CMSC320_P1

February 24, 2022

[1357]:
```
#Part 1
```

[1358]:
```
#Step 1
```

[1359]:
```
import requests
from bs4 import BeautifulSoup
from urllib.parse import urlparse
import pandas as pd
import numpy as np
```

[1360]:
```
res = requests.get("https://cmsc320.github.io/files/top-50-solar-flares.html")
#get request
```

[1361]:
```
root = BeautifulSoup(res.content)
```

[1362]:
```
a = root.find("table")
```

[1363]:
```
b = a.prettify()
```

[1364]:
```
all_frames = pd.read_html(b)
#use read_html to parser the html file
```

[1365]:
```
all_frames[0].columns = ['rank', 'x_classification', 'date', 'region',␣
 ↪'start_time', 'maximum_time', 'end_time', 'movie']
df = pd.DataFrame(all_frames[0])
#Build a dataframe
```

[1366]:
```
df
```

[1366]:

	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	Movie View archive
1	22:03	Movie View archive

2	11:24	Movie	View archive
3	18:03	Movie	View archive
4	13:55	Movie	View archive
5	21:01	Movie	View archive
6	12:01	Movie	View archive
7	12:10	Movie	View archive
8	10:45	Movie	View archive
9	17:39	Movie	View archive
10	16:31	Movie	View archive
11	07:26	Movie	View archive
12	08:08	Movie	View archive
13	19:00	Movie	View archive
14	20:36	Movie	View archive
15	14:35	Movie	View archive
16	10:43	Movie	View archive
17	19:31	Movie	View archive
18	00:40	Movie	View archive
19	21:17	Movie	View archive
20	08:49	Movie	View archive
21	17:04	Movie	View archive
22	01:03	Movie	View archive
23	22:28		View archive
24	00:47	Movie	View archive
25	16:56	Movie	View archive
26	10:19	Movie	View archive
27	21:50		View archive
28	10:07	Movie	View archive
29	06:49	Movie	View archive
30	10:08	Movie	View archive
31	14:01	Movie	View archive
32	00:39	Movie	View archive
33	02:57	Movie	View archive
34	21:32	Movie	View archive
35	22:15	Movie	View archive
36	21:54	Movie	View archive
37	06:13	Movie	View archive
38	01:20	Movie	View archive
39	22:13	Movie	View archive
40	01:31	Movie	View archive
41	20:14	Movie	View archive
42	16:16	Movie	View archive
43	08:14	Movie	View archive
44	08:32		View archive
45	22:15	Movie	View archive
46	01:45	Movie	View archive
47	08:20	Movie	View archive
48	23:31	Movie	View archive

```
49    11:09  Movie  View archive
```

[1367]:
```
#Step 2
```

[1368]:
```
df2 = df
df2 = df2.drop(['movie'],axis = 1)
#remove movie as mentioned in description
```

[1369]:
```
df2['start_datetime'] = pd.to_datetime(
                            df2['date'] + ' ' + df2['start_time'])
df2['max_datetime'] = pd.to_datetime(
                            df2['date'] + ' ' + df2['maximum_time'])
df2['end_datetime'] = pd.to_datetime(
                            df2['date'] + ' ' + df2['end_time'])
#Create three datetime columns by to_datetime
df2 = df2.drop(['date'],axis = 1)
df2 = df2.drop(['start_time'],axis = 1)
df2 = df2.drop(['maximum_time'],axis = 1)
df2 = df2.drop(['end_time'],axis = 1)
#Delete used columns
```

[1370]:
```
df2 = df2[['rank', 'x_classification', 'start_datetime', 'max_datetime',
       'end_datetime', 'region']]
#re-arrange the order of columns to better fit the sample output
```

[1371]:
```
df2.index = np.arange(1, 51)
# change index starting from 1
```

[1372]:
```
df2
```

[1372]:

