:00						
37	38	X3.3	1998-11-28	04:54:00	1998-11-28	05:52:00 1998-11-28
06:13:00						
38	39	X3.2	2013-05-14	00:00:00	2013-05-14	01:11:00 2013-05-14
01:20:00						
39	40	X3.1	2014-10-24	21:07:00	2014-10-24	21:41:00 2014-10-24

22:13:00						
40	41	X3.1	2002-08-24	00:49:00	2002-08-24	01:12:00 2002-08-24
01:31:00						
41	42	X3	2002-07-15	19:59:00	2002-07-15	20:08:00 2002-07-15
20:14:00						
42	43	X2.8	2013-05-13	15:48:00	2013-05-13	16:05:00 2013-05-13
16:16:00						
43	44	X2.8	2001-12-11	07:58:00	2001-12-11	08:08:00 2001-12-11
08:14:00						
44	45	X2.8	1998-08-18	08:14:00	1998-08-18	08:24:00 1998-08-18
08:32:00						
45	46	X2.7	2015-05-05	22:05:00	2015-05-05	22:11:00 2015-05-05
22:15:00						
46	47	X2.7	2003-11-03	01:09:00	2003-11-03	01:30:00 2003-11-03
01:45:00						
47	48	X2.7	1998-05-06	07:58:00	1998-05-06	08:09:00 1998-05-06
08:20:00						
48	49	X2.6	2005-01-15	22:25:00	2005-01-15	23:02:00 2005-01-15
23:31:00						
49	50	X2.6	2001-09-24	09:32:00	2001-09-24	10:38:00 2001-09-24
11:09:00						

	Region
0	486
1	9393
2	486
3	808
4	9415
5	486
6	8100
7	2673
8	930
9	486
10	2673
11	720
12	1263
13	930
14	808
15	9733
16	9077
17	9415
18	1429
19	808
20	486
21	9591
22	1990
23	8307
24	39
25	9236
26	488

```

27      8307
28      720
29      8384
30      808
31      649
32      365
33      930
34      9767
35      1890
36      39
37      8395
38      1748
39      2192
40      69
41      30
42      1748
43      9733
44      8307
45      2339
46      488
47      8210
48      720
49      9632

```

#Part 3

```

from bs4 import BeautifulSoup
import urllib3
import requests
#http = urllib3.PoolManager()
r = requests.get("https://cmssc320.github.io/files/waves_type2.html")
rSoup2 = BeautifulSoup(r.content, 'html.parser')
rTable2 = rSoup2.find('pre')
rText = rTable2.get_text()
entries = rText.split('\n')
row_num = 0
for i in range(0,12):
    entries.pop(0)
entries.pop(len(entries)-1)
entries.pop(len(entries)-1)
entries.pop(len(entries)-1)
df2 = pd.DataFrame(index=range(0, len(entries)), columns=['Start
Date', 'Start Time', 'End Date', 'End Time', 'Start Freq', 'End Freq',
'Flare Location', 'Flare Region', 'Flare Class', 'CME Date', 'CME
Time', 'CME Angle', 'CME Width', 'CME Speed'])
for entry in entries:
    cols = entry.split(' ')
    while '' in cols:
        cols.remove('')
    col_num = 0
    while col_num < 14:

```



```

df2.iat[row_num, col_num] = cols[col_num]
col_num += 1
row_num += 1

```

df2

	Start Date	Start Time	End Date	End Time	Start Freq	End Freq	\
0	1997/04/01	14:00	04/01	14:15	8000	4000	
1	1997/04/07	14:30	04/07	17:30	11000	1000	
2	1997/05/12	05:15	05/14	16:00	12000	80	
3	1997/05/21	20:20	05/21	22:00	5000	500	
4	1997/09/23	21:53	09/23	22:16	6000	2000	
...	
512	2017/07/23	05:27	07/23	06:12	4400	900	
513	2017/09/04	20:27	09/05	04:54	14000	210	
514	2017/09/06	12:05	09/07	08:00	16000	70	
515	2017/09/10	16:02	09/11	06:50	16000	150	
516	2017/09/12	07:38	09/12	07:43	16000	13000	

	Flare Location	Flare Region	Flare Class	CME Date	CME Time	CME
0	S25E16	8026	M1.3	04/01	15:18	
74						
1	S28E19	8027	C6.8	04/07	14:27	
Halo						
2	N21W08	8038	C1.3	05/12	05:30	
Halo						
3	N05W12	8040	M1.3	05/21	21:00	
263						
4	S29E25	8088	C1.4	09/23	22:02	
133						
...
.						
512	BACK	-----	----	07/23	04:48	
Halo						
513	S10W12	12673	M5.5	09/04	20:12	
Halo						
514	S08W33	12673	X9.3	09/06	12:24	
Halo						
515	S09W92	-----	X8.3	09/10	16:00	
Halo						
516	N08E48	12680	C3.0	09/12	08:03	
124						

	CME Width	CME Speed
0	79	312
1	360	878
2	360	464
3	165	296
4	155	712

```

..      ...      ...
512      360      1848
513      360      1418
514      360      1571
515      360      3163
516      96       252

```

[517 rows x 14 columns]

#Step 4

```

tidydf2 = df2.replace('----', 'NaN').replace('-----', 'NaN')
tidydf2['Is Halo?'] = tidydf2['CME Angle'].map(lambda i: i == 'Halo')
tidydf2 = tidydf2.replace('Halo', 'NA')
#if first char of string is >, it is a lower bound
tidydf2['Width Lower Bound?'] = tidydf2['CME Width'].map(lambda i:
(str(i))[0] == '>')
tidydf2['CME Width'] = tidydf2['CME Width'].map(lambda i:
"".join(filter(str.isdigit, i)))
tidydf2['Start DateTime'] = pd.to_datetime(tidydf2['Start Date'] + ' '
+ tidydf2['Start Time'])
tidydf2['End DateTime'] = ""
tidydf2['CME DateTime'] = ""

```

#to get the date right so we can to pd.to_datetime

```

for row in range(0, len(entries)):
    dateArray = tidydf2.values[row][0].split('/')
    year = dateArray[0]
    end_daymonth = tidydf2.values[row][2]
    end_dt = year + '/' + end_daymonth
    tidydf2.iat[row, 2] = end_dt
    cme_daymonth = tidydf2.values[row][9]
    cme_dt = year + '/' + cme_daymonth
    tidydf2.iat[row, 9] = cme_dt

```

display(tidydf2)

