

Using astropy

January 10, 2021

```
[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])
```

Tue Nov 24 12:18:28 2020

3.7.9 (default, Aug 31 2020, 07:22:35)

[Clang 10.0.0]

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/>

The astropy group received US\$ 1 million last year to found astropy for 3 years.

<https://www.slideshare.net/KelleCruz/astropy-project-update-for-adass>

<https://learn.astropy.org/>

```
[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>

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

4.0.2

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	6.6743e-11	m ³ / (kg s ²)	Gravitational constant
N_A	6.02214076e+23	1 / (mol)	Avogadro's number
R	8.31446262	J / (K mol)	Gas constant
Ryd	10973731.6	1 / (m)	Rydberg constant
a0	5.29177211e-11	m	Bohr radius
alpha	0.00729735257		Fine-structure constant
atm	101325	Pa	Standard atmosphere
b_wien	0.00289777196	m K	Wien wavelength displacement law constant
c	299792458	m / (s)	Speed of light in vacuum
e	1.60217663e-19	C	Electron charge
eps0	8.85418781e-12	F/m	Vacuum electric permittivity
g0	9.80665	m / s ²	Standard acceleration of gravity
h	6.62607015e-34	J s	Planck constant
hbar	1.05457182e-34	J s	Reduced Planck constant
k_B	1.380649e-23	J / (K)	Boltzmann constant
m_e	9.1093837e-31	kg	Electron mass
m_n	1.6749275e-27	kg	Neutron mass
m_p	1.67262192e-27	kg	Proton mass
mu0	1.25663706e-06	N/A ²	Vacuum magnetic permeability
muB	9.27401008e-24	J/T	Bohr magneton
sigma_T	6.65245873e-29	m ²	Thomson scattering cross-section
sigma_sb	5.67037442e-08	W / (K ⁴ m ²)	Stefan-Boltzmann constant
u	1.66053907e-27	kg	Atomic mass
GM_earth	3.986004e+14	m ³ / (s ²)	Nominal Earth mass parameter

GM_jup	1.2668653e+17	m3 / (s2)	Nominal Jupiter mass parameter
GM_sun	1.3271244e+20	m3 / (s2)	Nominal solar mass parameter
L_bol0	3.0128e+28	W	Luminosity for absolute bolometric magnitude 0
L_sun	3.828e+26	W	Nominal solar luminosity
M_earth	5.97216787e+24	kg	Earth mass
M_jup	1.8981246e+27	kg	Jupiter mass
M_sun	1.98840987e+30	kg	Solar mass
R_earth	6378100	m	Nominal Earth equatorial radius
R_jup	71492000	m	Nominal Jupiter equatorial radius
R_sun	695700000	m	Nominal solar radius
au	1.49597871e+11	m	Astronomical Unit
kpc	3.08567758e+19	m	Kiloparsec
pc	3.08567758e+16	m	Parsec
=====			

PACKAGE CONTENTS

```

astropyconst13
astropyconst20
astropyconst40
cgs
codata2010
codata2014
codata2018
config
constant
iau2012
iau2015
si
tests (package)
utils

```

SUBMODULES

```

codata
iaudata

```

FUNCTIONS

```

set_enabled_constants(modname)
    .. deprecated:: 4.0
        The set_enabled_constants function is deprecated and may be removed
in a future version.
        Use 'astropy.physical_constants' and
'astropy.astronomical_constants' instead.

Context manager to temporarily set values in the ``constants``
namespace to an older version.
See :ref:`astropy-constants-prior` for usage.

```

Parameters

modname : {'astropyconst13', 'astropyconst20'}
Name of the module containing an older version.

DATA