	rank	x_classification	start_datetime	max_datetime	\
1	1	X28+	2003-11-04 19:29:00	2003-11-04 19:53:00	
2	2	X20+	2001-04-02 21:32:00	2001-04-02 21:51:00	
3	3	X17.2+	2003-10-28 09:51:00	2003-10-28 11:10:00	
4	4	X17+	2005-09-07 17:17:00	2005-09-07 17:40:00	
5	5	X14.4	2001-04-15 13:19:00	2001-04-15 13:50:00	
6	6	X10	2003-10-29 20:37:00	2003-10-29 20:49:00	
7	7	X9.4	1997-11-06 11:49:00	1997-11-06 11:55:00	
8	8	X9.3	2017-09-06 11:53:00	2017-09-06 12:02:00	
9	9	X9	2006-12-05 10:18:00	2006-12-05 10:35:00	
10	10	X8.3	2003-11-02 17:03:00	2003-11-02 17:25:00	
11	11	X8.2	2017-09-10 15:35:00	2017-09-10 16:06:00	
12	12	X7.1	2005-01-20 06:36:00	2005-01-20 07:01:00	
13	13	X6.9	2011-08-09 07:48:00	2011-08-09 08:05:00	
14	14	X6.5	2006-12-06 18:29:00	2006-12-06 18:47:00	
15	15	X6.2	2005-09-09 19:13:00	2005-09-09 20:04:00	
16	16	X6.2	2001-12-13 14:20:00	2001-12-13 14:30:00	

```
17    17             X5.7 2000-07-14 10:03:00 2000-07-14 10:24:00
18    18             X5.6 2001-04-06 19:10:00 2001-04-06 19:21:00
19    19             X5.4 2012-03-07 00:02:00 2012-03-07 00:24:00
20    20             X5.4 2005-09-08 20:52:00 2005-09-08 21:06:00
21    21             X5.4 2003-10-23 08:19:00 2003-10-23 08:35:00
22    22             X5.3 2001-08-25 16:23:00 2001-08-25 16:45:00
23    23             X4.9 2014-02-25 00:39:00 2014-02-25 00:49:00
24    24             X4.9 1998-08-18 22:10:00 1998-08-18 22:19:00
25    25             X4.8 2002-07-23 00:18:00 2002-07-23 00:35:00
26    26               X4 2000-11-26 16:34:00 2000-11-26 16:48:00
27    27             X3.9 2003-11-03 09:43:00 2003-11-03 09:55:00
28    28             X3.9 1998-08-19 21:35:00 1998-08-19 21:45:00
29    29             X3.8 2005-01-17 06:59:00 2005-01-17 09:52:00
30    30             X3.7 1998-11-22 06:30:00 1998-11-22 06:42:00
31    31             X3.6 2005-09-09 09:42:00 2005-09-09 09:59:00
32    32             X3.6 2004-07-16 13:49:00 2004-07-16 13:55:00
33    33             X3.6 2003-05-28 00:17:00 2003-05-28 00:27:00
34    34             X3.4 2006-12-13 02:14:00 2006-12-13 02:40:00
35    35             X3.4 2001-12-28 20:02:00 2001-12-28 20:45:00
36    36             X3.3 2013-11-05 22:07:00 2013-11-05 22:12:00
37    37             X3.3 2002-07-20 21:04:00 2002-07-20 21:30:00
38    38             X3.3 1998-11-28 04:54:00 1998-11-28 05:52:00
39    39             X3.2 2013-05-14 00:00:00 2013-05-14 01:11:00
40    40             X3.1 2014-10-24 21:07:00 2014-10-24 21:41:00
41    41             X3.1 2002-08-24 00:49:00 2002-08-24 01:12:00
42    42               X3 2002-07-15 19:59:00 2002-07-15 20:08:00
43    43             X2.8 2013-05-13 15:48:00 2013-05-13 16:05:00
44    44             X2.8 2001-12-11 07:58:00 2001-12-11 08:08:00
45    45             X2.8 1998-08-18 08:14:00 1998-08-18 08:24:00
46    46             X2.7 2015-05-05 22:05:00 2015-05-05 22:11:00
47    47             X2.7 2003-11-03 01:09:00 2003-11-03 01:30:00
48    48             X2.7 1998-05-06 07:58:00 1998-05-06 08:09:00
49    49             X2.6 2005-01-15 22:25:00 2005-01-15 23:02:00
50    50             X2.6 2001-09-24 09:32:00 2001-09-24 10:38:00
```

```
          end_datetime  region
1  2003-11-04 20:06:00     486
2  2001-04-02 22:03:00    9393
3  2003-10-28 11:24:00     486
4  2005-09-07 18:03:00     808
5  2001-04-15 13:55:00    9415
6  2003-10-29 21:01:00     486
7  1997-11-06 12:01:00    8100
8  2017-09-06 12:10:00    2673
9  2006-12-05 10:45:00     930
10 2003-11-02 17:39:00     486
11 2017-09-10 16:31:00    2673
```

```
12 2005-01-20 07:26:00     720
13 2011-08-09 08:08:00    1263
14 2006-12-06 19:00:00     930
15 2005-09-09 20:36:00     808
16 2001-12-13 14:35:00    9733
17 2000-07-14 10:43:00    9077
18 2001-04-06 19:31:00    9415
19 2012-03-07 00:40:00    1429
20 2005-09-08 21:17:00     808
21 2003-10-23 08:49:00     486
22 2001-08-25 17:04:00    9591
23 2014-02-25 01:03:00    1990
24 1998-08-18 22:28:00    8307
25 2002-07-23 00:47:00      39
26 2000-11-26 16:56:00    9236
27 2003-11-03 10:19:00     488
28 1998-08-19 21:50:00    8307
29 2005-01-17 10:07:00     720
30 1998-11-22 06:49:00    8384
31 2005-09-09 10:08:00     808
32 2004-07-16 14:01:00     649
33 2003-05-28 00:39:00     365
34 2006-12-13 02:57:00     930
35 2001-12-28 21:32:00    9767
36 2013-11-05 22:15:00    1890
37 2002-07-20 21:54:00      39
38 1998-11-28 06:13:00    8395
39 2013-05-14 01:20:00    1748
40 2014-10-24 22:13:00    2192
41 2002-08-24 01:31:00      69
42 2002-07-15 20:14:00      30
43 2013-05-13 16:16:00    1748
44 2001-12-11 08:14:00    9733
45 1998-08-18 08:32:00    8307
46 2015-05-05 22:15:00    2339
47 2003-11-03 01:45:00     488
48 1998-05-06 08:20:00    8210
49 2005-01-15 23:31:00     720
50 2001-09-24 11:09:00    9632
```

