

Using_astropy

June 30, 2017

```
In [1]: # The following is to know when this notebook has been run and with which python version.
import time, sys
print(time.ctime())
print(sys.version.split('|')[0])
```

```
Fri Jun 30 13:30:36 2017
3.6.1
```

1 G The astropy package

The Astropy Project is a community effort to develop a single core package for Astronomy in Python and foster interoperability between Python astronomy packages. More informations here: <http://www.astropy.org/>

!!! WARNING !!!

To install atpy, one must use the `--no-deps` option when using pip (otherwise updates of numpy may be performed):

`pip install -U --no-deps astropy`

```
In [2]: %matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
```

1.0.1 Constants and Units

<http://docs.astropy.org/en/stable/constants/index.html>

<http://docs.astropy.org/en/stable/units/index.html>

```
In [3]: from astropy import constants as const
from astropy import units as u
help(const)
```

Help on package `astropy.constants` in `astropy`:

NAME

`astropy.constants`

DESCRIPTION

Contains astronomical and physical constants for use in Astropy or other places.

A typical use case might be::

```
>>> from astropy.constants import c, m_e
>>> # ... define the mass of something you want the rest energy of as m ...
>>> m = m_e
```

```
>>> E = m * c**2
>>> E.to('MeV') # doctest: +FLOAT_CMP
<Quantity 0.510998927603161 MeV>
```

The following constants are available:

Name	Value	Unit	Description
G	6.67384e-11	m ³ / (kg s ²)	Gravitational constant
L_bol0	3.0128e+28	W	Luminosity for absolute bolometric magnitude 0
L_sun	3.846e+26	W	Solar luminosity
M_earth	5.9742e+24	kg	Earth mass
M_jup	1.8987e+27	kg	Jupiter mass
M_sun	1.9891e+30	kg	Solar mass
N_A	6.02214129e+23	1 / (mol)	Avogadro's number
R	8.3144621	J / (K mol)	Gas constant
R_earth	6378136	m	Earth equatorial radius
R_jup	71492000	m	Jupiter equatorial radius
R_sun	695508000	m	Solar radius
Ryd	10973731.6	1 / (m)	Rydberg constant
a0	5.29177211e-11	m	Bohr radius
alpha	0.00729735257		Fine-structure constant
atmosphere	101325	Pa	Atmosphere
au	1.49597871e+11	m	Astronomical Unit
b_wien	0.0028977721	m K	Wien wavelength displacement law constant
c	299792458	m / (s)	Speed of light in vacuum
e	1.60217657e-19	C	Electron charge
eps0	8.85418782e-12	F/m	Electric constant
g0	9.80665	m / s ²	Standard acceleration of gravity
h	6.62606957e-34	J s	Planck constant
hbar	1.05457173e-34	J s	Reduced Planck constant
k_B	1.3806488e-23	J / (K)	Boltzmann constant
kpc	3.08567758e+19	m	Kiloparsec
m_e	9.10938291e-31	kg	Electron mass
m_n	1.67492735e-27	kg	Neutron mass
m_p	1.67262178e-27	kg	Proton mass
mu0	1.25663706e-06	N/A ²	Magnetic constant
muB	9.27400968e-24	J/T	Bohr magneton
pc	3.08567758e+16	m	Parsec
sigma_T	6.65245873e-29	m ²	Thomson scattering cross-section
sigma_sb	5.670373e-08	W / (K ⁴ m ²)	Stefan-Boltzmann constant
u	1.66053892e-27	kg	Atomic mass

PACKAGE CONTENTS

```
cgs
constant
setup_package
si
tests (package)
```

DATA

```
G = <Constant name='Gravitational constant' value=6...e-15 unit='m3 /...
```

```

L_bol0 = <Constant name='Luminosity for absolute bolometr...0.0 unit='...
L_sun = <Constant name='Solar luminosity' value=3.846e+2...rence="Alle...
M_earth = <Constant name='Earth mass' value=5.9742e+24 unc...rence="Al...
M_jup = <Constant name='Jupiter mass' value=1.8987e+27 u...rence="Alle...
M_sun = <Constant name='Solar mass' value=1.9891e+30 unc...rence="Alle...
N_A = <Constant name="Avogadro's number" value=6.02214...y=2.7e+16 uni...
R = <Constant name='Gas constant' value=8.3144621 un...5e-06 unit='J /...
R_earth = <Constant name='Earth equatorial radius' value=6...rence="Al...
R_jup = <Constant name='Jupiter equatorial radius' value...rence="Alle...
R_sun = <Constant name='Solar radius' value=695508000.0 ...rence="Alle...
Ryd = <Constant name='Rydberg constant' value=10973731...nty=5.5e-05 u...
a0 = <Constant name='Bohr radius' value=5.2917721092e...rtainty=1.7e-2...
absolute_import = _Feature((2, 5, 0, 'alpha', 1), (3, 0, 0, 'alpha', 0...
alpha = <Constant name='Fine-structure constant' value=0...ertainty=2...
atmosphere = <Constant name='Atmosphere' value=101325 uncertainty=0.0 ...
au = <Constant name='Astronomical Unit' value=1495978...=0.0 unit='m' ...
b_wien = <Constant name='Wien wavelength displacement law...ainty=2.6e...
c = <Constant name='Speed of light in vacuum' value=...rtainty=0.0 uni...
division = _Feature((2, 2, 0, 'alpha', 2), (3, 0, 0, 'alpha', 0), 8192...
e = <Constant name='Electron charge' value=1.6021765...rtainty=3.5e-27...
eps0 = <Constant name='Electric constant' value=8.85418...rtainty=0.0 ...
g0 = <Constant name='Standard acceleration of gravity...tainty=0.0 uni...
h = <Constant name='Planck constant' value=6.6260695...ainty=2.9e-41 u...
hbar = <Constant name='Reduced Planck constant' value=1...496649644e-4...
k_B = <Constant name='Boltzmann constant' value=1.3806...nty=1.3e-29 u...
kpc = <Constant name='Kiloparsec' value=3.085677581467...rtainty=0.0 u...
m_e = <Constant name='Electron mass' value=9.10938291e...ertainty=4e-3...
m_n = <Constant name='Neutron mass' value=1.674927351e...tainty=7.4e-3...
m_p = <Constant name='Proton mass' value=1.672621777e-...tainty=7.4e-3...
mu0 = <Constant name='Magnetic constant' value=1.25663...tainty=0.0 un...
muB = <Constant name='Bohr magneton' value=9.27400968e...ainty=2e-31 u...
pc = <Constant name='Parsec' value=3.0856775814671916...rtainty=0.0 un...
printfunction = _Feature((2, 6, 0, 'alpha', 2), (3, 0, 0, 'alpha', 0)...
sigma_T = <Constant name='Thomson scattering cross-section...tainty=1...
sigma_sb = <Constant name='Stefan-Boltzmann constant' value...1e-13 un...
u = <Constant name='Atomic mass' value=1.660538921e-...tainty=7.3e-35 ...
unicode_literals = _Feature((2, 6, 0, 'alpha', 2), (3, 0, 0, 'alpha', ...