```
G = <<class 'astropy.constants.codata2018.CODATA2018...e-15 unit='m3 /...
GM_earth = <<class 'astropy.constants.iau2015.IAU2015'> nam...it='m3 /...
GM_jup = <<class 'astropy.constants.iau2015.IAU2015'> nam...it='m3 / s...
GM_sun = <<class 'astropy.constants.iau2015.IAU2015'> nam...it='m3 / s...
L_bol0 = <<class 'astropy.constants.iau2015.IAU2015'> nam...0.0 unit='...
L_sun = <<class 'astropy.constants.iau2015.IAU2015'> nam...0.0 unit='W...
M_earth = <<class 'astropy.constants.iau2015.IAU2015'> nam...eference=...
M_jup = <<class 'astropy.constants.iau2015.IAU2015'> nam...eference='I...
M_sun = <<class 'astropy.constants.iau2015.IAU2015'> nam...eference='I...
N_A = <<class 'astropy.constants.codata2018.CODATA2018...ainty=0.0 uni...
R = <<class 'astropy.constants.codata2018.CODATA2018...y=0.0 unit='J /...
R_earth = <<class 'astropy.constants.iau2015.IAU2015'> nam...0.0 unit=...
R_jup = <<class 'astropy.constants.iau2015.IAU2015'> nam...0.0 unit='m...
R_sun = <<class 'astropy.constants.iau2015.IAU2015'> nam...0.0 unit='m...
Ryd = <<class 'astropy.constants.codata2018.CODATA2018...nty=2.1e-05 u...
a0 = <<class 'astropy.constants.codata2018.CODATA2018...certainty=8e-2...
alpha = <<class 'astropy.constants.codata2018.CODATA2018...ertainty=1...
atm = <<class 'astropy.constants.codata2018.CODATA2018...ncertainty=0...
au = <<class 'astropy.constants.iau2015.IAU2015'> nam...=0.0 unit='m' ...
b_wien = <<class 'astropy.constants.codata2018.CODATA2018...certainty=...
c = <<class 'astropy.constants.codata2018.CODATA2018...rtainty=0.0 uni...
e = <<class 'astropy.constants.codata2018.EMCODATA20...uncertainty=0.0...
eps0 = <<class 'astropy.constants.codata2018.EMCODATA20...nty=1.3e-21 ...
g0 = <<class 'astropy.constants.codata2018.CODATA2018...tainty=0.0 uni...
h = <<class 'astropy.constants.codata2018.CODATA2018...certainty=0.0 u...
hbar = <<class 'astropy.constants.codata2018.CODATA2018...certainty=0...
k_B = <<class 'astropy.constants.codata2018.CODATA2018...rtainty=0.0 u...
kpc = <<class 'astropy.constants.iau2015.IAU2015'> nam...ived from au ...
m_e = <<class 'astropy.constants.codata2018.CODATA2018...tainty=2.8e-4...
m_n = <<class 'astropy.constants.codata2018.CODATA2018...tainty=9.5e-3...
m_p = <<class 'astropy.constants.codata2018.CODATA2018...tainty=5.1e-3...
mu0 = <<class 'astropy.constants.codata2018.CODATA2018...ty=1.9e-16 un...
muB = <<class 'astropy.constants.codata2018.CODATA2018...nty=2.8e-33 u...
pc = <<class 'astropy.constants.iau2015.IAU2015'> nam...ived from au +...
sigma_T = <<class 'astropy.constants.codata2018.CODATA2018...ertainty=...
sigma_sb = <<class 'astropy.constants.codata2018.CODATA2018...y=0.0 un...
u = <<class 'astropy.constants.codata2018.CODATA2018...ertainty=5e-37 ...
```

FILE

/Users/christophemorisset/anaconda3/lib/python3.7/site-
packages/astropy/constants/__init__.py

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

```
Name    = Speed of light in vacuum
Value   = 299792458.0
Uncertainty = 0.0
Unit    = m / s
Reference = CODATA 2018
```

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

```
3.0660139378555056e-07 Mpc / yr
```

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

```
[6]:  $8.9875518 \times 10^{16} \frac{\text{m}^2}{\text{s}^2}$ 
```

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

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

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

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

```
[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)
```

```
5.965229612094153e+30 kg
8.225977685950412e+21 g m3 / (AU2 s2)
```

```
[10]: F
```

```
[10]:  $8.2259777 \times 10^{21} \frac{\text{m}^3 \text{g}}{\text{AU}^2 \text{s}^2}$ 
```

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

```
0.0003675671602160826 N
```

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

```
[13]: q**2
```

```
[13]: 1764 m2
```

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

```
1764.0
```

```
1764.0
```

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

```
<class 'astropy.units.quantity.Quantity'>
[2.803206e-09 2.803206e-09] m3 / (kg s2)
```

```
[16]: arr2 = np.ones(2) * q
arr2[1] = q*3
arr2
```

```
[16]: [42, 126] m
```

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

```
<class 'astropy.units.quantity.Quantity'>
[2.803206e-09 2.803206e-09] m4 / (kg s2)
```

```
[18]: # 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
```

```
[19]: x = 1.0 * u.parsec
print(x.to(u.km))
```

```
30856775814913.67 km
```

```
[20]: lam = 5007 * u.angstrom
```

```
[21]: print(lam.to(u.nm))
      print(lam.to(u.micron))
```

```
500.70000000000005 nm
0.5007000000000001 micron
```

```
[22]: # Some transformations needs extra information, available from u.special
      print(lam.to(u.GHz, equivalencies=u.spectral()))
```