[1373]:
```
#Step 3
```

[1374]:
```
res2 = requests.get("https://cmsc320.github.io/files/waves_type2.html")
```

[1375]:
```
root = BeautifulSoup(res2.content)
```

[1376]:
```
root = root.text
root = root.split("\n")
```

[1377]:
```
arr_2d = []
for x in range(15,533):
    arr_2d.append(root[x].split(" "))
#separate each element
```

[1378]:
```
arr_2d = [[i for i in item if i != ''] for item in arr_2d]
#remove empty in arr_2d
```

[1379]:
```
df3 = pd.DataFrame(arr_2d)
#build dataframe
```

[1380]:
```
for i in range (15,24):
    df3 = df3.drop(i,axis = 1)
#drop unused columns
```

[1381]:
```
df3.columns = ['start_date <chr>', 'start_time <chr>', 'end_date <chr>',␣
 ↪'end_time <chr>', 'start_frequency <chr>', 'end_frequency <chr>',␣
 ↪'flare_location <chr>', 'flare_region <chr>',
              'flare_classification <chr>', 'cme_date <chr>', 'cme_time <chr>',␣
 ↪'cme_angle <chr>', 'cme_width <chr>', 'cme_speed <chr>','plot <chr>']
#rename cloumns
```

[1382]:
```
df3
```

[1382]:

	start_date <chr>	start_time <chr>	end_date <chr>	end_time <chr>	\
0	1997/04/01	14:00	04/01	14:15	
1	1997/04/07	14:30	04/07	17:30	
2	1997/05/12	05:15	05/14	16:00	
3	1997/05/21	20:20	05/21	22:00	
4	1997/09/23	21:53	09/23	22:16	
..	…	…	…	…	
513	2017/09/04	20:27	09/05	04:54	
514	2017/09/06	12:05	09/07	08:00	
515	2017/09/10	16:02	09/11	06:50	
516	2017/09/12	07:38	09/12	07:43	
517	2017/09/17	11:45	09/17	12:35	

	start_frequency <chr>	end_frequency <chr>	flare_location <chr>	\
0	8000	4000	S25E16	
1	11000	1000	S28E19	
2	12000	80	N21W08	
3	5000	500	N05W12	
4	6000	2000	S29E25	
..	…	…	…	

```
513                14000                210              S10W12
514                16000                 70              S08W33
515                16000                150              S09W92
516                16000              13000              N08E48
517                16000                900             S08E170

    flare_region <chr> flare_classification <chr> cme_date <chr>  \
0                 8026                       M1.3          04/01
1                 8027                       C6.8          04/07
2                 8038                       C1.3          05/12
3                 8040                       M1.3          05/21
4                 8088                       C1.4          09/23
..                 …                          …             …
513              12673                       M5.5          09/04
514              12673                       X9.3          09/06
515              -----                       X8.3          09/10
516              12680                       C3.0          09/12
517              -----                       ----          09/17

    cme_time <chr> cme_angle <chr> cme_width <chr> cme_speed <chr> plot <chr>
0            15:18              74              79              312       PHTX
1            14:27            Halo             360              878       PHTX
2            05:30            Halo             360              464       PHTX
3            21:00             263             165              296       PHTX
4            22:02             133             155              712       PHTX
..             …               …               …               …         …
513          20:12            Halo             360             1418       PHTX
514          12:24            Halo             360             1571       PHTX
515          16:00            Halo             360             3163       PHTX
516          08:03             124              96              252       PHTX
517          12:00            Halo             360             1385       PHTX

[518 rows x 15 columns]
```