	Start Date	Start Time	End Date	End Time	Start Freq	End Freq	\
0	1997/04/01	14:00	1997/04/01	14:15	8000	4000	
1	1997/04/07	14:30	1997/04/07	17:30	11000	1000	
2	1997/05/12	05:15	1997/05/14	16:00	12000	80	
3	1997/05/21	20:20	1997/05/21	22:00	5000	500	
4	1997/09/23	21:53	1997/09/23	22:16	6000	2000	
..	
512	2017/07/23	05:27	2017/07/23	06:12	4400	900	
513	2017/09/04	20:27	2017/09/05	04:54	14000	210	
514	2017/09/06	12:05	2017/09/07	08:00	16000	70	
515	2017/09/10	16:02	2017/09/11	06:50	16000	150	
516	2017/09/12	07:38	2017/09/12	07:43	16000	13000	

Flare Location Flare Region Flare Class CME Date CME Time CME

Angle \					
0	S25E16	8026	M1.3	1997/04/01	15:18
74					
1	S28E19	8027	C6.8	1997/04/07	14:27
NA					
2	N21W08	8038	C1.3	1997/05/12	05:30
NA					
3	N05W12	8040	M1.3	1997/05/21	21:00
263					
4	S29E25	8088	C1.4	1997/09/23	22:02
133					
..
...					
512	BACK	NaN	NaN	2017/07/23	04:48
NA					
513	S10W12	12673	M5.5	2017/09/04	20:12
NA					
514	S08W33	12673	X9.3	2017/09/06	12:24
NA					
515	S09W92	NaN	X8.3	2017/09/10	16:00
NA					
516	N08E48	12680	C3.0	2017/09/12	08:03
124					

	CME Width	CME Speed	Is Halo?	Width Lower Bound?	Start
DateTime \					
0	79	312	False	False	1997-04-01
14:00:00					
1	360	878	True	False	1997-04-07
14:30:00					
2	360	464	True	False	1997-05-12
05:15:00					
3	165	296	False	False	1997-05-21
20:20:00					
4	155	712	False	False	1997-09-23
21:53:00					
..
...					
512	360	1848	True	False	2017-07-23
05:27:00					
513	360	1418	True	False	2017-09-04
20:27:00					
514	360	1571	True	False	2017-09-06
12:05:00					
515	360	3163	True	False	2017-09-10
16:02:00					
516	96	252	False	False	2017-09-12
07:38:00					

End DateTime CME DateTime

```

0
1
2
3
4
..      ...      ...
512
513
514
515
516

```

[517 rows x 19 columns]

```

from datetime import datetime
tidydf2 = tidydf2[['End Date', 'End Time', 'CME Date', 'CME Time',
'Start DateTime', 'End DateTime', 'Start Freq', 'End Freq', 'Flare
Location', 'Flare Region', 'Flare Class', 'CME DateTime', 'CME Angle',
'CME Width', 'CME Speed', 'Is Halo?', 'Width Lower Bound?']]
for row in range(0, len(entries)):
    timeCheckEnd = tidydf2.values[row][1].split(':')
    timeCheckCME = tidydf2.values[row][3].split(':')
    #if the time is 24, leave blank
    if (int(timeCheckEnd[0]) < 24):
        tidydf2.iat[row, 5] = tidydf2.values[row][0].replace('/', '-')
+ ' ' + tidydf2.values[row][1] + ':00'
    if (str(timeCheckCME[0]) == "--"):
        tidydf2.iat[row, 11] = ""
    else: tidydf2.iat[row, 11] = tidydf2.values[row]
[2].replace('/', '-') + ' ' + tidydf2.values[row][3] + ':00'
tidydf2 = tidydf2.drop('End Date', 1)
tidydf2 = tidydf2.drop('End Time', 1)
tidydf2 = tidydf2.drop('CME Date', 1)
tidydf2 = tidydf2.drop('CME Time', 1)
display(tidydf2)

```

/var/folders/yl/0_18jst15nb9gbl2n_j_z5tw0000gn/T/
ipykernel_3763/1317046877.py:12: FutureWarning: In a future version of
pandas all arguments of DataFrame.drop except for the argument
'labels' will be keyword-only.

```
tidydf2 = tidydf2.drop('End Date', 1)
```

/var/folders/yl/0_18jst15nb9gbl2n_j_z5tw0000gn/T/ipykernel_3763/131704
6877.py:13: FutureWarning: In a future version of pandas all arguments
of DataFrame.drop except for the argument 'labels' will be keyword-
only.

```
tidydf2 = tidydf2.drop('End Time', 1)
```

/var/folders/yl/0_18jst15nb9gbl2n_j_z5tw0000gn/T/ipykernel_3763/131704
6877.py:14: FutureWarning: In a future version of pandas all arguments
of DataFrame.drop except for the argument 'labels' will be keyword-
only.

```
tidydf2 = tidydf2.drop('CME Date', 1)
/var/folders/yl/0_18jst15nb9gbl2n_j_z5tw0000gn/T/ipykernel_3763/131704
6877.py:15: FutureWarning: In a future version of pandas all arguments
of DataFrame.drop except for the argument 'labels' will be keyword-
only.
```

```
tidydf2 = tidydf2.drop('CME Time', 1)
```

	Start DateTime	End DateTime	Start Freq	End Freq	\
0	1997-04-01 14:00:00	1997-04-01 14:15:00	8000	4000	
1	1997-04-07 14:30:00	1997-04-07 17:30:00	11000	1000	
2	1997-05-12 05:15:00	1997-05-14 16:00:00	12000	80	
3	1997-05-21 20:20:00	1997-05-21 22:00:00	5000	500	
4	1997-09-23 21:53:00	1997-09-23 22:16:00	6000	2000	
...	
512	2017-07-23 05:27:00	2017-07-23 06:12:00	4400	900	
513	2017-09-04 20:27:00	2017-09-05 04:54:00	14000	210	
514	2017-09-06 12:05:00	2017-09-07 08:00:00	16000	70	
515	2017-09-10 16:02:00	2017-09-11 06:50:00	16000	150	
516	2017-09-12 07:38:00	2017-09-12 07:43:00	16000	13000	