```

FILE

```

/home/morisset/anaconda2/envs/py3k6/lib/python3.6/site-packages/astropy/constants/__init__.py

```

```

In [4]: # Pretty printing
        print(const.c)

```

```

Name      = Speed of light in vacuum
Value     = 299792458.0
Uncertainty = 0.0
Unit      = m / s
Reference = CODATA 2010

```

```

In [5]: # .to change the unit
        print(const.c.to('Mpc/yr'))

```

```

3.0660139378795275e-07 Mpc / yr

```

```

In [6]: # basic operations are managed
        const.c * 2

Out[6]:
 $5.9958492 \times 10^8 \frac{\text{m}}{\text{s}}$ 

In [7]: np.sqrt(const.c)

Out[7]:
 $17314.516 \frac{\text{m}^{1/2}}{\text{s}^{1/2}}$ 

In [8]: print(np.sqrt(const.c))

17314.51581766005 m(1/2) / s(1/2)

In [9]: # Following the units
        M1 = 3 * const.M_sun
        M2 = 100 * u.g
        Dist = 2.2 * u.au
        F = const.G * M1 * M2 / Dist ** 2
        print(M1)
        print(F)

5.9673e+30 kg
8.228265585123966e+21 g m3 / (AU2 s2)

In [10]: F

Out[10]:
 $8.2282656 \times 10^{21} \frac{\text{m}^3 \text{g}}{\text{AU}^2 \text{s}^2}$ 

In [11]: # Convert in more classical unit
        print(F.to(u.N))

0.0003676693920278125 N

In [12]: q = 42.0 * u.meter

In [13]: q**2

Out[13]:
1764 m2

In [14]: # Extract only the value
        (q**2).value

Out[14]: 1764.0

In [15]: arr = np.array([q.value, q.value]) * const.G
        print(type(arr))
        print(arr)

<class 'astropy.units.quantity.Quantity'>
[ 2.80301280e-09  2.80301280e-09] m3 / (kg s2)

In [16]: arr = np.ones(2) * q * const.G
        print(type(arr))
        print(arr)

```

```

<class 'astropy.units.quantity.Quantity'>
[ 2.80301280e-09  2.80301280e-09] m4 / (kg s2)

In [17]: # Resolving redondant units
t = 3.0 * u.kilometer / (130.51 * u.meter / u.second)
print(t)
print(t.decompose())

0.022986744310780783 km s / m
22.986744310780782 s

In [18]: x = 1.0 * u.parsec
print(x.to(u.km))

30856775814671.914 km

In [19]: lam = 5007 * u.angstrom

In [20]: print(lam.to(u.nm))
print(lam.to(u.micron))

500.70000000000005 nm
0.5007000000000001 micron

In [21]: # Some transformations needs extra information, available from u.special
print(lam.to(u.Ry, equivalencies=u.spectral()))

0.18199861205330833 Ry

```

More in <http://docs.astropy.org/en/stable/units/index.html>

1.0.2 Data Table

<http://docs.astropy.org/en/stable/table/index.html>

```

In [22]: from astropy.table import Table

In [23]: # create a table with non homogeneous types
a = [1, 4, 5]
b = [2.0, 5.0, 8.2]
c = ['x', 'y', 'z']
t = Table([a, b, c], names=('a', 'b', 'c'), meta={'name': 'first table'})
print(t)

```

a	b	c
1	2.0	x
4	5.0	y
5	8.2	z

```

In [24]: # Pretty output
t

```

```

Out[24]: <Table length=3>
      a      b      c
    int64 float64 str1
-----
      1      2.0      x
      4      5.0      y
      5      8.2      z

```

```

In [25]: # One can change the output format
t['b'].format = '7.3f'
t['a'].format = '{:.4f}'
# and add units
t['b'].unit = 's'
t

Out[25]: <Table length=3>
      a      b      c
      s
  int64 float64 str1
-----
  1.0000  2.000    x
  4.0000  5.000    y
  5.0000  8.200    z

In [26]: t.show_in_browser(jsviewer=True)

In [27]: # access the column names
t.colnames

Out[27]: ['a', 'b', 'c']

In [28]: # length of the table (number of rows)
len(t)

Out[28]: 3

In [29]: # Acces one element
t['a'][1]

Out[29]: 4

In [30]: # Modify one element
t['a'][1] = 10
t

Out[30]: <Table length=3>
      a      b      c
      s
  int64 float64 str1
-----
  1.0000  2.000    x
 10.0000  5.000    y
  5.0000  8.200    z

In [31]: # easy add column:
t['d'] = [1, 2, 3]

In [32]: t

Out[32]: <Table length=3>
      a      b      c      d
      s
  int64 float64 str1 int64
-----
  1.0000  2.000    x      1
 10.0000  5.000    y      2
  5.0000  8.200    z      3

```

```
In [33]: t.rename_column('a', 'A')
t
```

```
Out[33]: <Table length=3>
      A      b      c      d
      s
  int64 float64 str1 int64
-----
  1.0000  2.000    x     1
 10.0000  5.000    y     2
  5.0000  8.200    z     3
```

```
In [34]: t.add_row([-8, -9.3, 'r', 10])
t
```

```
Out[34]: <Table length=4>
      A      b      c      d
      s
  int64 float64 str1 int64
-----
  1.0000  2.000    x     1
 10.0000  5.000    y     2
  5.0000  8.200    z     3
 -8.0000 -9.300    r    10
```

```
In [35]: t.add_row([-9, 40, 'q', 10])
t
```

```
Out[35]: <Table length=5>
      A      b      c      d
      s
  int64 float64 str1 int64
-----
  1.0000  2.000    x     1
 10.0000  5.000    y     2
  5.0000  8.200    z     3
 -8.0000 -9.300    r    10
 -9.0000 40.000    q    10
```

```
In [36]: # Masked values
t = Table([a, b, c], names=('a', 'b', 'c'), masked=True)
t['a'].mask = [True, True, False] # True is for the masked values!!
t
```

```
Out[36]: <Table masked=True length=3>
      a      b      c
  int64 float64 str1
-----
   --    2.0    x
   --    5.0    y
    5    8.2    z
```

```
In [37]: t['a'].mask = [True, False, False] # True is for the masked values!!
t
```

```
Out[37]: <Table masked=True length=3>
      a      b      c
```

	int64	float64	str1
--		2.0	x
4		5.0	y
5		8.2	z

```
In [38]: # Creat a table from a table
t2 = Table([t['a']**2, t['b']**2, t['a']**2 + t['b']**2], names=('a2', 'b2', 'a2+b2'))
t2
```

```
Out[38]: <Table masked=True length=3>
```

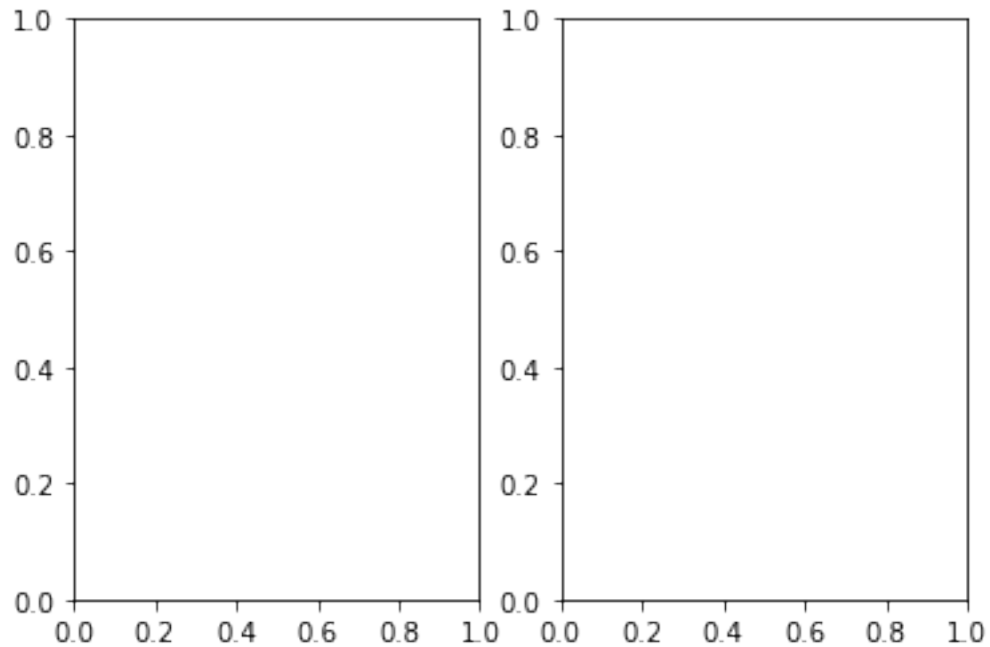
	a2	b2	a2+b2
	int64	float64	float64
--		4.0	--
16		25.0	41.0
25		67.24	92.24

```
In [39]: # Managing columns
from astropy.table import Column
```

```
In [40]: # Create a table combining different formats
a = (1, 4)
b = np.array([[2, 3], [5, 6]]) # vector column
c = Column(['x', 'y'], name='axis')
f, (ax1, ax2) = plt.subplots(1,2)
d = Column([ax1, ax2], name='axis obj')
arr = (a, b, c, d)
t3 = Table(arr) # Data column named "c" has a name "axis" in that table
t3
```

```
Out[40]: <Table length=2>
```

	col0	col1	[2]	axis	axis obj
	int64	int64	str1		object
1	2	..	3	x	Axes(0.125,0.125;0.352273x0.755)
4	5	..	6	y	Axes(0.547727,0.125;0.352273x0.755)



```
In [41]: # table from a dictionnary
```

```
rr = {'a': [1, 4],
      'b': [2.0, 5.0],
      'c': ('x', 'y')}
t4 = Table(rr)
t4
```

```
Out[41]: <Table length=2>
```

```
   a      b      c
int64 float64 str1
---- -
```

1	2.0	x
4	5.0	y

```
In [42]: # Create table row by row
```

```
t5 = Table(rows=[{'a': 5, 'b': 10}, {'a': 15, 'b': 30}])
t5
```

```
Out[42]: <Table length=2>
```

```
   a      b
int64 int64
---- -
```

5	10
15	30

```
In [44]: # Numpy structured array
```

```
arr = np.array([(1, 2.0, 'x'),
                (4, 5.0, 'y')],
               dtype=[('a', 'i8'), ('b', 'f8'), ('c', 'S2')])
print(arr)
t6 = Table(arr)
print(t6)
```

```
[(1, 2., b'x') (4, 5., b'y')]
a   b   c
--- --- ---
1 2.0   x
4 5.0   y
```