```
598746.6706610745 GHz
```

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>

```
[23]: from astropy.table import Table
```

```
[24]: # 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

```
[25]: # Pretty output
      t
```

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

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

```
[26]: <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
```

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

```
[28]: # access the column names
      t.colnames
```

```
[28]: ['a', 'b', 'c']
```

```
[29]: # length of the table (number of rows)
      len(t)
```

```
[29]: 3
```

```
[30]: # Acces one element
      t['a'][1]
```

```
[30]: 4
```

```
[31]: # Modify one element
      t['a'][1] = 10
      t
```

```
[31]: <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
```

```
[32]: # easy add column:
      t['d'] = [1, 2, 3]
```

```
[33]: t
```

```
[33]: <Table length=3>
      a      b      c      d
      s
int64 float64 str1 int64
```


int64	float64	str1	int64
1.0000	2.000	x	1
10.0000	5.000	y	2
5.0000	8.200	z	3

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

```
[34]: <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
```

```
[35]: t.add_row([-6.6, -9.3, 'r', 10])
t
```

```
[35]: <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
-6.0000 -9.300    r    10
```

```
[36]: t.add_row([-9, 40, 'q', 10.1])
t
```

```
[36]: <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
-6.0000 -9.300    r    10
-9.0000 40.000    q    10
```

```
[37]: # Masked values
t2 = Table(t, masked=True)
```

```
t2['A'].mask = [True, True, False, False, False, False] # True is for the
↳masked values!!
t2
```

[37]: <Table masked=True length=5>

A	b	c	d
s			
int64	float64	str1	int64
-----	-----	----	-----
--	2.000	x	1
--	5.000	y	2
5.0000	8.200	z	3
-6.0000	-9.300	r	10
-9.0000	40.000	q	10

```
[38]: t2['A'].mask = [True, False, False, False, False, False] # True is for the
↳masked values!!
t2
```

[38]: <Table masked=True length=5>

A	b	c	d
s			
int64	float64	str1	int64
-----	-----	----	-----
--	2.000	x	1
10.0000	5.000	y	2
5.0000	8.200	z	3
-6.0000	-9.300	r	10
-9.0000	40.000	q	10

```
[39]: # 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
```

[39]: <Table length=5>

a2	b2	a2+b2
s		
int64	float64	float64
-----	-----	-----
1.0000	4.000	5.0000
100.0000	25.000	125.0000
25.0000	67.240	92.2400
36.0000	86.490	122.4900
81.0000	1600.000	1681.0000

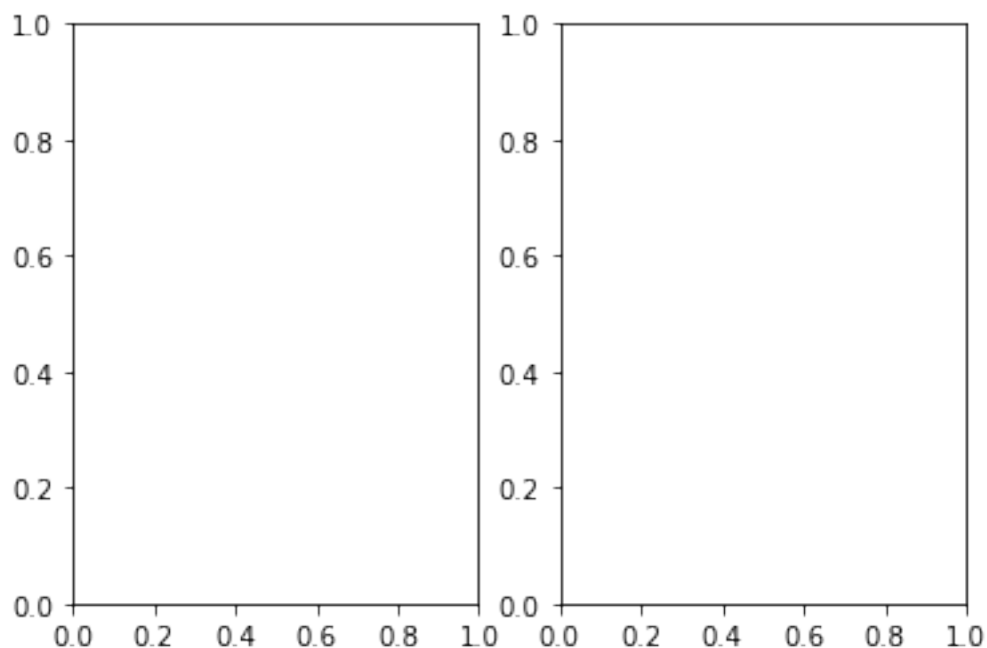
```
[40]: # Managing columns
from astropy.table import Column
```

```
[41]: # Create a table combining different formats
a = (1, 4)
b = np.array([[2, 5, 4, 5, 6, 3], [5, 7, 6]]) # vector column
c = Column(['x', 'y'], name='axis')
f, (ax1, ax2) = plt.subplots(1, 2)
d = Column([ax1, ax2], name='axis obj')
tup = (a, b, c, d)
t3 = Table(tup) # Data column named "c" has a name "axis" in that table
t3
```

