

# project1

October 2, 2022

## Part 1

```
[1]: import pandas as pd
import numpy as np
from bs4 import BeautifulSoup
import requests
```

```
[2]: # Step 1
# Get raw data from website
r = requests.get("https://cmssc320.github.io/files/top-50-solar-flares.html")
# Parse data to html
soup = BeautifulSoup(r.text, 'html.parser')
# Prettify the table to find the table
soup.prettify()
# print(soup.prettify)
# Find the desired table froup the html
table = soup.find("table", {"class": "table table-striped table-responsive-md"})
# Convert table html to dataframe
df = pd.read_html(str(table))[0]
# Rename dataframe columns
df = df.rename(columns={"Unnamed: 0": "Rank", "Unnamed: 1": "x_classification",
↪ "Unnamed: 2": "Date", "Start": "Start_Time", "Maximum": "Maximum_Time", "End":
↪ "End_Time", "Unnamed: 7": "Movie"})

print(df.head())
```

	Rank	x_classification	Date	Region	Start_Time	Maximum_Time	End_Time	\
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	

	Movie
0	MovieView archive
1	MovieView archive
2	MovieView archive
3	MovieView archive

#### 4 MovieView archive

```
[3]: # Step 2
# Drop the final column
# Note that rerunning this cell will error, trying to delete nonexistent row
df = df.drop("Movie", axis=1)
# Combine dates and times in time columns,
for row in range(len(df)):
    df.at[row, "Start_Time"] = pd.to_datetime(df.at[row, "Start_Time"] + " " +
    ↪df.at[row, "Date"])
    df.at[row, "Maximum_Time"] = pd.to_datetime(df.at[row, "Maximum_Time"] + " "
    ↪+ df.at[row, "Date"])
    df.at[row, "End_Time"] = pd.to_datetime(df.at[row, "End_Time"] + " " + df.
    ↪at[row, "Date"])
# then drop date column,
df = df.drop("Date", axis=1)
# rename columns,
df = df.rename(columns={"Start_Time": "Start_Datetime", "Maximum_Time":
    ↪"Max_Datetime", "End_Time": "End_Datetime"})
# and move Region to the end
df = df[["Rank", "x_classification", "Start_Datetime", "Max_Datetime",
    ↪"End_Datetime", "Region"]]
# Replace missing Region values with NAN
df = df.replace("-", np.nan)

print(df.head())
```

	Rank	x_classification	Start_Datetime	Max_Datetime \
0	1	X28+	2003-11-04 19:29:00	2003-11-04 19:53:00
1	2	X20+	2001-04-02 21:32:00	2001-04-02 21:51:00
2	3	X17.2+	2003-10-28 09:51:00	2003-10-28 11:10:00
3	4	X17+	2005-09-07 17:17:00	2005-09-07 17:40:00
4	5	X14.4	2001-04-15 13:19:00	2001-04-15 13:50:00

	End_Datetime	Region
0	2003-11-04 20:06:00	486
1	2001-04-02 22:03:00	9393
2	2003-10-28 11:24:00	486
3	2005-09-07 18:03:00	808
4	2001-04-15 13:55:00	9415

```
[4]: # Step 3
# Get raw data from website
r2 = requests.get("https://cmssc320.github.io/files/waves_type2.html")
# Parse data to html
soup2 = BeautifulSoup(r2.text, 'html.parser')
# Prettify html and notice it's stored as strings separated by newlines
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text2 = soup2.prettify()
# So, split lines for each newline
lst = text2.split("\n")
# Then, delete every line that doesn't start with a number,
# as only table rows start with a number
lst = [x for x in lst if len(x) > 0 and x[0].isnumeric()]

# Long process to convert each row to data entries
# Create list to append values to
biglst = []
# Loop through each line
for line in lst:
    # Split each line across spaces
    temp = line.split(" ")
    # Initialize loop variable; index for ^^ split
    y=0
    # To summarize,
    # in each row, split into strings whenever a space appears,
    # and then loop across those strings
    while y < len(temp):
        # If current string is empty (when table has two consecutive spaces),
        ↪ delete
        if len(temp[y]) == 0:
            del temp[y]
            y = y - 1
        # Many of our row elements are of the form:
        # <a href="some url">VALUE</a>
        # We want to extract VALUE
        # However, when splitting with spaces, "<a href" -> "<a", "href"
        # So, delete the first of those,
        elif temp[y] == "<a":
            del temp[y]
            y = y - 1
        # before parsing VALUE out of "href="some url">VALUE</a>
        elif "href" in temp[y]:
            tempstr = temp[y]
            templst = tempstr.split(">")
            temp[y] = templst[1][:-3]
            y = y + 1
        # Delete empty columns and add to list
        del temp[14:]
        biglst.append(temp)

# Finally, make a dataframe with appropriate column labels
df2 = pd.DataFrame(biglst, columns=["Start_Date", "Start_Time", "End_Date",
                                   "End_Time", "Start_Frequency",
                                   ↪ "End_Frequency",

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        "Flare_Location", "Flare_Region", \
        ↪ "Flare_Classification",
        "CME_Date", "CME_Time", "CME_Angle", \
        ↪ "CME_Width", "CME_Speed"])

print(df2.head())