Python arrays versus numpy arrays as input

There is a slightly subtle issue that is important to understand in the way that Table objects are created. Any data input that looks like a Python list (including a tuple) is considered to be a list of columns. In contrast an homogeneous numpy array input is interpreted as a list of rows:

```
In [45]: t7 = Table(((1,2,3), (4,5,6), (7,8,9)))
         t7
```

```
Out[45]: <Table length=3>
         col0  col1  col2
         int64 int64 int64
         -----
           1     4     7
           2     5     8
           3     6     9
```

```
In [47]: arr7 = np.array(((1,2,3), (4,5,6)))
         t7 = Table(arr7)
         print(arr7)
         print(t7)
```

```
[[1 2 3]
 [4 5 6]]
col0 col1 col2
---- ---- ----
   1    2    3
   4    5    6
```

```
In [48]: arr = np.array([(1, 2.0, 'x'),
                        (4, 5.0, 'y')],
                        dtype=[('a', 'i8'), ('b', 'f8'), ('c', 'S2')])
         t6 = Table(arr, copy=False) # pointing to the original data
         arr['a'][0] = 99
         print(arr)
         print(t6)
```

```
[(99, 2., b'x') ( 4, 5., b'y')]
a   b   c
--- --- ---
99 2.0   x
 4 5.0   y
```

```
In [49]: t6.columns
```

```
Out[49]: TableColumns([( 'a', <Column name='a' dtype='int64' length=2>
                        99
                        4), ('b', <Column name='b' dtype='float64' length=2>
                        2.0
                        5.0), ('c', <Column name='c' dtype='bytes2' length=2>
                        x
                        y)])
```

```
In [50]: t6.colnames
```

```
Out[50]: ['a', 'b', 'c']
```

```
In [51]: # One can obtain a numpy structured array from a Table
np.array(t6)
```

```
Out[51]: array([(99, 2., b'x'), ( 4, 5., b'y')],
              dtype=[('a', '<i8'), ('b', '<f8'), ('c', 'S2')])
```

```
In [52]: arr = np.arange(3000).reshape(100, 30) # 100 rows x 30 columns array
t = Table(arr)
print(t)
```

col0	col1	col2	col3	col4	col5	col6	...	col23	col24	col25	col26	col27	col28	col29
0	1	2	3	4	5	6	...	23	24	25	26	27	28	29
30	31	32	33	34	35	36	...	53	54	55	56	57	58	59
60	61	62	63	64	65	66	...	83	84	85	86	87	88	89
90	91	92	93	94	95	96	...	113	114	115	116	117	118	119
120	121	122	123	124	125	126	...	143	144	145	146	147	148	149
150	151	152	153	154	155	156	...	173	174	175	176	177	178	179
180	181	182	183	184	185	186	...	203	204	205	206	207	208	209
210	211	212	213	214	215	216	...	233	234	235	236	237	238	239
240	241	242	243	244	245	246	...	263	264	265	266	267	268	269
270	271	272	273	274	275	276	...	293	294	295	296	297	298	299
...
2670	2671	2672	2673	2674	2675	2676	...	2693	2694	2695	2696	2697	2698	2699
2700	2701	2702	2703	2704	2705	2706	...	2723	2724	2725	2726	2727	2728	2729
2730	2731	2732	2733	2734	2735	2736	...	2753	2754	2755	2756	2757	2758	2759
2760	2761	2762	2763	2764	2765	2766	...	2783	2784	2785	2786	2787	2788	2789
2790	2791	2792	2793	2794	2795	2796	...	2813	2814	2815	2816	2817	2818	2819
2820	2821	2822	2823	2824	2825	2826	...	2843	2844	2845	2846	2847	2848	2849
2850	2851	2852	2853	2854	2855	2856	...	2873	2874	2875	2876	2877	2878	2879
2880	2881	2882	2883	2884	2885	2886	...	2903	2904	2905	2906	2907	2908	2909
2910	2911	2912	2913	2914	2915	2916	...	2933	2934	2935	2936	2937	2938	2939
2940	2941	2942	2943	2944	2945	2946	...	2963	2964	2965	2966	2967	2968	2969
2970	2971	2972	2973	2974	2975	2976	...	2993	2994	2995	2996	2997	2998	2999

Length = 100 rows

```
In [53]: t.show_in_browser(jsviewer=True)
```

```
In [54]: # create a simple table to play with
arr = np.arange(15).reshape(5, 3)
t = Table(arr, names=('a', 'b', 'c'), meta={'keywords': {'key1': 'val1'}})
t
```

```
Out[54]: <Table length=5>
   a      b      c
int64 int64 int64
-----
   0      1      2
   3      4      5
   6      7      8
   9     10     11
  12     13     14
```

```
In [55]: t['a'] = [1, -2, 3, -4, 5] # Set all
t
```

```
Out[55]: <Table length=5>
      a      b      c
int64 int64 int64
-----
      1      1      2
     -2      4      5
      3      7      8
     -4     10     11
      5     13     14
```

```
In [56]: t['a'][2] = 30 # set one
t
```

```
Out[56]: <Table length=5>
      a      b      c
int64 int64 int64
-----
      1      1      2
     -2      4      5
     30      7      8
     -4     10     11
      5     13     14
```

```
In [57]: # set one row
t[1] = (8, 9, 10)
t
```

```
Out[57]: <Table length=5>
      a      b      c
int64 int64 int64
-----
      1      1      2
      8      9     10
     30      7      8
     -4     10     11
      5     13     14
```

```
In [58]: # Set a whole column
t['a'] = 99
t
```

```
Out[58]: <Table length=5>
      a      b      c
int64 int64 int64
-----
     99      1      2
     99      9     10
     99      7      8
     99     10     11
     99     13     14
```

```
In [59]: # Add a column
t.add_column(Column(np.array([1,2,3,4,5]), name='d'))
t
```

```
Out[59]: <Table length=5>
```

a	b	c	d
int64	int64	int64	int64
99	1	2	1
99	9	10	2
99	7	8	3
99	10	11	4
99	13	14	5

```
In [60]: # remove a column
t.remove_column('b')
t
```

```
Out[60]: <Table length=5>
```

a	c	d
int64	int64	int64
99	2	1
99	10	2
99	8	3
99	11	4
99	14	5

```
In [61]: # add a row
t.add_row([-8, -9, 10])
t
```

```
Out[61]: <Table length=6>
```

a	c	d
int64	int64	int64
99	2	1
99	10	2
99	8	3
99	11	4
99	14	5
-8	-9	10

```
In [62]: # Remove some rows
t.remove_rows([1, 2])
t
```

```
Out[62]: <Table length=4>
```