	Flare Location	Flare Region	Flare Class	CME DateTime	CME
Angle \					
0	S25E16	8026	M1.3	1997-04-01 15:18:00	
74					
1	S28E19	8027	C6.8	1997-04-07 14:27:00	
NA					
2	N21W08	8038	C1.3	1997-05-12 05:30:00	
NA					
3	N05W12	8040	M1.3	1997-05-21 21:00:00	
263					
4	S29E25	8088	C1.4	1997-09-23 22:02:00	
133					
...	
...					
512	BACK	NaN	NaN	2017-07-23 04:48:00	
NA					
513	S10W12	12673	M5.5	2017-09-04 20:12:00	
NA					
514	S08W33	12673	X9.3	2017-09-06 12:24:00	
NA					
515	S09W92	NaN	X8.3	2017-09-10 16:00:00	
NA					
516	N08E48	12680	C3.0	2017-09-12 08:03:00	
124					

	CME Width	CME Speed	Is Halo?	Width Lower Bound?
0	79	312	False	False
1	360	878	True	False
2	360	464	True	False
3	165	296	False	False

4	155	712	False	False
...
512	360	1848	True	False
513	360	1418	True	False
514	360	1571	True	False
515	360	3163	True	False
516	96	252	False	False

[517 rows x 13 columns]

#Part 2

#Question 1

```
top50SF = tidydf2.loc[tidydf2['Flare Class'].str.contains('X')]
top50SF['Flare Class'] = top50SF['Flare Class'].str.lstrip('X')
```

#float to compare

```
top50SF = top50SF.astype({'Flare Class': float})
top50SF = top50SF.sort_values('Flare Class', ascending = False)
top50SF = top50SF.head(50)
top50SF = top50SF.astype({'Flare Class': str})
top50SF['Flare Class'] = "X" + top50SF['Flare Class']
display(top50SF)
```

/var/folders/yl/0_18jst15nb9gbl2n_j_z5tw0000gn/T/

ipykernel_3763/1127006976.py:4: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation:

https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
top50SF['Flare Class'] = top50SF['Flare Class'].str.lstrip('X')
```

	Start DateTime	End DateTime	Start Freq	End Freq	\
240	2003-11-04 20:00:00		10000	200	
117	2001-04-02 22:05:00	2001-04-03 02:30:00	14000	250	
233	2003-10-28 11:10:00		14000	40	
126	2001-04-15 14:05:00	2001-04-16 13:00:00	14000	40	
234	2003-10-29 20:55:00		11000	500	
8	1997-11-06 12:20:00	1997-11-07 08:30:00	14000	100	
514	2017-09-06 12:05:00	2017-09-07 08:00:00	16000	70	
328	2006-12-05 10:50:00	2006-12-05 20:00:00	14000	250	
237	2003-11-02 17:30:00	2003-11-03 01:00:00	12000	250	
515	2017-09-10 16:02:00	2017-09-11 06:50:00	16000	150	
288	2005-01-20 07:15:00	2005-01-20 16:30:00	14000	25	
359	2011-08-09 08:20:00	2011-08-09 08:35:00	16000	4000	
331	2006-12-06 19:00:00		16000	30	
317	2005-09-09 19:45:00	2005-09-09 22:00:00	10000	50	
82	2000-07-14 10:30:00	2000-07-15 14:30:00	14000	80	
121	2001-04-06 19:35:00	2001-04-07 01:50:00	14000	230	
375	2012-03-07 01:00:00	2012-03-08 19:00:00	16000	30	
135	2001-08-25 16:50:00	2001-08-25 23:00:00	8000	170	

443	2014-02-25 00:56:00	2014-02-25 11:28:00	14000	100
193	2002-07-23 00:50:00	2002-07-23 04:00:00	11000	400
104	2000-11-26 17:00:00	2000-11-26 17:15:00	14000	7000
239	2003-11-03 10:00:00	2003-11-03 12:30:00	6000	400
286	2005-01-17 10:00:00	2005-01-17 10:35:00	6100	1500
222	2003-05-28 01:00:00	2003-05-29 00:30:00	1000	200
332	2006-12-13 02:45:00	2006-12-13 10:40:00	12000	150
160	2001-12-28 20:35:00	2001-12-29 03:00:00	14000	350
192	2002-07-20 21:30:00	2002-07-20 22:20:00	10000	2000
404	2013-05-14 01:16:00	2013-05-14 08:20:00	16000	240
201	2002-08-24 01:45:00	2002-08-24 03:25:00	5000	400
403	2013-05-13 16:15:00	2013-05-13 19:10:00	16000	300
487	2015-05-05 22:24:00	2015-05-05 23:14:00	14000	500
19	1998-05-06 08:25:00	1998-05-06 08:35:00	14000	5000
238	2003-11-03 01:15:00	2003-11-03 01:25:00	3000	1500
284	2005-01-15 23:00:00	2005-01-17 00:00:00	3000	40
142	2001-09-24 10:45:00	2001-09-25 20:00:00	7000	30
9	1997-11-27 13:30:00	1997-11-27 14:00:00	14000	7000
276	2004-11-10 02:25:00	2004-11-10 03:40:00	14000	1000
123	2001-04-10 05:24:00		14000	100
99	2000-11-24 15:25:00	2000-11-24 22:00:00	14000	200
73	2000-06-06 15:20:00	2000-06-08 09:00:00	14000	40
345	2011-02-15 02:10:00	2011-02-15 07:00:00	16000	400
318	2005-09-10 21:45:00	2005-09-11 01:00:00	14000	200
361	2011-09-06 22:30:00	2011-09-07 15:40:00	16000	150
420	2013-10-25 15:08:00	2013-10-25 22:32:00	16000	200
7	1997-11-04 06:00:00	1997-11-05 04:30:00	14000	100