```

	Start_Date	Start_Time	End_Date	End_Time	Start_Frequency	End_Frequency	\
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	

	Flare_Location	Flare_Region	Flare_Classification	CME_Date	CME_Time	\
0	S25E16	8026	M1.3	04/01	15:18	
1	S28E19	8027	C6.8	04/07	14:27	
2	N21W08	8038	C1.3	05/12	05:30	
3	N05W12	8040	M1.3	05/21	21:00	
4	S29E25	8088	C1.4	09/23	22:02	

	CME_Angle	CME_Width	CME_Speed
0	74	79	312
1	Halo	360	878
2	Halo	360	464
3	263	165	296
4	133	155	712

```

[5]: # Step 4

# Note that for efficiency, all of these for loops could be condensed into one.
# However, to preserve the structure of the assignment, they are separate.

# Find all missing entry types, then recode as NaN
# Loop through each row
for row in range(len(df2)):
    # Then loop through list of columns
    for col in df2.columns:
        # All missing entries start with a hyphen
        if df2.at[row, col][0] == "-":
            df2.at[row, col] = np.nan

# Create column of booleans if cme_angle is halo or not; change halo values in
↪ cme_angle to NaN
# Initialize list of booleans for halo or not
halo = []
# For each row, if CME angle is halo, append true and change to nan, else false

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for row in range(len(df2)):
    if df2.at[row, "CME_Angle"] == "Halo":
        halo.append(True)
        df2.at[row, "CME_Angle"] = np.nan
    else:
        halo.append(False)
df2["Is_Halo"] = halo

# cme_width indicates if value is lower bound (int or >int);
# create column of bools for lower bound or not and make cme_width int only
# The same concept and code as for halo

lower = []
for row in range(len(df2)):
    # Assignment to avoid duplicate .at calls
    curr = df2.at[row, "CME_Width"]
    # Needed case since NaN, as missing entry replacement, is type float, not
    ↪ string
    if type(curr) == float:
        lower.append(False)
    elif curr[0] == ">":
        lower.append(True)
        # Slice used to update value as all but first character, which is >
        df2.at[row, "CME_Width"] = curr[1:]
    else:
        lower.append(False)
df2["Width_Lower_Bound"] = lower

# Combine date/time columns for start, end, and cme
# Prob gonna need to take year off of start date for end date and cme date;
↪ visit each row once
# Combine dates and times in time columns:
for row in range(len(df2)):
    # Repeating block to convert 24:00 to 00:00 for to_datetime to parse
    time = df2.at[row, "Start_Time"]
    # If time is missing entry (nan is a float), keep nan
    if type(time) == float:
        df2.at[row, "Start_Time"] = np.nan
    else:
        if time[:2] == "24":
            time = "00" + time[2:]
        df2.at[row, "Start_Time"] = pd.to_datetime(time + " " + df2.at[row,
↪ "Start_Date"])

# Save year since it only appears in start time
year = df2.at[row, "Start_Date"][:4]

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time = df2.at[row, "CME_Time"]
if type(time) == float:
    df2.at[row, "Start_Time"] = np.nan
else:
    if time[:2] == "24":
        time = "00" + time[2:]
    df2.at[row, "CME_Time"] = pd.to_datetime(time + " " + year + "/" + df2.
    at[row, "CME_Date"])

time = df2.at[row, "End_Time"]
if type(time) == float:
    df2.at[row, "Start_Time"] = np.nan
else:
    if time[:2] == "24":
        time = "00" + time[2:]
    df2.at[row, "End_Time"] = pd.to_datetime(time + " " + year + "/" + df2.
    at[row, "End_Date"])

# then drop date columns,
df2 = df2.drop("Start_Date", axis=1).drop("End_Date", axis=1).drop("CME_Date",
    axis=1)
# and rename columns
df2 = df2.rename(columns={"Start_Time": "Start_Datetime",
    "CME_Time": "CME_Datetime", "End_Time": "End_Datetime"})

print(df2.head())

