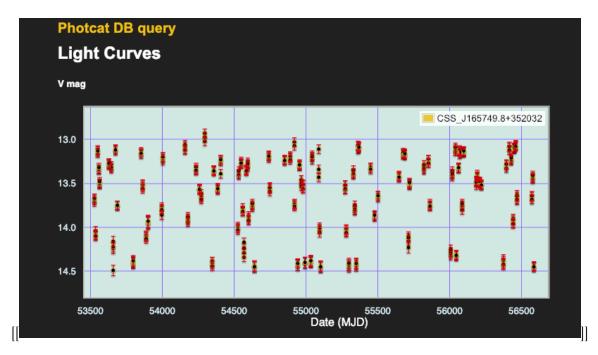
Ph21 Problem Set 1

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Introduction

Here we are looking at data from the Catalina Real-Time Transient Survey. Earlier in the set, the website http://nesssi.cacr.caltech.edu/cgi-bin/getcssconedbid_release2.cgi was searched manually for the Her X-1 binary system. This is the plot the website generated in terms of intensity over time:



We will now attempt to recreate this plot by accessing the website data using python. We will parse the data first using string methods and next using astopy.io.votable.

Imports

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import urllib.request
from astropy.io.votable import parse
sns.set(font_scale=1.5)
```

Reading in the Data

These value querys were found by inspecting the source code of website, checking the values of 'name =' fields:

Now we need to decode this to convert it back to a string:

```
data_str = html.decode('ascii')
```

Parsing the Output Using String Methods

Here we find that the real table begins somewhere in the middle of the file, after a tag combination. We'll truncate this and see what we get:

```
ind_str = ''
trunced_data = data_str[data_str.index(ind_str):]

print(trunced_data[:1000])
```

135075253.0187535.2777812.8251.246251.24633.842

Here we see about four sections separated by newline characters before getting to the actual Photometry data table. So we just need to split over the newline character and then take all the elements beyond:

```
table_data = trunced_data.split('\n')[4:]
```

Let's check out the data now:

```
table_data[:10]
```

```
['113507504547713.340.05254.457535.342353557.33004
'13507504547713.340.05254.457535.342453557.33004
'13507504547713.320.05254.457635.342453557.33418
'13507504547713.510.05254.457635.342353557.33834
'13507504547713.510.05254.457535.342353562.24598
'13507504547713.500.05254.457535.342353562.24598
'13507504547713.500.05254.457535.342353562.25304
'13507504547713.500.05254.457535.342353562.26014
'13507504547713.480.05254.457535.342353562.26718
'13507504547713.710.05254.457535.342353526.30884
'13507504547713.690.05254.457535.342353526.30884
'13507504547713.690.05254.457535.342353526.30884
'13507504547713.690.05254.457535.342353526.31949
'13507504547713.690.05254.457535.342353526.31949
'13507504547713.670.05254.457535.342353526.31949
'13507504547713.670.05254.457535.342353526.31949
'13507504547713.670.05254.254.457535.3423342353526.31949
```

Here we see that we just need to take out the s and s found in each line:

```
def parse_line(table_line):
    splitted = table_line.split('')
    # remove first element, which is just 
    tr_rem = splitted[1:]
    # get rid of 
    from the last element
    tr_rem[-1] = tr_rem[-1].strip('
    //
    # make every item a float
    return list(map(float, tr_rem))
```

```
parse_line(table_data[1])
```

[1135075045477.0, 13.34, 0.05, 254.4575, 35.3424, 53557.33418]

Looks like the parser works. Now we can just parse over the whole table. But, the end of data has some other elements not from the table, so we need to check and see which lines only contain the table.

```
table_data[-4:]
```

```
['113507504547714.450.05254.457635.342456588.10323', '<br>', '<br></HTML>', '']
```

So, we'll parse over everything but the last 3 lines:

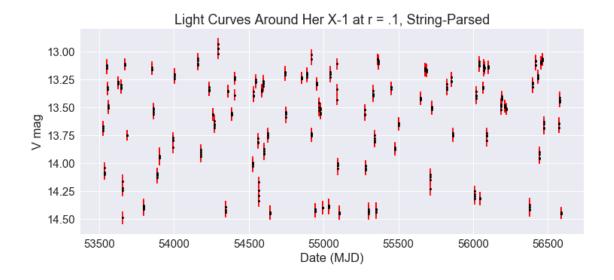
```
parsed_data = np.array(list(map(parse_line, table_data[:-3])))
```

Now, in np-array form, we can just pull out the magnitude, uncertainties, and times of each line from the headers in the HTML code earlier:

```
mags = parsed_data[:, 1]
uncerts = parsed_data[:, 2]
MJDs = parsed_data[:, 5]
```

Plotting the Parsed Data

Now, we just need to plot this with error bars:



Parsing the data from the XML file in VOTable Format:

Now we can get the VOtable XML data from online and use the voparser to parser the data in a much easier fashion:

```
votable = parse("Documents/Ph21/result_web_fileDR8owK.vot", pedantic=False)
vo_data = votable.get_first_table().to_table()
```

vo_data

```
<Table masked=True length=378>
            RAJ2000
                     DEJ2000
MasterID
                              ObsTime [1]
                                           Mag [1]
                                                      Magerr [1] Blend [1]
deg
                     d
object
                                float32
                                           float32
                                                      float32
                                                                   int32
           float32
                    float32
1135075045477 254.45757 35.34235
                                    53557.324 13.338659 0.052161247
                                                                             0
1135075045477 254.45753 35.34234
                                     53557.33 13.339489 0.052163184
                                                                             0
1135075045477 254.45753 35.34235
                                    53557.336 13.342177 0.052178945
                                                                             0
1135075045477 254.45757 35.34235
                                                                             0
                                     53557.34 13.324588 0.05213488
1135075045477 254.45750 35.34233
                                    53562.246 13.510649 0.052456874
                                                                             0
1135075045477 254.45750 35.34233
                                    53562.254 13.49694 0.052435752
                                                                             0
1135075045477 254.45750 35.34231
                                     53562.26 13.500379 0.052439135
                                                                             0
1135075045477 254.45750 35.34234
                                    53562.266 13.475543 0.052392595
                                                                             0
1135075045477 254.45750 35.34233
                                     53526.31 13.70976 0.052931767
                                                                             0
1135075045477 254.45757 35.34233
                                    56574.137 13.675402 0.05290823
                                                                             0
```

```
1135075045477 254.45757 35.34234
                                     56574.14 13.675675 0.052921385
                                                                             0
1135075045477 254.45753 35.34234
                                                                             0
                                    56580.086 13.435135 0.052297443
1135075045477 254.45757 35.34232
                                     56580.09 13.445594 0.052325822
                                                                             0
1135075045477 254.45757 35.34232
                                    56580.098 13.42858 0.052298788
                                                                             0
1135075045477 254.45755 35.34232
                                                                             0
                                    56580.105 13.405985 0.052251775
1135075045477 254.45757 35.34234
                                    56588.082 14.451928
                                                         0.05430378
                                                                             0
1135075045477 254.45755 35.34236
                                     56588.09 14.453248
                                                          0.0543382
                                                                             0
1135075045477 254.45757 35.34235
                                    56588.098 14.436412
                                                         0.05427364
                                                                             0
1135075045477 254.45757 35.34235
                                      56588.1 14.448011 0.054356683
                                                                             0
```

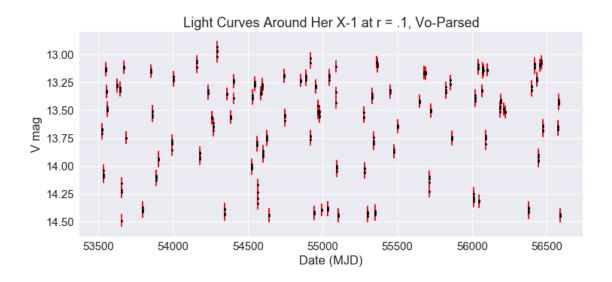
Here we see the relevant fields we need to find:

```
vo_mags = np.array(vo_data['Mag'])
vo_errs = np.array(vo_data['Magerr'])
vo_MJDs = np.array(vo_data['ObsTime'])
```

Plotting the VOTable Data

Now we can plot the data just as before:

```
plot_mag_data(vo_mags, vo_errs, vo_MJDs)
plt.title('Light Curves Around Her X-1 at r = .1, Vo-Parsed')
plt.show()
```



Looks like both plots match each other, as well as the plot generated by the website.