98	2000-11-24 05:10:00	2000-11-24 15:00:00	14000	100
125	2001-04-12 10:20:00	2001-04-12 10:40:00	14000	7000
274	2004-11-07 16:25:00	2004-11-08 20:00:00	14000	60
285	2005-01-17 09:25:00	2005-01-17 16:00:00	14000	30
102	2000-11-25 19:00:00	2000-11-25 19:35:00	6000	2000

	Flare Location	Flare Region	Flare Class	CME DateTime	CME
Angle \					
240	S19W83	10486	X28.0	2003-11-04 19:54:00	
NA					
117	N19W72	9393	X20.0	2001-04-02 22:06:00	
261					
233	S16E08	10486	X17.0	2003-10-28 11:30:00	
NA					
126	S20W85	9415	X14.0	2001-04-15 14:06:00	
245					
234	S15W02	10486	X10.0	2003-10-29 20:54:00	
NA					
8	S18W63	8100	X9.4	1997-11-06 12:10:00	
NA					
514	S08W33	12673	X9.3	2017-09-06 12:24:00	
NA					
328	S07E68	10930	X9.0		

NaN				
237	S14W56	10486	X8.3	2003-11-02 17:30:00
NA				
515	S09W92	NaN	X8.3	2017-09-10 16:00:00
NA				
288	N14W61	10720	X7.1	2005-01-20 06:54:00
NA				
359	N17W69	11263	X6.9	2011-08-09 08:12:00
NA				
331	S05E64	10930	X6.5	
NaN				
317	S12E67	10808	X6.2	2005-09-09 19:48:00
NA				
82	N22W07	9077	X5.7	2000-07-14 10:54:00
NA				
121	S21E31	9415	X5.6	2001-04-06 19:30:00
NA				
375	N17E27	11429	X5.4	2012-03-07 00:24:00
NA				
135	S17E34	9591	X5.3	2001-08-25 16:50:00
NA				
443	S12E82	11990	X4.9	2014-02-25 01:25:00
NA				
193	S13E72	10039	X4.8	2002-07-23 00:42:00
NA				
104	N18W38	9236	X4.0	2000-11-26 17:06:00
NA				
239	N08W77	10488	X3.9	2003-11-03 10:06:00
293				
286	N15W25	10720	X3.8	2005-01-17 09:54:00
NA				
222	S07W20	10365	X3.6	2003-05-28 00:50:00
NA				
332	S06W23	10930	X3.4	2006-12-13 02:54:00
NA				
160	S26E90	9756	X3.4	2001-12-28 20:30:00
NA				
192	S13E90	10039	X3.3	2002-07-20 22:06:00
NA				
404	N08E77	11748	X3.2	2013-05-14 01:25:00
NA				
201	S02W81	10069	X3.1	2002-08-24 01:27:00
NA				
403	N11E85	11748	X2.8	2013-05-13 16:07:00
NA				
487	N15E79	12339	X2.7	2015-05-05 22:24:00
NA				
19	S11W65	8210	X2.7	1998-05-06 08:29:00
309				
238	N10W83	10488	X2.7	2003-11-03 01:59:00

304				
284	N15W05	10720	X2.6	2005-01-15 23:06:00
NA				
142	S16E23	9632	X2.6	2001-09-24 10:30:00
NA				
9	N17E63	8113	X2.6	1997-11-27 13:56:00
98				
276	N09W49	10696	X2.5	2004-11-10 02:26:00
NA				
123	S23W09	9415	X2.3	2001-04-10 05:30:00
NA				
99	N22W07	9236	X2.3	2000-11-24 15:30:00
NA				
73	N20E18	9026	X2.3	2000-06-06 15:54:00
NA				
345	S20W12	11158	X2.2	2011-02-15 02:24:00
NA				
318	S13E47	10808	X2.1	2005-09-10 21:52:00
NA				
361	N14W18	11283	X2.1	2011-09-06 23:05:00
NA				
420	S06E69	11882	X2.1	2013-10-25 15:12:00
NA				
7	S14W33	8100	X2.1	1997-11-04 06:10:00
NA				
98	N20W05	9236	X2.0	2000-11-24 05:30:00
NA				
125	S19W43	9415	X2.0	2001-04-12 10:31:00
NA				
274	N09W17	10696	X2.0	2004-11-07 16:54:00
NA				
285	N15W25	10720	X2.0	2005-01-17 09:30:00
NA				
102	N20W23	9236	X1.9	2000-11-25 19:31:00
NA				

	CME Width	CME Speed	Is Halo?	Width Lower Bound?
240	360	2657	True	False
117	244	2505	False	False
233	360	2459	True	False
126	167	1199	False	False
234	360	2029	True	False
8	360	1556	True	False
514	360	1571	True	False
328		NaN	False	False
237	360	2598	True	False
515	360	3163	True	False
288	360	882	True	False
359	360	1610	True	False
331		NaN	False	False

317	360	2257	True	False
82	360	1674	True	False
121	360	1270	True	False
375	360	2684	True	False
135	360	1433	True	False
443	360	2147	True	False
193	360	2285	True	False
104	360	980	True	False
239	103	1420	False	False
286	360	2547	True	False
222	360	1366	True	False
332	360	1774	True	False
160	360	2216	True	False
192	360	1941	True	False
404	360	2625	True	False
201	360	1913	True	False
403	360	1850	True	False
487	360	715	True	False
19	190	1099	False	False
238	65	827	False	False
284	360	2861	True	False
142	360	2402	True	False
9	91	441	False	False
276	360	3387	True	False
123	360	2411	True	False
99	360	1245	True	False
73	360	1119	True	False
345	360	669	True	False
318	360	1893	True	False
361	360	575	True	False
420	360	1081	True	False
7	360	785	True	False
98	360	1289	True	False
125	360	1184	True	False
274	360	1759	True	False
285	360	2094	True	False
102	360	671	True	False