```

	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	

	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	

	CME_Angle	CME_Width	CME_Speed	Is_Halo	Width_Lower_Bound
0	74	79	312	False	False
1	NaN	360	878	True	False
2	NaN	360	464	True	False
3	263	165	296	False	False
4	133	155	712	False	False

## Part 2

```
[6]: # Question 1
# Is it possible to replicate the top 50 solar flares from NASA data? (Rank
↳based on x classification)
# Make a list of all classifications
classes = df2.loc[:, "Flare_Classification"]
# Initialize and populate a list of just the x classifications (highest rank)
onlyx = []
for c in range(len(classes)):
    if type(classes[c]) == str and classes[c][0] == "X":
        onlyx.append((c, classes[c]))
# Sort the list based on the floats after the X
sort = sorted(onlyx, key = lambda x : float(x[1][1:]), reverse=True)
# Take the top 50 of those floats
top50 = sort[:50]
# And find the associated indices
top50i = [i for (i, c) in top50]
# Then, make a dataframe with the same columns as the NASA dataframe
df50 = pd.DataFrame(columns=df2.columns)
# And add the rows of the top 50 flares to that dataframe
j = 0
for i in top50i:
    df50.loc[j] = df2.iloc[i]
    j = j + 1

# print(df["Start_Datetime"])
# print(df50["Start_Datetime"])

# By simply comparing the lists of dates from the original 50 and the NASA 50,
# we see within the first ten entries multiple discrepancies.
# For one, NASA is missing a few, and has some NaN that the original doesn't
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[7]: # Question 2
# For each top 50 flare, find best matching row in NASA
# Add column to NASA indicating rank, if it appears in that dataset

# We use dates as our first point of comparison
# Regardless of differences between measurements or classifications,
# dates are absolute enough that the only errors that may appear are near ends
↳of days,
# which is a small margin, whereas classification, for example, could be off by
# a few tenths in a harder way to predict.
# Then, if a date has two or no flares in the NASA dataset, we pick the higher
↳classification
# or skip that rank, respectively
```

```

# We create a list of all the start dates in the NASA dataset
start_dates = [date.date() if type(date) is not float else np.nan for date in df2["Start_Datetime"]]
matchlist = []
# Then, we loop through the top 50 dataset
for row in range(len(df)):
    date = df.loc[row, "Start_Datetime"].date()
    # Find all matching dates
    matches = np.where(np.array(start_dates) == date)[0]
    # And select our best match
    # Arbitrary to select if only one match
    if len(matches) == 1:
        matchlist.append(matches[0])
    # If two matches, pick the larger classification, preferring x as a prefix
    elif len(matches) > 1:
        maxclass = 0
        maxidx = 0
        for index in matches:
            cla = df2.loc[index, "Flare_Classification"]
            if type(cla) is not float:
                if cla[0] == "X":
                    old = maxclass
                    maxclass = max(maxclass, float(cla[1:]))
                    if maxclass is not old:
                        maxidx = index
        matchlist.append(maxidx)
    # If no matches, add a placeholder to the list to skip this rank
    else:
        matchlist.append(np.nan)
# Create a new column with -1 as unranked for type consistency
df2["Rank"] = [-1] * len(df2)
rank = 1
# Loop through matches, assigning appropriate rank to where matches appear
for i in matchlist:
    if i is np.nan:
        rank = rank + 1
    else:
        df2.at[i, "Rank"] = rank
        rank = rank + 1
# Tail is shown as result because it has ranked entries
print(df2.tail())