a	c	d
int64	int64	int64
99	2	1
99	11	4
99	14	5
-8	-9	10

```
In [63]: # sort the Table using one column
t.sort('c')
t
```

```
Out[63]: <Table length=4>
      a      c      d
    int64 int64 int64
-----
      -8     -9     10
      99      2      1
      99     11      4
      99     14      5
```

```
In [64]: filter = (t['a'] > 50) & (t['d'] > 3)
          print(filter)
```

```
[False False  True  True]
```

```
In [65]: t[filter]
```

```
Out[65]: <Table length=2>
      a      c      d
    int64 int64 int64
-----
      99     11      4
      99     14      5
```

```
In [66]: %%writefile tab1.dat
#name      obs_date      mag_b  mag_v
M31        2012-01-02    17.0    17.5
M31        2012-01-02    17.1    17.4
M101       2012-01-02    15.1    13.5
M82        2012-02-14    16.2    14.5
M31        2012-02-14    16.9    17.3
M82        2012-02-14    15.2    15.5
M101       2012-02-14    15.0    13.6
M82        2012-03-26    15.7    16.5
M101       2012-03-26    15.1    13.5
M101       2012-03-26    14.8    14.3
```

Overwriting tab1.dat

```
In [67]: # directly read a Table from an ascii file
          obs = Table.read('tab1.dat', format='ascii')
```

```
In [69]: print(obs)
```

```
name  obs_date  mag_b  mag_v
-----
M31 2012-01-02  17.0  17.5
M31 2012-01-02  17.1  17.4
M101 2012-01-02  15.1  13.5
M82 2012-02-14  16.2  14.5
M31 2012-02-14  16.9  17.3
M82 2012-02-14  15.2  15.5
M101 2012-02-14  15.0  13.6
M82 2012-03-26  15.7  16.5
M101 2012-03-26  15.1  13.5
M101 2012-03-26  14.8  14.3
```

```

In [70]: # Group data
         obs_by_name = obs.group_by('name')
         obs_by_name

Out[70]: <Table length=10>
         name  obs_date  mag_b  mag_v
         str4   str10   float64 float64
         ----  -
M101  2012-01-02    15.1    13.5
M101  2012-02-14    15.0    13.6
M101  2012-03-26    15.1    13.5
M101  2012-03-26    14.8    14.3
M31   2012-01-02    17.0    17.5
M31   2012-01-02    17.1    17.4
M31   2012-02-14    16.9    17.3
M82   2012-02-14    16.2    14.5
M82   2012-02-14    15.2    15.5
M82   2012-03-26    15.7    16.5

In [71]: print(obs_by_name.groups.keys)

name
----
M101
M31
M82

In [72]: # Using 2 keys to group
         print(obs.group_by(['name', 'obs_date']).groups.keys)

name  obs_date
----  -
M101  2012-01-02
M101  2012-02-14
M101  2012-03-26
M31   2012-01-02
M31   2012-02-14
M82   2012-02-14
M82   2012-03-26

In [73]: # Extracting a group
         print(obs_by_name.groups[1])

name  obs_date  mag_b  mag_v
----  -
M31   2012-01-02  17.0  17.5
M31   2012-01-02  17.1  17.4
M31   2012-02-14  16.9  17.3

In [74]: # Using a mask to select entries
         mask = obs_by_name.groups.keys['name'] == 'M101'
         print(mask)
         print(obs_by_name.groups[mask])

[ True False False]
name  obs_date  mag_b  mag_v

```

```

-----
M101 2012-01-02 15.1 13.5
M101 2012-02-14 15.0 13.6
M101 2012-03-26 15.1 13.5
M101 2012-03-26 14.8 14.3

```

```

In [75]: # Some functions can be applied to the elements of a group
         obs_mean = obs_by_name.groups.aggregate(np.mean)
         print(obs_mean)

```

```

name mag_b mag_v
-----
M101 15.0 13.725
M31 17.0 17.4
M82 15.7 15.5

```

WARNING: Cannot aggregate column 'obs_date' with type '<U10' [astropy.table.groups]

```

In [76]: print(obs_by_name['name', 'mag_v', 'mag_b'].groups.aggregate(np.mean))

```

```

name mag_v mag_b
-----
M101 13.725 15.0
M31 17.4 17.0
M82 15.5 15.7

```

```

In [77]: # creat a new Table on the fly
         obs1 = Table.read("""name      obs_date      mag_b  logLx
M31      2012-01-02  17.0   42.5
M82      2012-10-29  16.2   43.5
M101     2012-10-31  15.1   44.5""", format='ascii')

```

```

In [78]: # this is used to stack Tables
         from astropy.table import vstack

```

```

In [79]: tvs = vstack([obs, obs1])
         tvs

```

```

Out[79]: <Table masked=True length=13>
         name obs_date mag_b mag_v logLx
         str4 str10 float64 float64 float64
         -----
M31 2012-01-02 17.0 17.5 --
M31 2012-01-02 17.1 17.4 --
M101 2012-01-02 15.1 13.5 --
M82 2012-02-14 16.2 14.5 --
M31 2012-02-14 16.9 17.3 --
M82 2012-02-14 15.2 15.5 --
M101 2012-02-14 15.0 13.6 --
M82 2012-03-26 15.7 16.5 --
M101 2012-03-26 15.1 13.5 --
M101 2012-03-26 14.8 14.3 --
M31 2012-01-02 17.0 -- 42.5
M82 2012-10-29 16.2 -- 43.5
M101 2012-10-31 15.1 -- 44.5

```



```
In [80]: %%writefile data6.dat
Line      Iobs      lambda  rel_er  Obs_code
H 1  4861A  1.00000    4861.  0.08000  Anabel
H 1  6563A  2.8667     6563.  0.19467  Anabel
H 1  4340A  0.4933     4340.  0.03307  Anabel
H 1  4102A  0.2907     4102.  0.02229  Anabel
H 1  3970A  0.1800     3970.  0.01253  Anabel
N 2  6584A  2.1681     6584.  0.08686  Anabel
N 2 121.7m  0.0044621217000.  0.20000  Liu
O 1  6300A  0.0147      6300.  0.00325  Anabel
TOTL 2326A  0.07900    2326.  0.20000  Adams
C 2 157.6m  0.00856 1576000.  0.20000  Liu
O 1  63.17m 0.13647  631700.  0.10000  Liu
O 1 145.5m 0.00446 1455000.  0.200    Liu
TOTL 3727A  0.77609    3727.  0.200    Torres-Peimbert
S II 4070A  0.06174    4070.  0.200    Torres-Peimbert
S II 4078A  0.06174    4078.  0.200    Torres-Peimbert
```

Overwriting data6.dat

```
In [81]: d = Table.read('data6.dat', format='ascii.fixed_width',
                        col_starts=(0, 12, 20, 29, 38))
d
```

```
Out[81]: <Table length=15>
      Line      Iobs      lambda  rel_er  Obs_code
      str11    float64    float64  float64    str15
-----
H 1  4861A      1.0      4861.0    0.08      Anabel
H 1  6563A      2.8667    6563.0  0.19467    Anabel
H 1  4340A      0.4933    4340.0  0.03307    Anabel
H 1  4102A      0.2907    4102.0  0.02229    Anabel
H 1  3970A      0.18      3970.0  0.01253    Anabel
N 2  6584A      2.1681    6584.0  0.08686    Anabel
N 2 121.7m  0.004462 1217000.0    0.2      Liu
O 1  6300A      0.0147    6300.0  0.00325    Anabel
TOTL 2326A      0.079    2326.0    0.2      Adams
C 2 157.6m  0.00856 1576000.0    0.2      Liu
O 1  63.17m  0.13647  631700.0    0.1      Liu
O 1 145.5m  0.00446 1455000.0    0.2      Liu
TOTL 3727A      0.77609    3727.0    0.2 Torres-Peimbert
S II 4070A      0.06174    4070.0    0.2 Torres-Peimbert
S II 4078A      0.06174    4078.0    0.2 Torres-Peimbert
```

```
In [82]: d.group_by('Obs_code')
```

```
Out[82]: <Table length=15>
      Line      Iobs      lambda  rel_er  Obs_code
      str11    float64    float64  float64    str15
-----
TOTL 2326A      0.079    2326.0    0.2      Adams
H 1  4861A      1.0      4861.0    0.08      Anabel
H 1  6563A      2.8667    6563.0  0.19467    Anabel
H 1  4340A      0.4933    4340.0  0.03307    Anabel
H 1  4102A      0.2907    4102.0  0.02229    Anabel
```

H	1	3970A	0.18	3970.0	0.01253	Anabel
N	2	6584A	2.1681	6584.0	0.08686	Anabel
O	1	6300A	0.0147	6300.0	0.00325	Anabel
N	2	121.7m	0.004462	1217000.0	0.2	Liu
C	2	157.6m	0.00856	1576000.0	0.2	Liu
O	1	63.17m	0.13647	631700.0	0.1	Liu
O	1	145.5m	0.00446	1455000.0	0.2	Liu
TOTL		3727A	0.77609	3727.0	0.2	Torres-Peimbert
S II		4070A	0.06174	4070.0	0.2	Torres-Peimbert
S II		4078A	0.06174	4078.0	0.2	Torres-Peimbert

There is a lot of possibilities of joining Tables, see <http://docs.astropy.org/en/stable/table/operations.html>

```
In [83]: t = Table.read("ftp://cdsarc.u-strasbg.fr/pub/cats/J/other/RMxAA/45.261/digeda.dat",
                        format='ascii.cds',
                        readme='ftp://cdsarc.u-strasbg.fr/pub/cats/J/other/RMxAA/45.261/ReadMe')
```

Downloading <ftp://cdsarc.u-strasbg.fr/pub/cats/J/other/RMxAA/45.261/digeda.dat> [Done]

Downloading <ftp://cdsarc.u-strasbg.fr/pub/cats/J/other/RMxAA/45.261/ReadMe> [Done]

```
In [84]: t
```

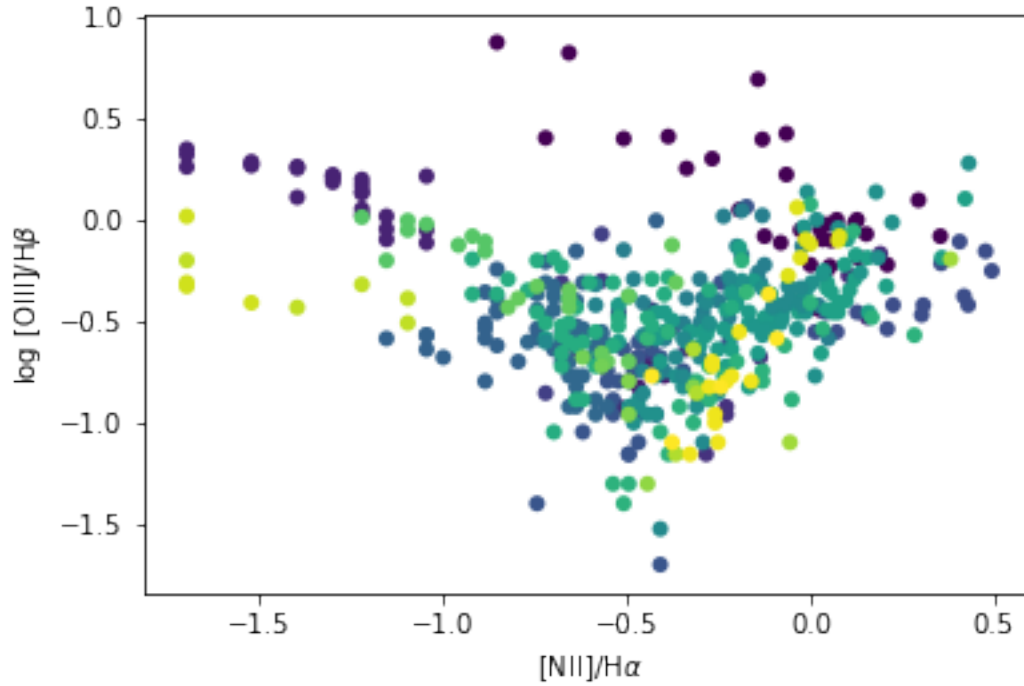
```
Out[84]: <Table masked=True length=1061>
```

ObsID	Pos	I3727	I4363	IHb	I4959	...	MType	Slit	Region	GalID	RefN
	pc					...					
int64	float64	float64	float64	float64	float64	...	int64	int64	int64	int64	int64
1	0.03	--	--	1.0	0.2	...	12	3	1	2	1
2	0.03	--	--	1.0	0.33	...	12	3	1	2	1
3	0.05	--	--	1.0	0.32	...	12	3	1	2	1
4	0.06	--	--	1.0	0.12	...	12	3	1	2	1
5	0.07	--	--	1.0	0.27	...	12	3	1	2	1
6	0.12	--	--	1.0	0.31	...	12	3	1	2	1
7	0.13	--	--	1.0	0.29	...	12	3	1	2	1
8	0.15	--	--	1.0	0.3	...	12	3	1	2	1
9	0.15	--	--	1.0	0.57	...	12	3	1	2	1
...
1052	-1.0	--	--	0.35	--	...	2	3	3	92	44
1053	-1.0	--	--	0.35	--	...	2	3	3	92	44
1054	-1.0	--	--	0.35	--	...	2	3	3	92	44
1055	-1.0	--	--	0.35	--	...	2	3	1	92	44
1056	-1.0	--	--	0.35	--	...	2	3	3	92	44
1057	-1.0	--	--	0.35	--	...	2	3	1	92	44
1058	-1.0	--	--	0.35	--	...	2	3	3	92	44
1059	-1.0	--	--	0.35	--	...	2	3	3	92	44
1060	-1.0	--	--	0.35	--	...	2	3	1	92	44
1061	-1.0	--	--	0.35	--	...	2	3	3	92	44

```
In [85]: t.show_in_browser(jsviewer=True)
```

```
In [86]: plt.scatter(np.log10(t['I6583']), np.log10(t['I5007']), c=t['RefN'], edgecolor='None')
           plt.xlabel(r'[NII]/H$\alpha$')
           plt.ylabel(r'log [OIII]/H$\beta$')
```

```
Out[86]: <matplotlib.text.Text at 0x7fea59a0c978>
```



```
In [87]: t = Table.read("ftp://cdsarc.u-strasbg.fr/pub/cats/VII/253/snrs.dat",
readme="ftp://cdsarc.u-strasbg.fr/pub/cats/VII/253/ReadMe",
format="ascii.cds")
```

Downloading ftp://cdsarc.u-strasbg.fr/pub/cats/VII/253/snrs.dat [Done]

Downloading ftp://cdsarc.u-strasbg.fr/pub/cats/VII/253/ReadMe [Done]

```
In [88]: t
```

```
Out[88]: <Table masked=True length=274>
```

SNR	RAh	RAm	RA s	DE-	...	u.S(1GHz)	Sp-Index	u.Sp-Index	Names
	h	min	s		...				
str11	int64	int64	int64	str1	...	str1	float64	str1	str26
-----	-----	-----	-----	-----	-----	-----	-----	-----	-----
G000.0+00.0	17	45	44	-	?	0.8	? Sgr A East	
G000.3+00.0	17	46	15	-	--	0.6	--	--
G000.9+00.1	17	47	21	-	?	--	v	--
G001.0-00.1	17	48	30	-	--	0.6	?	--
G001.4-00.1	17	49	39	-	?	--	?	--
G001.9+00.3	17	48	45	-	--	0.6	--	--
G003.7-00.2	17	55	26	-	--	0.65	--	--
G003.8+00.3	17	52	55	-	?	0.6	--	--
G004.2-03.5	18	8	55	-	?	0.6	?	--
...
G356.3-00.3	17	37	56	-	?	--	?	--
G356.3-01.5	17	42	35	-	?	--	?	--
G357.7-00.1	17	40	29	-	--	0.4	-- MSH 17-39	
G357.7+00.3	17	38	35	-	--	0.4	?	--
G358.0+03.8	17	26	0	-	?	--	?	--

G358.1+00.1	17	37	0	- ...	?	--	?	--
G358.5-00.9	17	46	10	- ...	?	--	?	--
G359.0-00.9	17	46	50	- ...	--	0.5	--	--
G359.1-00.5	17	45	30	- ...	--	0.4	?	--
G359.1+00.9	17	39	36	- ...	?	--	?	--

```
In [89]: t.show_in_browser(jsviewer=True)
```

```
In [91]: t[0:10].write('tab_cds1.tex', format='latex', overwrite=True)
```

```
In [92]: !cat tab_cds1.tex
```

```
\begin{table}
\begin{tabular}{cccccccccccccccc}
SNR & RAh & RAh & RAS & DE- & DEd & DEd & MajDiam & --- & MinDiam & u_MinDiam & type & l_S(1GHz) & S(1GHz) & u_S(1GHz) & Sp-Index & u_Sp-Index \\
& & $ \mathrm{h} $ & & $ \mathrm{min} $ & & $ \mathrm{s} $ & & & $ \mathrm{min} $ & & & & & & & & \\
G000.0+00.0 & 17 & 45 & 44 & - & 29 & 0 & 3.5 & x & 2.5 & & S & & 100.0 & ? & 0.8 & ? & Sgr A East \\
G000.3+00.0 & 17 & 46 & 15 & - & 28 & 38 & 15.0 & x & 8.0 & & S & & 22.0 & & 0.6 & & \\
G000.9+00.1 & 17 & 47 & 21 & - & 28 & 9 & 8.0 & & & C & & 18.0 & ? & & v & & \\
G001.0-00.1 & 17 & 48 & 30 & - & 28 & 9 & 8.0 & & & S & & 15.0 & & 0.6 & ? & & \\
G001.4-00.1 & 17 & 49 & 39 & - & 27 & 46 & 10.0 & & & S & & 2.0 & ? & & ? & & \\
G001.9+00.3 & 17 & 48 & 45 & - & 27 & 10 & 1.5 & & & S & & 0.6 & & 0.6 & & & \\
G003.7-00.2 & 17 & 55 & 26 & - & 25 & 50 & 14.0 & x & 11.0 & & S & & 2.3 & & 0.65 & & \\
G003.8+00.3 & 17 & 52 & 55 & - & 25 & 28 & 18.0 & & & S? & & 3.0 & ? & 0.6 & & & \\
G004.2-03.5 & 18 & 8 & 55 & - & 27 & 3 & 28.0 & & & S & & 3.2 & ? & 0.6 & ? & & \\
G004.5+06.8 & 17 & 30 & 42 & - & 21 & 29 & 3.0 & & & S & & 19.0 & & 0.64 & & Kepler, SN1604, 3C471 \\
\end{tabular}
\end{table}
```

```
In [95]: t[10:20].write('tab_cds1.ascii', format='ascii', delimiter='|', formats={'Sp-Index': '%0.2f'},
```

```
In [96]: !cat tab_cds1.ascii
```

```
SNR|RAh|RAh|RAS|DE-|DEd|DEd|MajDiam|---|MinDiam|u_MinDiam|type|l_S(1GHz)|S(1GHz)|u_S(1GHz)|Sp-Index|u_Sp-Index
G004.8+06.2|17|33|25|-|21|34|18.0|||S||3.0||0.60||
G005.2-02.6|18|7|30|-|25|45|18.0|||S||2.6||0.60|?|
G005.4-01.2|18|2|10|-|24|54|35.0|||C?||35.0|?|0.20|?|Milne 56
G005.5+00.3|17|57|4|-|24|0|15.0|x|12.0||S||5.5||0.70||
G005.9+03.1|17|47|20|-|22|16|20.0|||S||3.3|?|0.40|?|
G006.1+00.5|17|57|29|-|23|25|18.0|x|12.0||S||4.5||0.90||
G006.1+01.2|17|54|55|-|23|5|30.0|x|26.0|F||4.0|?|0.30|?|
G006.4-00.1|18|0|30|-|23|26|48.0|||C||310.0|||v|W28
G006.4+04.0|17|45|10|-|21|22|31.0|||S||1.3|?|0.40|?|
G006.5-00.4|18|2|11|-|23|34|18.0|||S||27.0||0.60||
```

```
In [98]: t[10:20].write('tab_cds2.ascii', format='ascii.fixed_width', delimiter='|', formats={'Sp-Index': '%0.2f'},
```

```
In [99]: !cat tab_cds2.ascii
```

SNR	RAh	RAh	RAS	DE-	DEd	DEd	MajDiam	---	MinDiam	u_MinDiam	type	l_S(1GHz)	S(1GHz)	u_S(1GHz)	Sp-Index	u_Sp-Index
G004.8+06.2	17	33	25	-	21	34	18.0				S					
G005.2-02.6	18	7	30	-	25	45	18.0				S					
G005.4-01.2	18	2	10	-	24	54	35.0				C?					
G005.5+00.3	17	57	4	-	24	0	15.0	x	12.0		S					
G005.9+03.1	17	47	20	-	22	16	20.0				S					
G006.1+00.5	17	57	29	-	23	25	18.0	x	12.0		S					

G006.1+01.2	17	54	55	-	23	5	30.0	x	26.0		F
G006.4-00.1	18	0	30	-	23	26	48.0				C
G006.4+04.0	17	45	10	-	21	22	31.0				S
G006.5-00.4	18	2	11	-	23	34	18.0				S

The astropy Table can also read FITS files (if containing tables), VO tables and hdf5 format. See more there: <http://docs.astropy.org/en/stable/io/unified.html>

1.0.3 Time and Dates

The astropy.time package provides functionality for manipulating times and dates. Specific emphasis is placed on supporting time scales (e.g. UTC, TAI, UT1, TDB) and time representations (e.g. JD, MJD, ISO 8601) that are used in astronomy and required to calculate, e.g., sidereal times and barycentric corrections. It uses Cython to wrap the C language ERFA time and calendar routines, using a fast and memory efficient vectorization scheme. More here: <http://docs.astropy.org/en/stable/time/index.html>

1.0.4 Coordinates

The coordinates package provides classes for representing a variety of celestial/spatial coordinates, as well as tools for converting between common coordinate systems in a uniform way.

```
In [100]: from astropy import units as u
          from astropy.coordinates import SkyCoord

In [101]: c = SkyCoord(ra=10.5*u.degree, dec=41.2*u.degree, frame='icrs')
          c

Out[101]: <SkyCoord (ICRS): (ra, dec) in deg
          ( 10.5,  41.2)>

In [102]: c = SkyCoord('0 42 00 +41 12 00', frame='icrs', unit=(u.hourangle, u.deg))
          c

Out[102]: <SkyCoord (ICRS): (ra, dec) in deg
          ( 10.5,  41.2)>

In [103]: print(c.ra, c.dec)
10d30m00s 41d12m00s

In [104]: c.to_string('decimal')
Out[104]: '10.5 41.2'

In [105]: print(c.dec.to_string(format='latex'))
$41^{\circ}12'00''$
```

$41^{\circ}12'00''$

1.0.5 Modeling

astropy.modeling provides a framework for representing models and performing model evaluation and fitting. It currently supports 1-D and 2-D models and fitting with parameter constraints.

It is designed to be easily extensible and flexible. Models do not reference fitting algorithms explicitly and new fitting algorithms may be added without changing the existing models (though not all models can be used with all fitting algorithms due to constraints such as model linearity).

The goal is to eventually provide a rich toolset of models and fitters such that most users will not need to define new model classes, nor special purpose fitting routines (while making it reasonably easy to do when necessary).

<http://docs.astropy.org/en/stable/modeling/index.html>

1.0.6 Convolution and filtering

`astropy.convolution` provides convolution functions and kernels that offers improvements compared to the `scipy.ndimage` convolution routines, including:

- Proper treatment of NaN values
- A single function for 1-D, 2-D, and 3-D convolution
- Improved options for the treatment of edges
- Both direct and Fast Fourier Transform (FFT) versions
- Built-in kernels that are commonly used in Astronomy

More on <http://docs.astropy.org/en/stable/convolution/index.html>