```

compareNASA = top50SF[['Flare Class', 'Start DateTime', 'CME
DateTime', 'End DateTime', 'Flare Region']]
compareNASA = compareNASA.rename(columns={'Flare Class': 'X Class',
'CME DateTime': 'Max DateTime', 'Flare Region': 'Region'})
compareSWL = tidydf[['X Class', 'Start DateTime', 'Max DateTime', 'End
DateTime', 'Region']]
compareSWL.iat[0,0] = 'X28.0'
compareSWL.iat[1,0] = 'X20.0'
compareSWL.iat[2,0] = 'X17.2'
compareSWL.iat[3,0] = 'X17.0'
for row in range(0,50):
    if (compareNASA.iat[row,4] != 'NaN'):

```

```

tempInt = int(compareNASA.iat[row,4])
if (tempInt >= 10000):
    tempInt -= 10000
    compareNASA.iat[row,4] = tempInt
display(compareSWL)
display(compareNASA)
#Most of the top 50 solar flares on spaceweatherlive.com are the same
ones that I acquired from NASA.
#While some of the information is missing (end datetimes, cme
datetimes), most of the relevant information is there.
#Also, while most of the regions are different, those that have the
same region in the SWL top 50
#also contain the same region in NASA's data.

```

	X Class	Start DateTime	Max DateTime	End DateTime
Region				
0	X28.0	2003-11-04 19:29:00	2003-11-04 19:53:00	2003-11-04 20:06:00
486				
1	X20.0	2001-04-02 21:32:00	2001-04-02 21:51:00	2001-04-02 22:03:00
9393				
2	X17.2	2003-10-28 09:51:00	2003-10-28 11:10:00	2003-10-28 11:24:00
486				
3	X17.0	2005-09-07 17:17:00	2005-09-07 17:40:00	2005-09-07 18:03:00
808				
4	X14.4	2001-04-15 13:19:00	2001-04-15 13:50:00	2001-04-15 13:55:00
9415				
5	X10	2003-10-29 20:37:00	2003-10-29 20:49:00	2003-10-29 21:01:00
486				
6	X9.4	1997-11-06 11:49:00	1997-11-06 11:55:00	1997-11-06 12:01:00
8100				
7	X9.3	2017-09-06 11:53:00	2017-09-06 12:02:00	2017-09-06 12:10:00
2673				
8	X9	2006-12-05 10:18:00	2006-12-05 10:35:00	2006-12-05 10:45:00
930				
9	X8.3	2003-11-02 17:03:00	2003-11-02 17:25:00	2003-11-02 17:39:00
486				
10	X8.2	2017-09-10 15:35:00	2017-09-10 16:06:00	2017-09-10 16:31:00
2673				
11	X7.1	2005-01-20 06:36:00	2005-01-20 07:01:00	2005-01-20 07:26:00
720				
12	X6.9	2011-08-09 07:48:00	2011-08-09 08:05:00	2011-08-09 08:08:00
1263				
13	X6.5	2006-12-06 18:29:00	2006-12-06 18:47:00	2006-12-06 19:00:00
930				
14	X6.2	2005-09-09 19:13:00	2005-09-09 20:04:00	2005-09-09 20:36:00
808				
15	X6.2	2001-12-13 14:20:00	2001-12-13 14:30:00	2001-12-13 14:35:00
9733				
16	X5.7	2000-07-14 10:03:00	2000-07-14 10:24:00	2000-07-14 10:43:00
9077				

17 9415	X5.6	2001-04-06	19:10:00	2001-04-06	19:21:00	2001-04-06	19:31:00
18 1429	X5.4	2012-03-07	00:02:00	2012-03-07	00:24:00	2012-03-07	00:40:00
19 808	X5.4	2005-09-08	20:52:00	2005-09-08	21:06:00	2005-09-08	21:17:00
20 486	X5.4	2003-10-23	08:19:00	2003-10-23	08:35:00	2003-10-23	08:49:00
21 9591	X5.3	2001-08-25	16:23:00	2001-08-25	16:45:00	2001-08-25	17:04:00
22 1990	X4.9	2014-02-25	00:39:00	2014-02-25	00:49:00	2014-02-25	01:03:00
23 8307	X4.9	1998-08-18	22:10:00	1998-08-18	22:19:00	1998-08-18	22:28:00
24 39	X4.8	2002-07-23	00:18:00	2002-07-23	00:35:00	2002-07-23	00:47:00
25 9236	X4	2000-11-26	16:34:00	2000-11-26	16:48:00	2000-11-26	16:56:00
26 488	X3.9	2003-11-03	09:43:00	2003-11-03	09:55:00	2003-11-03	10:19:00
27 8307	X3.9	1998-08-19	21:35:00	1998-08-19	21:45:00	1998-08-19	21:50:00
28 720	X3.8	2005-01-17	06:59:00	2005-01-17	09:52:00	2005-01-17	10:07:00
29 8384	X3.7	1998-11-22	06:30:00	1998-11-22	06:42:00	1998-11-22	06:49:00
30 808	X3.6	2005-09-09	09:42:00	2005-09-09	09:59:00	2005-09-09	10:08:00
31 649	X3.6	2004-07-16	13:49:00	2004-07-16	13:55:00	2004-07-16	14:01:00
32 365	X3.6	2003-05-28	00:17:00	2003-05-28	00:27:00	2003-05-28	00:39:00
33 930	X3.4	2006-12-13	02:14:00	2006-12-13	02:40:00	2006-12-13	02:57:00
34 9767	X3.4	2001-12-28	20:02:00	2001-12-28	20:45:00	2001-12-28	21:32:00
35 1890	X3.3	2013-11-05	22:07:00	2013-11-05	22:12:00	2013-11-05	22:15:00
36 39	X3.3	2002-07-20	21:04:00	2002-07-20	21:30:00	2002-07-20	21:54:00
37 8395	X3.3	1998-11-28	04:54:00	1998-11-28	05:52:00	1998-11-28	06:13:00
38 1748	X3.2	2013-05-14	00:00:00	2013-05-14	01:11:00	2013-05-14	01:20:00
39 2192	X3.1	2014-10-24	21:07:00	2014-10-24	21:41:00	2014-10-24	22:13:00
40 69	X3.1	2002-08-24	00:49:00	2002-08-24	01:12:00	2002-08-24	01:31:00
41 30	X3	2002-07-15	19:59:00	2002-07-15	20:08:00	2002-07-15	20:14:00