```

	Start_Datetime	End_Datetime	Start_Frequency	End_Frequency	\
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	\
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	Is_Halo	Width_Lower_Bound	Rank
513	NaN	360	1418	True	False	-1
514	NaN	360	1571	True	False	8
515	NaN	360	3163	True	False	11
516	124	96	252	False	False	-1
517	NaN	360	1385	True	False	-1

```
[8]: # Question 3
# Prepare a plot showing the top 50 in context with all NASA data

from matplotlib import pyplot as plt

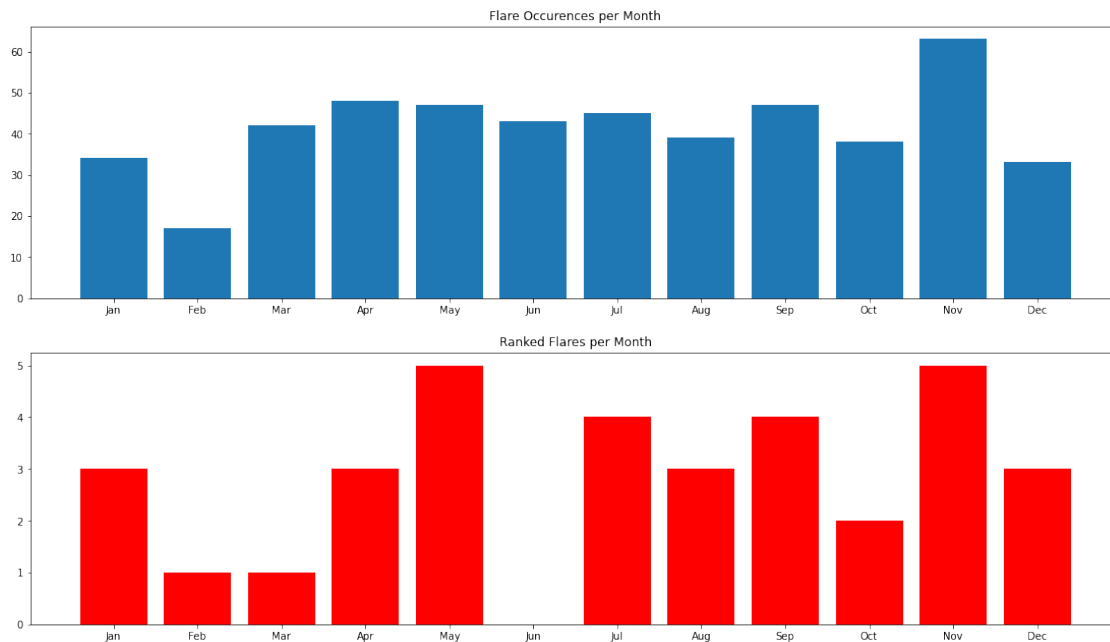
# One interesting question about a lot of enviromental behaviour on Earth is
# when it happens
# With seasons and human-induced climate change, we care about when certain
# things happen
# While these factors don't necessarily exist on the sun, the question remains:
# Does time have a noticeable impact on strong flare occurences?

months = [0] * 12
ranked = [0] * 12
for row in range(len(df2)):
    month = df2.at[row, "Start_Datetime"]
    if month is not np.nan:
        month = month.date().month
        months[month - 1] += 1
        if df2.at[row, "Rank"] > 0:
            ranked[month - 1] += 1

fig, ax = plt.subplots(2, figsize=(19.2,10.8))
ax[0].bar(height=months,
    x=["Jan", "Feb", "Mar", "Apr", "May", "Jun", "Jul", "Aug", "Sep", "Oct", "Nov", "Dec"])
ax[0].set_title("Flare Occurences per Month")
ax[1].bar(height=ranked,
    x=["Jan", "Feb", "Mar", "Apr", "May", "Jun", "Jul", "Aug", "Sep", "Oct", "Nov", "Dec"],
    color="red")
ax[1].set_title("Ranked Flares per Month")
fig.show()
```

```
# Here, we plot the number of flare occurrences in each month
# Below that, we show the number of top 50 flares per month

# There are a few observations we can draw from these graphs.
# Firstly, February and November are the lowest and highest in both graphs,
# ↪ respectively.
# We might expect a slight dip in February simply due to its lower number of
# ↪ days,
# but not to the extent that appears here.
# November sticks out significantly. If I were investigating flares, I would
# try to plan experiments around November for more results, or if flares effect
# ↪ my tests,
# I would work in the lower occurrence months such as December-January.
```



[ ]: