

Lec15-astropy2-RyanSponzilli

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1 ASTR 310 Lecture 15 - astropy 2

1.0.1 1. Reading a table and converting coordinates/times (10 points total)

- Download the file `fermi_lat_grbs.dat` from today's exercise page on the website. It contains 1000 gamma-ray bursts observed by the Fermi satellite between 2010 and 2018.
- Use the `astropy.io.ascii.read()` routine to read the file into a table. You need to specify the `header_start`, `data_start`, and `delimiter` arguments.
- Display the column headers using the `columns` or `colnames` method. Print the units on the `ra` and `dec` columns.

[3 pts]

```
[112]: import astropy.io.ascii
import astropy.coordinates
import astropy.units as u
import numpy as np
```

```
[113]: table = astropy.io.ascii.read("fermi_lat_grbs.dat", guess=False,
    ↪header_start=2, data_start=3, delimiter='|')
print(table.colnames)
print(table['ra'].unit)
print(table['dec'].unit)
```

```
['col0', 'name', 'ra', 'dec', 'trigger_time', 't90', 't90_error', 't90_start',
'fluence', 'fluence_error', 'flux_1024', 'flux_1024_error', 'flux_1024_time',
'flux_64', 'flux_64_error', 'flnc_band_ampl', 'flnc_band_ampl_pos_err',
'flnc_band_ampl_neg_err', 'flnc_band_epeak', 'flnc_band_epeak_pos_err',
'flnc_band_epeak_neg_err', 'flnc_band_alpha', 'flnc_band_alpha_pos_err',
'flnc_band_alpha_neg_err', 'flnc_band_beta', 'flnc_band_beta_pos_err',
'flnc_band_beta_neg_err', 'flnc_spectrum_start', 'flnc_spectrum_stop',
'pflx_best_fitting_model', 'pflx_best_model_redchisq',
'flnc_best_fitting_model', 'flnc_best_model_redchisq', '_1']
```

None

None

- Notice that RA and DEC are in sexagesimal format, represented as strings. Use `astropy.coordinates.Angle` to convert them to decimal degrees. Print the table or try

the `show_in_notebook` method. (The method sometimes gets carried away, so you might like to restrict it to printing the first 10 rows.) [3 pts]

```
[114]: table['ra'] = astropy.coordinates.Angle(table['ra'], unit=u.degree)
table['dec'] = astropy.coordinates.Angle(table['dec'], unit=u.degree)
table
```

```
[114]: <Table length=1000>
      col0      name      ra      ... flnc_best_model_redchisq      _1
      int64      str12      float64      ...      float64      int64
-----
-- GRB120830212  22.524666666666665 ...      0.891  --
-- GRB140323433  23.797277777777778 ...      1.131  --
-- GRB181007737  14.081333333333333 ...      --      --
-- GRB100325246  13.942666666666668 ...      1.003  --
-- GRB100507577  0.19333333333333333 ...      0.998  --
-- GRB171206122      0.632 ...      1.001  --
-- GRB140518709  16.269333333333332 ...      1.025  --
-- GRB130623488  1.3815277777777778 ...      1.061  --
-- GRB110212550  20.755333333333333 ...      1.169  --
...
-- GRB120121101  15.711333333333332 ...      0.979  --
-- GRB170510217  10.660722222222223 ...      1.205  --
-- GRB170802638  3.4866666666666667 ...      0.993  --
-- GRB150523690  13.484666666666666 ...      1.121  --
-- GRB140619490  20.330666666666666 ...      1.08   --
-- GRB170607946  16.444055555555554 ...      1.297  --
-- GRB120926426  3.9813333333333336 ...      0.886  --
-- GRB130627372  12.296333333333333 ...      0.894  --
-- GRB101102840  18.978666666666665 ...      0.987  --
```

- Do some brief analysis:
 - a) Construct a boolean mask array that selects only those bursts with durations $t_{90} < 2$ seconds and relative errors in the duration $\frac{t_{90_error}}{t_{90}} < 50\%$.
 - b) Construct a second boolean mask array that selects $t_{90} > 2$ seconds with the same relative error.
 - c) Compare the median peak energy (flnc_band_epeak, in keV) of the two samples. Use the NumPy `median` function to compute the values. [4 pts]

```
[115]: table['t90'].unit = u.second
table['t90_error'].unit = u.second
table['flnc_band_epeak'].unit = u.keV

mask1 = (table['t90'] < 2*u.second) & ((table['t90_error'] / table['t90']) < 0.
↪5)
```

```
mask2 = (table['t90'] > 2*u.second) & ((table['t90_error'] / table['t90']) < 0.
↪5)
```

```
print(np.median(table['flnc_band_epeak'][mask1].quantity))
print(np.median(table['flnc_band_epeak'][mask2].quantity))
```

359.3643 keV

135.48020000000002 keV

1.0.2 2. Working with and modifying a table (10 points total)

Download the files “mcxc.dat” and “mcxc.readme” from today’s exercise page on the website. These files contain a catalog of X-ray-detected clusters of galaxies from Piffaretti et al. (2011) obtained through the VizieR service at the University of Strasbourg.

1. Use `astropy.io.ascii.read()` to read the table and its metadata into Python. This table is in “CDS” format, and you specify the metadata file using the `readme` argument.
2. Extract the log of L500 (luminosity in units of 10^{44} erg/s), log of M500 (mass in 10^{14} solar masses), and z (redshift) columns from the data into 1D arrays. Create a mask array selecting those clusters with redshift < 0.1 .
3. Now construct the array r containing $r = \log L - 1.64 \log M$. The X-ray luminosity and mass of galaxy clusters are correlated roughly such that $L \propto M^{5/3}$, so the range in r is small.
4. Modify the table to add a masked column (`MaskedColumn` object) for r , and use the description “ $r = \log L - 1.64 \log M$ ” for that column. Use the mask array you created to mask those clusters you don’t want to store r values for. Print the summary information for the table, and print the table to check whether your column was added correctly. You should see “–” for masked-out rows.
5. Write the modified table to a file named “mcxc_new.csv” in comma-separated value (CSV) format.

```
[116]: import astropy.table
```

```
t = astropy.io.ascii.read("mcxc.dat", readme="mcxc.readme", format='cds')
log_l500 = np.log10(t['L500'])
log_m500 = np.log10(t['M500'])
z = t['z']

z_mask = z < 0.1

r = log_l500 - 1.64*log_m500

t['r'] = astropy.table.MaskedColumn(r, mask=~z_mask)
t['r'].description = "r = log(L) - 1.64 * log(M)"
```

```
[117]: t
```

[117]: <Table length=1743>

MCXC	OName	...	L500r4	r
		...		1e+37 W
str12	str18	...	float64	float64
-----	-----	...	-----	-----
J0000.1+0816	RXC J0000.1+0816	...	--	-0.49006049429220355
J0000.4-0237	RXC J0000.4-0237	...	--	-0.4908979504552985
J0001.6-1540	RXC J0001.6-1540	...	--	--
J0001.9+1204	RXC J0001.9+1204	...	--	--
J0003.1-0605	RXCJ0003.1-0605	...	--	--
J0003.2-3555	RXCJ0003.2-3555	...	--	-0.4855196142700461
J0003.8+0203	RXCJ0003.8+0203	...	--	-0.46430026883136716
J0004.9+1142	RXC J0004.9+1142	...	--	-0.4723642366254167
J0005.3+1612	RXC J0005.3+1612	...	--	--
...
J2355.1-1500	BVH2007 242	...	--	-0.4676040941865678
J2355.6+1120	RXC J2355.6+1120	...	--	-0.474121679853096
J2355.7+1138	A2678	...	--	-0.4739568466752166
J2355.8+3423	RXC J2355.8+3423	...	--	--
J2357.0-3445	RXCJ2357.0-3445	...	--	-0.4862703253095651
J2359.3-6042	RXCJ2359.3-6042	...	--	-0.4610630708726696
J2359.4-3418	MS2356.9-3434	...	--	--
J2359.5-3211	RX J2359.5-3211	...	--	--
J2359.9-3928	RXCJ2359.9-3928	...	--	--

[118]: t.write("mcxc_new.csv", format='csv', overwrite=True)