/Users/christophemorisset/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:3: VisibleDeprecationWarning: Creating an ndarray from ragged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray

This is separate from the ipykernel package so we can avoid doing imports until

```
[41]: <Table length=2>
      col0      col1      axis      axis obj
      int64      object      str1      object
-----
      1 [2, 5, 4, 5, 6, 3]    x  AxesSubplot(0.125,0.125;0.352273x0.755)
      4      [5, 7, 6]      y  AxesSubplot(0.547727,0.125;0.352273x0.755)
```



```
[42]: # table from a dictionary
```

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

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

```
   a      b      c  
int64 float64 str1  
----  
   1     2.0     x  
   4     5.0     y
```

```
[43]: # Create table row by row
```

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

```
[43]: <Table length=2>
```

```
   a      b      c  
int64 int64 int64  
----  
   5     10     --  
  --     30     15
```

```
[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:

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

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

```
[46]: 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
```

```
[47]: 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
```

```
[48]: t6.columns
```

```
[48]: <TableColumns names=('a','b','c')>
```

```
[49]: t6.colnames
```

```
[49]: ['a', 'b', 'c']
```

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

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

```
[51]: arr = np.arange(9000).reshape(100, 90) # 100 rows x 90 columns array
      t = Table(arr)
      print(t)
```

col0	col1	col2	col3	col4	col5	col6	...	col83	col84	col85	col86	col87	col88	col89
0	1	2	3	4	5	6	...	83	84	85	86	87	88	89
90	91	92	93	94	95	96	...	173	174	175	176	177	178	179
180	181	182	183	184	185	186	...	263	264	265	266	267	268	269
270	271	272	273	274	275	276	...	353	354	355	356	357	358	359
360	361	362	363	364	365	366	...	443	444	445	446	447	448	449
450	451	452	453	454	455	456	...	533	534	535	536	537	538	539
540	541	542	543	544	545	546	...	623	624	625	626	627	628	629
630	631	632	633	634	635	636	...	713	714	715	716	717	718	719
720	721	722	723	724	725	726	...	803	804	805	806	807	808	809
810	811	812	813	814	815	816	...	893	894	895	896	897	898	899
...
8010	8011	8012	8013	8014	8015	8016	...	8093	8094	8095	8096	8097	8098	8099
8100	8101	8102	8103	8104	8105	8106	...	8183	8184	8185	8186	8187	8188	8189
8190	8191	8192	8193	8194	8195	8196	...	8273	8274	8275	8276	8277	8278	8279
8280	8281	8282	8283	8284	8285	8286	...	8363	8364	8365	8366	8367	8368	8369
8370	8371	8372	8373	8374	8375	8376	...	8453	8454	8455	8456	8457	8458	8459
8460	8461	8462	8463	8464	8465	8466	...	8543	8544	8545	8546	8547	8548	8549
8550	8551	8552	8553	8554	8555	8556	...	8633	8634	8635	8636	8637	8638	8639
8640	8641	8642	8643	8644	8645	8646	...	8723	8724	8725	8726	8727	8728	8729
8730	8731	8732	8733	8734	8735	8736	...	8813	8814	8815	8816	8817	8818	8819
8820	8821	8822	8823	8824	8825	8826	...	8903	8904	8905	8906	8907	8908	8909
8910	8911	8912	8913	8914	8915	8916	...	8993	8994	8995	8996	8997	8998	8999

Length = 100 rows

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