[1383]:
```
#Step 4
```

[1384]:
```
df4 = df3
```

[1385]:
```
for i in range(0,518):
    for j in range(0,14):
        if df4.iat[i,j][0] == "-":
            df4.iat[i,j] = "NaN"
#Using for lopp to replace not avaible values by NaN
```

[1386]:
```
for i in range(0,518):
    df4.iat[i,2] = df4.iat[i,0][0]+df4.iat[i,0][1]+df4.iat[i,0][2] + df4.
 ↪iat[i,0][3]+df4.iat[0,0][4]+ df4.iat[i,2]
```

```
    if df4.iat[i,9] != "NaN":
        df4.iat[i,9] = df4.iat[i,0][0]+df4.iat[i,0][1]+df4.iat[i,0][2] + df4.
 ↪iat[i,0][3]+df4.iat[0,0][4]+ df4.iat[i,9]
# Add year to end_date and cme_date then date has the form YYYY/MM/DD
# If cme_date is NaN, we skip
```

[1387]:
```
for i in range(0,518):
    df4.iat[i,1] = df4.iat[i,1]+":00"
    df4.iat[i,3] = df4.iat[i,3]+":00"
    if df4.iat[i,9] != "NaN":
        df4.iat[i,10] = df4.iat[i,10]+":00"
# Add second to start_time, end_time and cme_time then date has the form HH/MM/
 ↪SS
# If cme_date is NaN, we skip
```

[1388]:
```
df4['start_datetime'] = pd.to_datetime(df4['start_date <chr>']) + pd.
 ↪to_timedelta(df4['start_time <chr>'])
df4['end_datetime'] = pd.to_datetime(df4['end_date <chr>']) + pd.
 ↪to_timedelta(df4['end_time <chr>'])
df4['cme_datetime'] = pd.to_datetime(df4['cme_date <chr>']) + pd.
 ↪to_timedelta(df4['cme_time <chr>'])
# Create three datetime columns using to_datetime
```

[1389]:
```
df4.columns = ['start_date', 'start_time', 'end_date', 'end_time',␣
 ↪'start_frequency', 'end_frequency', 'flare_location', 'flare_region',
              'flare_classification', 'cme_date', 'cme_time', 'cme_angle',␣
 ↪'cme_width',␣
 ↪'cme_speed','plot','start_datetime','end_datetime','cme_datetime']
#rename columns
```

[1390]:
```
#lambda expression !!!!!
list1 = []
for i in range(0,518):
    if df4.iat[i,11] == "Halo":
        df4.iat[i,11] = "NA"
        list1.append(True)
    else :
        list1.append(False)
df4["is_halo"] = list1
# Create "is_halo" column by iterating all elements.
# Also repace all "Halo" to "NA"
```

[1391]:
```
list2 = []
for i in range(0,518):
    if df4.iat[i,12][0] == ">":
        df4.iat[i,12] = df4.iat[i,12][1:]
```

```
        list2.append(True)
    else :
        list2.append(False)
df4["width_lower_bound"] = list2
# Create "width_lower_bound" column by iterating all elements and checking if␣
 ↪there is a ">".
# Also delete ">" if there is one
```

[1392]:
```
df4 = df4.drop(['start_time'],axis = 1)
df4 = df4.drop(['start_date'],axis = 1)
df4 = df4.drop(['end_time'],axis = 1)
df4 = df4.drop(['end_date'],axis = 1)
df4 = df4.drop(['cme_time'],axis = 1)
df4 = df4.drop(['cme_date'],axis = 1)
# Drop columns to better fit with sample output
```

[1393]:
```
df4 = df4[['start_datetime','end_datetime','start_frequency', 'end_frequency',␣
 ↪'flare_location', 'flare_region',
            'flare_classification', 'cme_datetime', 'cme_angle', 'cme_width',␣
 ↪'cme_speed','plot','is_halo','width_lower_bound']]
# Change columns olders to better fit with sample output
```