42 1748	X2.8	2013-05-13	15:48:00	2013-05-13	16:05:00	2013-05-13	16:16:00
43 9733	X2.8	2001-12-11	07:58:00	2001-12-11	08:08:00	2001-12-11	08:14:00
44 8307	X2.8	1998-08-18	08:14:00	1998-08-18	08:24:00	1998-08-18	08:32:00
45 2339	X2.7	2015-05-05	22:05:00	2015-05-05	22:11:00	2015-05-05	22:15:00
46 488	X2.7	2003-11-03	01:09:00	2003-11-03	01:30:00	2003-11-03	01:45:00
47 8210	X2.7	1998-05-06	07:58:00	1998-05-06	08:09:00	1998-05-06	08:20:00
48 720	X2.6	2005-01-15	22:25:00	2005-01-15	23:02:00	2005-01-15	23:31:00
49 9632	X2.6	2001-09-24	09:32:00	2001-09-24	10:38:00	2001-09-24	11:09:00

	X Class	Start DateTime	Max DateTime	End
240	X28.0	2003-11-04 20:00:00	2003-11-04 19:54:00	
117 02:30:00	X20.0	2001-04-02 22:05:00	2001-04-02 22:06:00	2001-04-03
233	X17.0	2003-10-28 11:10:00	2003-10-28 11:30:00	
126 13:00:00	X14.0	2001-04-15 14:05:00	2001-04-15 14:06:00	2001-04-16
234	X10.0	2003-10-29 20:55:00	2003-10-29 20:54:00	
8 08:30:00	X9.4	1997-11-06 12:20:00	1997-11-06 12:10:00	1997-11-07
514 08:00:00	X9.3	2017-09-06 12:05:00	2017-09-06 12:24:00	2017-09-07
328 20:00:00	X9.0	2006-12-05 10:50:00		2006-12-05
237 01:00:00	X8.3	2003-11-02 17:30:00	2003-11-02 17:30:00	2003-11-03
515 06:50:00	X8.3	2017-09-10 16:02:00	2017-09-10 16:00:00	2017-09-11
288 16:30:00	X7.1	2005-01-20 07:15:00	2005-01-20 06:54:00	2005-01-20
359 08:35:00	X6.9	2011-08-09 08:20:00	2011-08-09 08:12:00	2011-08-09
331	X6.5	2006-12-06 19:00:00		
317 22:00:00	X6.2	2005-09-09 19:45:00	2005-09-09 19:48:00	2005-09-09
82 14:30:00	X5.7	2000-07-14 10:30:00	2000-07-14 10:54:00	2000-07-15
121	X5.6	2001-04-06 19:35:00	2001-04-06 19:30:00	2001-04-07

01:50:00						
375	X5.4	2012-03-07 01:00:00	2012-03-07 00:24:00	2012-03-08 19:00:00		
135	X5.3	2001-08-25 16:50:00	2001-08-25 16:50:00	2001-08-25 23:00:00		
443	X4.9	2014-02-25 00:56:00	2014-02-25 01:25:00	2014-02-25 11:28:00		
193	X4.8	2002-07-23 00:50:00	2002-07-23 00:42:00	2002-07-23 04:00:00		
104	X4.0	2000-11-26 17:00:00	2000-11-26 17:06:00	2000-11-26 17:15:00		
239	X3.9	2003-11-03 10:00:00	2003-11-03 10:06:00	2003-11-03 12:30:00		
286	X3.8	2005-01-17 10:00:00	2005-01-17 09:54:00	2005-01-17 10:35:00		
222	X3.6	2003-05-28 01:00:00	2003-05-28 00:50:00	2003-05-29 00:30:00		
332	X3.4	2006-12-13 02:45:00	2006-12-13 02:54:00	2006-12-13 10:40:00		
160	X3.4	2001-12-28 20:35:00	2001-12-28 20:30:00	2001-12-29 03:00:00		
192	X3.3	2002-07-20 21:30:00	2002-07-20 22:06:00	2002-07-20 22:20:00		
404	X3.2	2013-05-14 01:16:00	2013-05-14 01:25:00	2013-05-14 08:20:00		
201	X3.1	2002-08-24 01:45:00	2002-08-24 01:27:00	2002-08-24 03:25:00		
403	X2.8	2013-05-13 16:15:00	2013-05-13 16:07:00	2013-05-13 19:10:00		
487	X2.7	2015-05-05 22:24:00	2015-05-05 22:24:00	2015-05-05 23:14:00		
19	X2.7	1998-05-06 08:25:00	1998-05-06 08:29:00	1998-05-06 08:35:00		
238	X2.7	2003-11-03 01:15:00	2003-11-03 01:59:00	2003-11-03 01:25:00		
284	X2.6	2005-01-15 23:00:00	2005-01-15 23:06:00	2005-01-17 00:00:00		
142	X2.6	2001-09-24 10:45:00	2001-09-24 10:30:00	2001-09-25 20:00:00		
9	X2.6	1997-11-27 13:30:00	1997-11-27 13:56:00	1997-11-27 14:00:00		
276	X2.5	2004-11-10 02:25:00	2004-11-10 02:26:00	2004-11-10 03:40:00		
123	X2.3	2001-04-10 05:24:00	2001-04-10 05:30:00			
99	X2.3	2000-11-24 15:25:00	2000-11-24 15:30:00	2000-11-24 22:00:00		
73	X2.3	2000-06-06 15:20:00	2000-06-06 15:54:00	2000-06-08 09:00:00		
345	X2.2	2011-02-15 02:10:00	2011-02-15 02:24:00	2011-02-15		

07:00:00						
318	X2.1	2005-09-10 21:45:00	2005-09-10 21:52:00	2005-09-11 01:00:00		
361	X2.1	2011-09-06 22:30:00	2011-09-06 23:05:00	2011-09-07 15:40:00		
420	X2.1	2013-10-25 15:08:00	2013-10-25 15:12:00	2013-10-25 22:32:00		
7	X2.1	1997-11-04 06:00:00	1997-11-04 06:10:00	1997-11-05 04:30:00		
98	X2.0	2000-11-24 05:10:00	2000-11-24 05:30:00	2000-11-24 15:00:00		
125	X2.0	2001-04-12 10:20:00	2001-04-12 10:31:00	2001-04-12 10:40:00		
274	X2.0	2004-11-07 16:25:00	2004-11-07 16:54:00	2004-11-08 20:00:00		
285	X2.0	2005-01-17 09:25:00	2005-01-17 09:30:00	2005-01-17 16:00:00		
102	X1.9	2000-11-25 19:00:00	2000-11-25 19:31:00	2000-11-25 19:35:00		

	Region
240	486
117	9393
233	486
126	9415
234	486
8	8100
514	2673
328	930
237	486
515	NaN
288	720
359	1263
331	930
317	808
82	9077
121	9415
375	1429
135	9591
443	1990
193	39
104	9236
239	488
286	720
222	365
332	930
160	9756
192	39
404	1748
201	69

```

403    1748
487    2339
19     8210
238     488
284     720
142    9632
9      8113
276     696
123    9415
99     9236
73     9026
345    1158
318     808
361    1283
420    1882
7      8100
98     9236
125    9415
274     696
285     720
102    9236

```

#Question 2

```
compareNASA['Best Match?'] = pd.Series('NaN', index=compareNASA.index)
```

```
#display(compareNASA)
```

```
#Function that finds which flare it best matches with (if any)
```

```
def bestMatch(index):
```

```
    for row in range(0,50):
```

```
        #temp = row[5][1:]
```

```
        if (str(compareNASA.iat[index,0]) ==
```

```
str(compareSWL.iat[row,0]) and int(compareNASA.iat[index,4]) ==
```

```
int(compareSWL.iat[row,4])):
```

```
        #return best match rank if found
```

```
        return row
```

```
    # if no best match is found
```

```
    return 'NaN'
```

```
for i in range(0,50):
```

```
    if (compareNASA.iat[i, 4] != 'NaN'):
```

```
        temp = bestMatch(i)
```

```
        compareNASA.iat[i, 5] = temp
```

```
display(compareNASA)
```

	X Class	Start DateTime	Max DateTime	End
240	X28.0	2003-11-04 20:00:00	2003-11-04 19:54:00	
117	X20.0	2001-04-02 22:05:00	2001-04-02 22:06:00	2001-04-03 02:30:00
233	X17.0	2003-10-28 11:10:00	2003-10-28 11:30:00	

126	X14.0	2001-04-15	14:05:00	2001-04-15	14:06:00	2001-04-16
			13:00:00			
234	X10.0	2003-10-29	20:55:00	2003-10-29	20:54:00	
8	X9.4	1997-11-06	12:20:00	1997-11-06	12:10:00	1997-11-07
			08:30:00			
514	X9.3	2017-09-06	12:05:00	2017-09-06	12:24:00	2017-09-07
			08:00:00			
328	X9.0	2006-12-05	10:50:00			2006-12-05
			20:00:00			
237	X8.3	2003-11-02	17:30:00	2003-11-02	17:30:00	2003-11-03
			01:00:00			
515	X8.3	2017-09-10	16:02:00	2017-09-10	16:00:00	2017-09-11
			06:50:00			
288	X7.1	2005-01-20	07:15:00	2005-01-20	06:54:00	2005-01-20
			16:30:00			
359	X6.9	2011-08-09	08:20:00	2011-08-09	08:12:00	2011-08-09
			08:35:00			
331	X6.5	2006-12-06	19:00:00			
317	X6.2	2005-09-09	19:45:00	2005-09-09	19:48:00	2005-09-09
			22:00:00			
82	X5.7	2000-07-14	10:30:00	2000-07-14	10:54:00	2000-07-15
			14:30:00			
121	X5.6	2001-04-06	19:35:00	2001-04-06	19:30:00	2001-04-07
			01:50:00			
375	X5.4	2012-03-07	01:00:00	2012-03-07	00:24:00	2012-03-08
			19:00:00			
135	X5.3	2001-08-25	16:50:00	2001-08-25	16:50:00	2001-08-25
			23:00:00			
443	X4.9	2014-02-25	00:56:00	2014-02-25	01:25:00	2014-02-25
			11:28:00			
193	X4.8	2002-07-23	00:50:00	2002-07-23	00:42:00	2002-07-23
			04:00:00			
104	X4.0	2000-11-26	17:00:00	2000-11-26	17:06:00	2000-11-26
			17:15:00			
239	X3.9	2003-11-03	10:00:00	2003-11-03	10:06:00	2003-11-03
			12:30:00			
286	X3.8	2005-01-17	10:00:00	2005-01-17	09:54:00	2005-01-17
			10:35:00			
222	X3.6	2003-05-28	01:00:00	2003-05-28	00:50:00	2003-05-29
			00:30:00			
332	X3.4	2006-12-13	02:45:00	2006-12-13	02:54:00	2006-12-13
			10:40:00			
160	X3.4	2001-12-28	20:35:00	2001-12-28	20:30:00	2001-12-29
			03:00:00			
192	X3.3	2002-07-20	21:30:00	2002-07-20	22:06:00	2002-07-20
			22:20:00			
404	X3.2	2013-05-14	01:16:00	2013-05-14	01:25:00	2013-05-14
			08:20:00			

201	X3.1	2002-08-24 01:45:00	2002-08-24 01:27:00	2002-08-24 03:25:00
403	X2.8	2013-05-13 16:15:00	2013-05-13 16:07:00	2013-05-13 19:10:00
487	X2.7	2015-05-05 22:24:00	2015-05-05 22:24:00	2015-05-05 23:14:00
19	X2.7	1998-05-06 08:25:00	1998-05-06 08:29:00	1998-05-06 08:35:00
238	X2.7	2003-11-03 01:15:00	2003-11-03 01:59:00	2003-11-03 01:25:00
284	X2.6	2005-01-15 23:00:00	2005-01-15 23:06:00	2005-01-17 00:00:00
142	X2.6	2001-09-24 10:45:00	2001-09-24 10:30:00	2001-09-25 20:00:00
9	X2.6	1997-11-27 13:30:00	1997-11-27 13:56:00	1997-11-27 14:00:00
276	X2.5	2004-11-10 02:25:00	2004-11-10 02:26:00	2004-11-10 03:40:00
123	X2.3	2001-04-10 05:24:00	2001-04-10 05:30:00	
99	X2.3	2000-11-24 15:25:00	2000-11-24 15:30:00	2000-11-24 22:00:00
73	X2.3	2000-06-06 15:20:00	2000-06-06 15:54:00	2000-06-08 09:00:00
345	X2.2	2011-02-15 02:10:00	2011-02-15 02:24:00	2011-02-15 07:00:00
318	X2.1	2005-09-10 21:45:00	2005-09-10 21:52:00	2005-09-11 01:00:00
361	X2.1	2011-09-06 22:30:00	2011-09-06 23:05:00	2011-09-07 15:40:00
420	X2.1	2013-10-25 15:08:00	2013-10-25 15:12:00	2013-10-25 22:32:00
7	X2.1	1997-11-04 06:00:00	1997-11-04 06:10:00	1997-11-05 04:30:00
98	X2.0	2000-11-24 05:10:00	2000-11-24 05:30:00	2000-11-24 15:00:00
125	X2.0	2001-04-12 10:20:00	2001-04-12 10:31:00	2001-04-12 10:40:00
274	X2.0	2004-11-07 16:25:00	2004-11-07 16:54:00	2004-11-08 20:00:00
285	X2.0	2005-01-17 09:25:00	2005-01-17 09:30:00	2005-01-17 16:00:00
102	X1.9	2000-11-25 19:00:00	2000-11-25 19:31:00	2000-11-25 19:35:00

	Region	Best Match?
240	486	0
117	9393	1
233	486	NaN
126	9415	NaN

234	486	NaN
8	8100	6
514	2673	7
328	930	NaN
237	486	9
515	NaN	NaN
288	720	11
359	1263	12
331	930	13
317	808	14
82	9077	16
121	9415	17
375	1429	18
135	9591	21
443	1990	22
193	39	24
104	9236	NaN
239	488	26
286	720	28
222	365	32
332	930	33
160	9756	NaN
192	39	36
404	1748	38
201	69	40
403	1748	42
487	2339	45
19	8210	47
238	488	46
284	720	48
142	9632	49
9	8113	NaN
276	696	NaN
123	9415	NaN
99	9236	NaN
73	9026	NaN
345	1158	NaN
318	808	NaN
361	1283	NaN
420	1882	NaN
7	8100	NaN
98	9236	NaN
125	9415	NaN
274	696	NaN
285	720	NaN
102	9236	NaN

*#I did a couple things here to determine if the data I got from NASA
best matched with
#any ranks in the SWL top 50. To do this, I changed those with classes*

```

of X_+ to X_.0,
#if they had not already had a decimal, as well as decreasing any
region with a number
#more than 10000 by 10000, since I had realized the numbers were the
same otherwise.
#I assume this was likely due to some number of digits maximum, since
it only applied to
#those with regions that were 5 digits. Since many of the DateTime
values were also either
#missing or close to the actual times, I decided to first only focus
on those which had
#the same X Class and Region values, classifying those as flares with
a "Best Match".

#Basically, I went through every row in my top 50 solar flares I got
from NASA. For each
#row, I iterated through every row in the SWL data. If the X Class and
Region values
#are the same, I would put that ranking under the 'Best Match?' column
in my new dataframe.
#Therefore, there are many "best matches" in my NASA dataframe. Since
there is more than one
#SWL entry that best matches, I then looked towards the start and end
datetimes to determine
#which one is the "best match". I then compared all of the start/end
datetimes of best matches
#to see which one had the least difference. This, using the code
below, was determined to be
#solar flares 23 on the NASA data and 46 on SWL's list.
import datetime
bestMatches = pd.DataFrame(index=range(0,26), columns=['X Class',
'Region', 'Matching Rank', 'StartDT Diff', 'EndDT Diff', 'Total
DateTime Difference'])
bestMatches = bestMatches.astype({'StartDT Diff': str, 'EndDT Diff':
str, 'Total DateTime Difference': float})
i = 0
for row in range(0,50):
    if (compareNASA.iat[row,5] != 'NaN' and compareNASA.iat[row,3] !=
''):
        bestMatches.iat[i,0] = compareNASA.iat[row,0]
        bestMatches.iat[i,1] = compareNASA.iat[row,4]
        bestMatches.iat[i,2] = compareNASA.iat[row,5]
        for rank in range(0,50):
            if (str(compareNASA.iat[row,0]) ==
str(compareSWL.iat[rank,0]) and int(compareNASA.iat[row,4]) ==
int(compareSWL.iat[rank,4])):
                swlStartDT = compareSWL.iat[rank, 1]
                nasaStartDT = compareNASA.iat[row, 1]
                a = datetime.datetime.strptime(str(swlStartDT), "%Y-
%m-%d %H:%M:%S")

```

```

        b = datetime.datetime.strptime(str(nasaStartDT), "%Y-%m-%d %H:%M:%S")
        c = a-b
        swlEndDT = compareSWL.iat[rank, 3]
        nasaEndDT = compareNASA.iat[row, 3]
        d= datetime.datetime.strptime(str(swlEndDT), "%Y-%m-%d %H:%M:%S")
        e = datetime.datetime.strptime(str(nasaEndDT), "%Y-%m-%d %H:%M:%S")
        f = d-e
        bestMatches.iat[i,3] = c.total_seconds()
        bestMatches.iat[i,4] = f.total_seconds()
        i += 1
bestMatches = bestMatches.astype({'Total DateTime Difference': float})
for j in range(0,26):
    bestMatches.iat[j,5] = float(abs(bestMatches.iat[j,3]) +
abs(bestMatches.iat[j,4]))
bestMatches = bestMatches.sort_values('Total DateTime Difference',
ascending = True)
display(bestMatches)

```

	X	Class	Region	Matching	Rank	StartDT	Diff	EndDT	Diff \
23	X2.7	488		46		-360.0		1200.0	
22	X2.7	8210		47		-1620.0		-900.0	
17	X3.3	39		36		-1560.0		-1560.0	
5	X6.9	1263		12		-1920.0		-1620.0	
21	X2.7	2339		45		-1140.0		-3540.0	
6	X6.2	808		14		-1920.0		-5040.0	
13	X3.9	488		26		-1020.0		-7860.0	
19	X3.1	69		40		-3360.0		-6840.0	
20	X2.8	1748		42		-1620.0		-10440.0	
14	X3.8	720		28		-10860.0		-1680.0	
12	X4.8	39		24		-1920.0		-11580.0	
0	X20.0	9393		1		-1980.0		-16020.0	
10	X5.3	9591		21		-1620.0		-21360.0	
8	X5.6	9415		17		-1500.0		-22740.0	
3	X8.3	486		9		-1620.0		-26460.0	
16	X3.4	930		33		-1860.0		-27780.0	
18	X3.2	1748		38		-4560.0		-25200.0	
4	X7.1	720		11		-2340.0		-32640.0	
11	X4.9	1990		22		-1020.0		-37500.0	
2	X9.3	2673		7		-720.0		-71400.0	
1	X9.4	8100		6		-1860.0		-73740.0	
15	X3.6	365		32		-2580.0		-85860.0	
24	X2.6	720		48		-2100.0		-88140.0	
7	X5.7	9077		16		-1620.0		-100020.0	
25	X2.6	9632		49		-4380.0		-118260.0	
9	X5.4	1429		18		-3480.0		-152400.0	

	Total DateTime Difference
23	1560.0
22	2520.0
17	3120.0
5	3540.0
21	4680.0
6	6960.0
13	8880.0
19	10200.0
20	12060.0
14	12540.0
12	13500.0
0	18000.0
10	22980.0
8	24240.0
3	28080.0
16	29640.0
18	29760.0
4	34980.0
11	38520.0
2	72120.0
1	75600.0
15	88440.0
24	90240.0
7	101640.0
25	122640.0
9	155880.0

#Question 3

```
import numpy as np
import matplotlib_inline
import matplotlib
trueForSome = 0
falseforSome = trueForAll = falseForAll = trueForSome
for index, row in top50SF.iterrows():
    if row['Is Halo?']: trueForSome += 1
    else: falseforSome += 1
for index, row in tidydf2.iterrows():
    if row['Is Halo?']: trueForAll += 1
    else: falseForAll += 1
halo_data = np.array([[ '', 'NASA Top 50 Flares', 'All NASA Flares'],
                     [True, trueForSome, trueForAll],
                     [False, falseforSome, falseForAll]])
halo_df = pd.DataFrame(data=halo_data[1:,1:],
                      index=halo_data[1:,0],
                      columns=halo_data[0,1:])
halo_df['NASA Top 50 Flares'] = halo_df['NASA Top 50
Flares'].astype(str).astype(int)
halo_df['All NASA Flares'] = halo_df['All NASA
Flares'].astype(str).astype(int)
```

```
halo_df.plot.bar(color=['green', 'purple'])  
# This graph signifies that there is a correlation between being a top  
# 50 solar flare  
# and having a Halo CME.
```

<AxesSubplot:>