```
[53]: # 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
```

```
[53]: <Table length=5>
      a      b      c
      int64 int64 int64
      ----
      0      1      2
      3      4      5
      6      7      8
```

9	10	11
12	13	14

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

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

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

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

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

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

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

[57]: <Table length=5>

a	b	c
int64	int64	int64
-----	-----	-----
99	1	2
99	9	10
99	7	8
99	10	11
99	13	14

[58]: *# Add a column*

```
t.add_column(Column(np.array([1,2,3,4,5]), name='d'))
t
```

[58]: <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

[59]: *# remove a column*

```
t.remove_column('b')
t
```

[59]: <Table length=5>

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

[60]: *# add a row*

```
t.add_row([-8, -9, 10])
t
```

[60]: <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

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

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

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

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

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

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

```
[64]: t[filter]
```

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

```
[65]: %%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

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

```
[67]: obs
```

```
[67]: <Table length=10>
name  obs_date      mag_b      mag_v
str4   str10      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
```

```
[68]: # Group data
obs_by_name = obs.group_by('name')
obs_by_name
```

```
[68]: <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

```
[69]: print(obs_by_name.groups.keys)
```

```
name
----
M101
  M31
  M82
```

```
[70]: # Using 2 keys to group
print(obs.groupby(['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
```

```
[71]: # 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
```

```
[72]: # 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

```
[73]: # 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	15.000000000000002	13.725000000000001
M31	17.0	17.400000000000002
M82	15.699999999999998	15.5

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

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

name	mag_v	mag_b
M101	13.725000000000001	15.000000000000002
M31	17.400000000000002	17.0
M82	15.5	15.699999999999998

```
[75]: # 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')
```

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

```
[77]: tvs = vstack([obs, obs1])
tvs
```

```
[77]: <Table 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

```
[78]: %%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 0.004462 1217000. 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

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

```
[79]: <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
```

```
[80]: d.group_by('Obs_code')
```

```
[80]: <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>

1.0.3 Pandas and Table

```
[81]: df = t.to_pandas()
```

```
[82]: df
```

```
[82]:
   a   c   d
0 -8  -9  10
1 99   2   1
2 99  11   4
3 99  14   5
```

```
[83]: t2 = Table.from_pandas(df)
      t2
```

```
[83]: <Table length=4>
      a      c      d
    int64 int64 int64
-----
      -8      -9      10
      99       2       1
```

```

99    11    4
99    14    5

```

1.0.4 Downloading from CDS

Look for data on “Diffuse gas” at Vizier: <https://vizier.u-strasbg.fr/viz-bin/VizieR>

```
[84]: t = Table.read("https://cdsarc.unistra.fr/ftp/J/other/RMxAA/45.261/digeda.dat",
                    format='ascii.cds',
                    readme='https://cdsarc.unistra.fr/ftp/J/other/RMxAA/45.261/
                    ↪ReadMe')
```

```
[85]: 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')
```

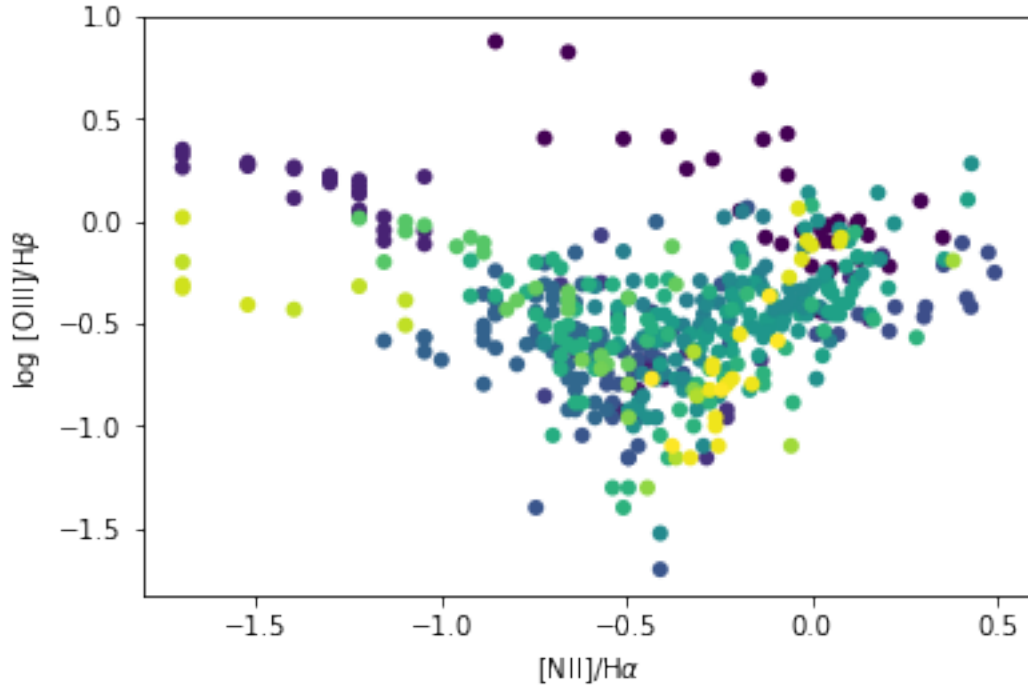
```
[86]: t
```

```
[86]: <Table 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
```