[1394]:
```
df4
```

[1394]:

	start_datetime	end_datetime	start_frequency	end_frequency	\
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	
..	…	…	…	…	
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	
517	2017-09-17 11:45:00	2017-09-17 12:35:00	16000	900	

	flare_location	flare_region	flare_classification	cme_datetime	\
0	S25E16	8026	M1.3	1997-04-01 15:18:00	
1	S28E19	8027	C6.8	1997-04-07 14:27:00	
2	N21W08	8038	C1.3	1997-05-12 05:30:00	
3	N05W12	8040	M1.3	1997-05-21 21:00:00	
4	S29E25	8088	C1.4	1997-09-23 22:02:00	
..	…	…	…	…	
513	S10W12	12673	M5.5	2017-09-04 20:12:00	
514	S08W33	12673	X9.3	2017-09-06 12:24:00	
515	S09W92	NaN	X8.3	2017-09-10 16:00:00	

```
516         N08E48        12680                 C3.0 2017-09-12 08:03:00
517        S08E170          NaN                  NaN 2017-09-17 12:00:00

    cme_angle cme_width cme_speed  plot  is_halo  width_lower_bound
0          74        79       312  PHTX    False              False
1          NA       360       878  PHTX     True              False
2          NA       360       464  PHTX     True              False
3         263       165       296  PHTX    False              False
4         133       155       712  PHTX    False              False
..        …       …       …   …      …                 …
513        NA       360      1418  PHTX     True              False
514        NA       360      1571  PHTX     True              False
515        NA       360      3163  PHTX     True              False
516       124        96       252  PHTX    False              False
517        NA       360      1385  PHTX     True              False

[518 rows x 14 columns]
```

[1395]:
```
#Part 2 Analysis
```

[1396]:
```
#Q1
```

[1397]:
```
list3 = []
#for this sorting, we convert flare_classification to weighted numbers.
#As researched, Classification has a letter which is A B C M X(increasing␣
 ↪order) and a number.
for i in range(0,518):
    if df4.iat[i,6] != "NaN" and df4.iat[i,6] != "FILA" and df4.iat[i,6] !=␣
 ↪"DSF":
        folat_str = df4.iat[i,6][1:]
        list3.append(1000*ord(df4.iat[i,6][0])+float(folat_str))
        #Thus, we can give the letter a weight of 1000 and the number weight of␣
 ↪1.
        #(Since the number is much less likely to influence the rank compared␣
 ↪to the letter.)
    else:
        list3.append(0)
        #If there is no data, we assign 0.
df5 = df4
df5["weighted_flare_rank"] = list3
#We added our new column to the table.
```

[1398]:
```
df5 = df5.drop(['start_frequency'], axis = 1)
df5 = df5.drop(['end_frequency'], axis = 1)
df5 = df5.drop(['flare_location'], axis = 1)
df5 = df5.drop(['cme_angle'], axis = 1)
df5 = df5.drop(['cme_width'], axis = 1)
```

```
df5 = df5.drop(['cme_speed'], axis = 1)
df5 = df5.drop(['plot'], axis = 1)
df5 = df5.drop(['is_halo'], axis = 1)
df5 = df5.drop(['width_lower_bound'], axis = 1)
df5 = df5.drop(['cme_datetime'], axis = 1)
#we delele unshown columns compared to SWL
```

[1399]:
```
df5.sort_values(by = 'weighted_flare_rank',ascending=False).head(50)
#We sort by the weighted_flare_rank in decending order and only pick top 50.
```

[1399]:

	start_datetime	end_datetime	flare_region	flare_classification	\
240	2003-11-04 20:00:00	2003-11-05 00:00:00	10486	X28.	
117	2001-04-02 22:05:00	2001-04-03 02:30:00	9393	X20.	
233	2003-10-28 11:10:00	2003-10-30 00:00:00	10486	X17.	
126	2001-04-15 14:05:00	2001-04-16 13:00:00	9415	X14.	
234	2003-10-29 20:55:00	2003-10-30 00:00:00	10486	X10.	
8	1997-11-06 12:20:00	1997-11-07 08:30:00	8100	X9.4	
514	2017-09-06 12:05:00	2017-09-07 08:00:00	12673	X9.3	
328	2006-12-05 10:50:00	2006-12-05 20:00:00	10930	X9.0	
515	2017-09-10 16:02:00	2017-09-11 06:50:00	NaN	X8.3	
237	2003-11-02 17:30:00	2003-11-03 01:00:00	10486	X8.3	
288	2005-01-20 07:15:00	2005-01-20 16:30:00	10720	X7.1	
359	2011-08-09 08:20:00	2011-08-09 08:35:00	11263	X6.9	
331	2006-12-06 19:00:00	2006-12-09 00:00:00	10930	X6.5	
317	2005-09-09 19:45:00	2005-09-09 22:00:00	10808	X6.2	
82	2000-07-14 10:30:00	2000-07-15 14:30:00	9077	X5.7	
121	2001-04-06 19:35:00	2001-04-07 01:50:00	9415	X5.6	
375	2012-03-07 01:00:00	2012-03-08 19:00:00	11429	X5.4	
135	2001-08-25 16:50:00	2001-08-25 23:00:00	9591	X5.3	
443	2014-02-25 00:56:00	2014-02-25 11:28:00	11990	X4.9	
193	2002-07-23 00:50:00	2002-07-23 04:00:00	10039	X4.8	
104	2000-11-26 17:00:00	2000-11-26 17:15:00	9236	X4.0	
239	2003-11-03 10:00:00	2003-11-03 12:30:00	10488	X3.9	
286	2005-01-17 10:00:00	2005-01-17 10:35:00	10720	X3.8	
222	2003-05-28 01:00:00	2003-05-29 00:30:00	10365	X3.6	
332	2006-12-13 02:45:00	2006-12-13 10:40:00	10930	X3.4	
160	2001-12-28 20:35:00	2001-12-29 03:00:00	9756	X3.4	
192	2002-07-20 21:30:00	2002-07-20 22:20:00	10039	X3.3	
404	2013-05-14 01:16:00	2013-05-14 08:20:00	11748	X3.2	
201	2002-08-24 01:45:00	2002-08-24 03:25:00	10069	X3.1	
403	2013-05-13 16:15:00	2013-05-13 19:10:00	11748	X2.8	
487	2015-05-05 22:24:00	2015-05-05 23:14:00	12339	X2.7	
19	1998-05-06 08:25:00	1998-05-06 08:35:00	8210	X2.7	
238	2003-11-03 01:15:00	2003-11-03 01:25:00	10488	X2.7	
284	2005-01-15 23:00:00	2005-01-17 00:00:00	10720	X2.6	
142	2001-09-24 10:45:00	2001-09-25 20:00:00	9632	X2.6	
9	1997-11-27 13:30:00	1997-11-27 14:00:00	8113	X2.6	

```
276 2004-11-10 02:25:00 2004-11-10 03:40:00        10696                X2.5
73  2000-06-06 15:20:00 2000-06-08 09:00:00         9026                X2.3
123 2001-04-10 05:24:00 2001-04-11 00:00:00         9415                X2.3
99  2000-11-24 15:25:00 2000-11-24 22:00:00         9236                X2.3
345 2011-02-15 02:10:00 2011-02-15 07:00:00        11158                X2.2
318 2005-09-10 21:45:00 2005-09-11 01:00:00        10808                X2.1
361 2011-09-06 22:30:00 2011-09-07 15:40:00        11283                X2.1
420 2013-10-25 15:08:00 2013-10-25 22:32:00        11882                X2.1
7   1997-11-04 06:00:00 1997-11-05 04:30:00         8100                X2.1
98  2000-11-24 05:10:00 2000-11-24 15:00:00         9236                X2.0
125 2001-04-12 10:20:00 2001-04-12 10:40:00         9415                X2.0
274 2004-11-07 16:25:00 2004-11-08 20:00:00        10696                X2.0
285 2005-01-17 09:25:00 2005-01-17 16:00:00        10720                X2.0
102 2000-11-25 19:00:00 2000-11-25 19:35:00         9236                X1.9
```

```
     weighted_flare_rank
240              88028.0
117              88020.0
233              88017.0
126              88014.0
234              88010.0
8                88009.4
514              88009.3
328              88009.0
515              88008.3
237              88008.3
288              88007.1
359              88006.9
331              88006.5
317              88006.2
82               88005.7
121              88005.6
375              88005.4
135              88005.3
443              88004.9
193              88004.8
104              88004.0
239              88003.9
286              88003.8
222              88003.6
332              88003.4
160              88003.4
192              88003.3
404              88003.2
201              88003.1
403              88002.8
487              88002.7
```

```
19              88002.7
238             88002.7
284             88002.6
142             88002.6
9               88002.6
276             88002.5
73              88002.3
123             88002.3
99              88002.3
345             88002.2
318             88002.1
361             88002.1
420             88002.1
7               88002.1
98              88002.0
125             88002.0
274             88002.0
285             88002.0
102             88001.9
```

[1400]:
```
#We cannot replicate the data exactly. There are some missing solar flare␣
 ↪events.
#The data we get from NASA is not identical to SpaceWeatherLive but most rows␣
 ↪can
#replicate the SpaceWeatherLive by indetical flare_region and␣
 ↪flare_classification(times may differ).
#To be noticed, some regions have a leading "1" and some flare_classifications␣
 ↪have "+" at the end.
#We consider it is same when other parts are identical.
```

[1401]:
```
#Q2
```

[1402]:
```
df4_NASA = df5
df5_SWL = df2
```

[1403]:
```
#To find best matches, firstly compare the x-classification, region.
#If those are matched, we consider it is a match.
#If many match to the same row, we then compare the difference of their date␣
 ↪and time. We picked the smallest one.
#
```