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

```
[88]: 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$\beta$')
```

```
[88]: Text(0, 0.5, 'log [OIII]/H$\beta$')
```



```
[89]: 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")
```

```
[90]: t
```

```
[90]: <Table 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	- ...	?	--	?	--

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

```
[92]: t[0:10].write('tab_cds1.tex', format='latex', overwrite=True,
    ↪ formats={'Sp-Index': '%0.2f'})
```

```
[93]: !cat tab_cds1.tex
```

```
\begin{table}
\begin{tabular}{cccccccccccccccc}
SNR & RAh & RAh & RAs & DE- & DEd & DEm & MajDiam & --- & MinDiam & u_MinDiam & &
type & l_S(1GHz) & S(1GHz) & u_S(1GHz) & Sp-Index & u_Sp-Index & Names & \\\
& $\mathrm{h}$ & $\mathrm{min}$ & $\mathrm{s}$ & & $\mathrm{deg}$ & &
$\mathrm{arcmin}$ & $\mathrm{arcmin}$ & & $\mathrm{arcmin}$ & & &
$\mathrm{Jy}$ & & & & \\\
G000.0+00.0 & 17 & 45 & 44 & - & 29 & 0 & 3.5 & x & 2.5 & & S & & 100.0 & ? & &
0.80 & ? & Sgr A East & \\\
G000.3+00.0 & 17 & 46 & 15 & - & 28 & 38 & 15.0 & x & 8.0 & & S & & 22.0 & & &
0.60 & & & \\\
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 & & 0.60 &
? & & \\\
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 & & 0.60 &
& \\\
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 & ? & &
0.60 & & & \\\
G004.2-03.5 & 18 & 8 & 55 & - & 27 & 3 & 28.0 & & & & S & & 3.2 & ? & 0.60 &
? & & \\\end{tabular}
```

```
G004.5+06.8 & 17 & 30 & 42 & - & 21 & 29 & 3.0 & & & S & & 19.0 & & 0.64
& & Kepler, SN1604, 3C358 \\
\end{tabular}
\end{table}
```

```
[94]: t[10:20].write('tab_cds1.ascii', format='ascii', delimiter=';',
↳formats={'Sp-Index': '%0.2f'}, overwrite=True)
```

```
[95]: !cat tab_cds1.ascii
```

```
SNR;RAh;RAm;RAs;DE-;DEd;DEm;MajDiam;---;MinDiam;u_MinDiam;type;l_S(1GHz);S(1GHz)
;u_S(1GHz);Sp-Index;u_Sp-Index;Names
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;;
```

```
[96]: t[10:20].write('tab_cds2.ascii', format='ascii.fixed_width', delimiter=' ',
↳formats={'Sp-Index': '%0.2f'}, overwrite=True)
```

```
[97]: !cat tab_cds2.ascii
```

	SNR	RAh	RAm	RAs	DE-	DEd	DEm	MajDiam	---	MinDiam	u_MinDiam	
type	l_S(1GHz)	S(1GHz)	u_S(1GHz)	Sp-Index	u_Sp-Index	Names						
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
```

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>

1.0.5 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>

1.0.6 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.

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

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

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

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

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

```
[101]: print(c.ra, c.dec)
```

```
10d30m00s 41d12m00s
```

```
[102]: c.to_string('decimal')
```

```
[102]: '10.5 41.2'
```

```
[103]: print(c.dec.to_string(format='latex'))
```

```
$41^{\circ}12'\prime$
```

```
41°12'00"
```

1.0.7 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>

More examples: <https://learn.astropy.org/rst-tutorials/Models-Quick-Fit.html>

1.0.8 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>

1.0.9 CCD reduction

`Ccdproc` is an Astropy affiliated package for basic data reductions of CCD images. It provides the essential tools for processing of CCD images in a framework that provides error propagation and bad pixel tracking throughout the reduction process.

<https://ccdproc.readthedocs.io/en/latest/>