[1404]:
```
rank = [-1]*518
time_difference = [-1]*518
#rank is the ranks in SWL for NASA values

#time_difference is the sum of absolute value of
```

```
#(SWl_start_datetime - NASA_start_datetime) and
# absolute value of (SWl_end_datetime - NASA_end_datetime)

#We firstly initialize ranks and time_difference to be -1, which means not␣
 ↪avaible.

for i in range (0,50):
    if df5_SWL.iat[i,1][-1] == "+":
        SWL_x_class = df5_SWL.iat[i,1][:-1]
    else:
        SWL_x_class = df5_SWL.iat[i,1]
    #Since some X_class has a "+", we will remove that for comparision.

    SWL_region = df5_SWL.iat[i,5]
    SWl_start_datetime = df5_SWL.iat[i,2]
    SWl_end_datetime = df5_SWL.iat[i,4]
   #We get region,start and end for SWL data.

    for j in range(0,518):#(0,518)
        if df4_NASA.iat[j,3][-1] == ".":
            NASA_x_class = df4_NASA.iat[j,3][:-1]
        else:
            NASA_x_class = df4_NASA.iat[j,3]
            #Since some X_class has a ".", we will remove that for comparision.

        NASA_region = df4_NASA.iat[j,2]
        if(len(str(NASA_region)) == 5):
            NASA_region = int(str(NASA_region)[1:])
        #Since some region has a "1" at beginning, we will remove that for␣
 ↪comparision.

        NASA_start_datetime = df4_NASA.iat[j,0]
        NASA_end_datetime = df4_NASA.iat[j,1]
        time_difference_new = abs(SWl_start_datetime - NASA_start_datetime) + \
        abs(SWl_end_datetime - NASA_end_datetime)

        if SWL_x_class == NASA_x_class and SWL_region == NASA_region :
        #If x_class and region matches, we think it is a match.
            if rank[j] != -1:
                #If the value already has a match, we compare by their␣
 ↪time_difference.
                if time_difference_new < time_difference[j]:
                    #If new value has smaller time_difference, we update our␣
 ↪rank and time_difference
                    rank[j] = i+1
                    time_difference[j] = time_difference_new
            else :
```

```
                #If the value doesn't have a match, we update our rank and␣
 ↪time_difference
                rank[j] = i+1
                time_difference[j] = time_difference_new
df4["rank"] = rank
#We add rank to the dataframe.
#df4["time_dif"] = time_difference
```

[1405]:
```
df4 = df4.replace(-1,"")
df4
```

[1405]:

	start_datetime	end_datetime	start_frequency	end_frequency
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
..	…	…	…	…
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
517	2017-09-17 11:45:00	2017-09-17 12:35:00	16000	900

	flare_location	flare_region	flare_classification	cme_datetime
0	S25E16	8026	M1.3	1997-04-01 15:18:00
1	S28E19	8027	C6.8	1997-04-07 14:27:00
2	N21W08	8038	C1.3	1997-05-12 05:30:00
3	N05W12	8040	M1.3	1997-05-21 21:00:00
4	S29E25	8088	C1.4	1997-09-23 22:02:00
..	…	…	…	…
513	S10W12	12673	M5.5	2017-09-04 20:12:00
514	S08W33	12673	X9.3	2017-09-06 12:24:00
515	S09W92	NaN	X8.3	2017-09-10 16:00:00
516	N08E48	12680	C3.0	2017-09-12 08:03:00
517	S08E170	NaN	NaN	2017-09-17 12:00:00

	cme_angle	cme_width	cme_speed	plot	is_halo	width_lower_bound
0	74	79	312	PHTX	False	False
1	NA	360	878	PHTX	True	False
2	NA	360	464	PHTX	True	False
3	263	165	296	PHTX	False	False
4	133	155	712	PHTX	False	False
..	…	…	…	…	…	…
513	NA	360	1418	PHTX	True	False
514	NA	360	1571	PHTX	True	False
515	NA	360	3163	PHTX	True	False

```
516       124        96       252  PHTX    False              False
517        NA       360      1385  PHTX     True              False

     weighted_flare_rank rank
0                77001.3
1                67006.8
2                67001.3
3                77001.3
4                67001.4
..                   …  …
513              77005.5
514              88009.3    8
515              88008.3
516              67003.0
517                  0.0

[518 rows x 16 columns]
```

[1406]:
```
# there is no duplicates SWL entries "best matches".
```

[1407]:
```
#Q3
#Intention: To check if there is a relationship between higher X_classification␣
 ↪and Halo CME.
#The variation is the x_classification, which means some of the strongest solar␣
 ↪flares with others.
# We compare the height difference (or proportion) of Halo CMEs in the top 50␣
 ↪flares vs. the dataset as a whole.
```

[1408]:
```
import matplotlib.pyplot as plt
```

[1409]:
```
count_top50 = 0
count_top50_halo = 0
for i in range(0,518):
    if df4.iat[i,15]!= "" :
        count_top50 += 1
        if df4.iat[i,12] == True:
            count_top50_halo +=1
print(count_top50)
print(count_top50_halo)
```

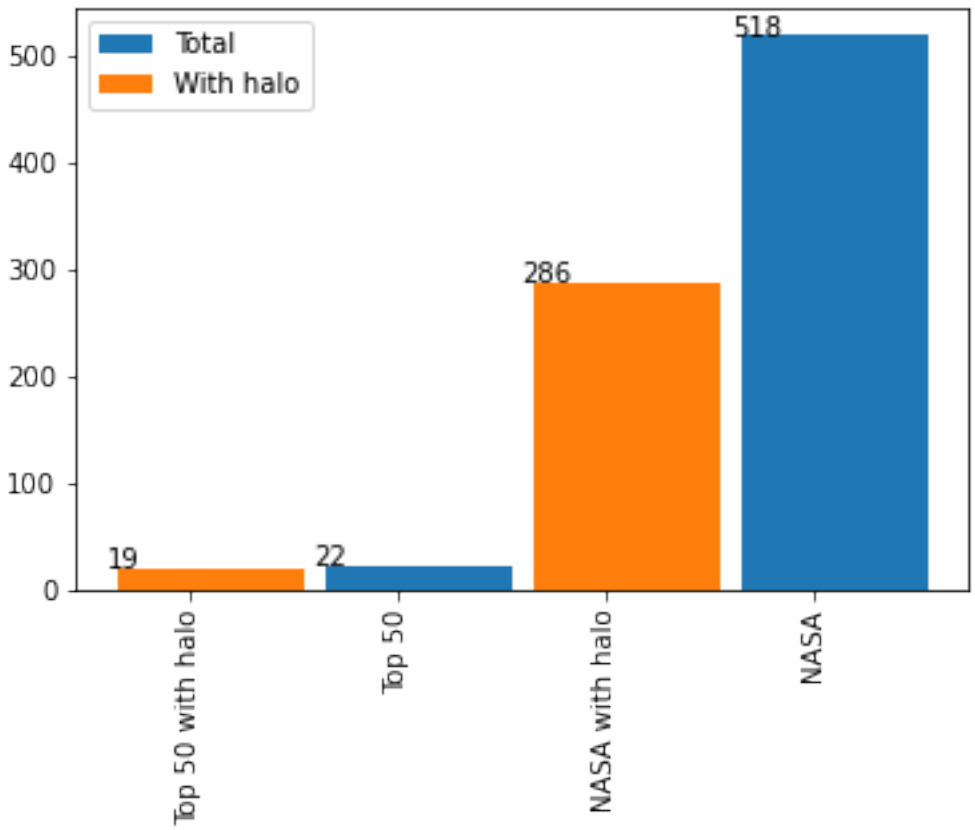
```
22
19
```

[1410]:
```
count_all = 0
count_all_halo = 0
for i in range(0,518):
    count_all += 1
```

```
    if df4.iat[i,12] == True:
        count_all_halo +=1
print(count_all)
print(count_all_halo)
```

```
518
286
```

[1411]:
```
barWidth = 0.9
bars1 = [22, 518]
bars2 = [19, 286]
bars4 = bars1+bars2

r1 = [2,4]
r2 = [1,3]
r4 = r1+r2

label = ['22', '518','19', '286']

plt.bar(r1, bars1, width = barWidth, label = 'Total')
plt.bar(r2, bars2, width = barWidth, label = 'With halo')

plt.legend()

plt.xticks([r + barWidth for r in range(len(r4))], ['Top 50 with halo','Top␣
 ↪50', 'NASA with halo', 'NASA'], rotation=90)

for i in range(len(r4)):
    plt.text(x = r4[i]-0.5 , y = bars4[i]+0.1, s = label[i], size = 10)

plt.show()
```



[1412]:
```
#Description of Plot:
# The y-axis is the number of solar flares and x-axis has four types:
# Top 50 with halo','Top 50', 'NASA with halo', 'NASA'.
# Organe color means with Halo and blue means without Halo.
# The height represents the number.
```

[1413]:
```
# Yes, flares in the top 50 tend to have Halo CMEs. Based on the graph,
# the difference of height between top50 is much less than all data.
# Also, 19/22 = 0.86... and 286/518 = 0.552...
# Thus, flares in the top 50 tend to have Halo CMEs.
# Thus, we can conclude that as solar flares are stronger, there is more␣
 ↪possibility that it is with Halo CME.
```

[ ]:

[ ]: