
Programming and Data Structures with Python

Lecture 17, 22 November 2021

▾ Using numpy

- Arrays and lists
- Arrays are "homogenous" with regular structure
- Lists are flexible

▾ Load numpy

```
1 import numpy as np

1 a = np.array([1,2,3])
2 a
   array([1, 2, 3])

1 b = np.array(range(10))
2 b
   array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])

1 c = np.array([[0,1],[2,3]])
2 c
   array([[0, 1],
          [2, 3]])
```

▾ Indexing and slicing

```
1 a = np.arange(10)**3
2 a
   array([ 0,  1,  8, 27, 64, 125, 216, 343, 512, 729])

1 a[2], a[2:5]
   (8, array([ 8, 27, 64]))

1 a[:6:2] = -1000 # equivalent to a[0:6:2] = -1000
2 a
   array([-1000,  1, -1000,  27, -1000,  125,  216,  343,  512,
           729])

1 def f(x,y):
2     return(10*x + y)

1 f(5,7)
   57

1 b = np.fromfunction(f,(5,4),dtype=int)
2 b
   array([[ 0,  1,  2,  3],
          [10, 11, 12, 13],
          [20, 21, 22, 23],
          [30, 31, 32, 33],
          [40, 41, 42, 43]])

1 b[2,3] # Not b[2][3]
   23

1 b[0:5, 1] # each row in the second column of b
   array([ 1, 11, 21, 31, 41])
```

```

1 b[:,1] # equivalent to the previous example
   array([ 1, 11, 21, 31, 41])

1 b[1:3, :] # each column in the second and third row of b
   array([[10, 11, 12, 13],
          [20, 21, 22, 23]])

1 b[1:4,1:3]
   array([[11, 12],
          [21, 22],
          [31, 32]])

```

▼ Iterating over elements

```

1 print(b)
[[ 0  1  2  3]
 [10 11 12 13]
 [20 21 22 23]
 [30 31 32 33]
 [40 41 42 43]]

1 for row in b:
2     print(row)

[0 1 2 3]
[10 11 12 13]
[20 21 22 23]
[30 31 32 33]
[40 41 42 43]

1 for element in b.flat:
2     print(element,end=' ')

0 1 2 3 10 11 12 13 20 21 22 23 30 31 32 33 40 41 42 43

```

▼ Stacking arrays

```

1 a = np.zeros((5,7),dtype=int)
2 a

array([[0, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0],
       [0, 0, 0, 0, 0, 0, 0]])

1 a = np.floor(10*np.random.random((2,2)))
2 b = np.floor(10*np.random.random((2,2)))
3 print(a,b)

[[2. 8.]
 [2. 7.]] [[6. 7.]
 [1. 7.]]

1 np.vstack((a,b))

array([[2., 8.],
       [2., 7.],
       [6., 7.],
       [1., 7.]])

1 c = np.floor(10*np.random.random((3,3)))

1 c

array([[7., 5., 2.],
       [1., 9., 9.],
       [3., 9., 8.]])

1 np.vstack((a,c))

```



```
-----
ValueError                                Traceback (most recent call last)
<ipython-input-24-eb5bc5f1560c> in <module>()
----> 1 np.vstack((a,c))

<_array_function__ internals> in vstack(*args, **kwargs)

/usr/local/lib/python3.7/dist-packages/numpy/core/shape_base.py in vstack(tup)
    281     if not isinstance(arrs, list):
    282         arrs = [arrs]
--> 283     return _nx.concatenate(arrs, 0)
    284
    285

<_array_function__ internals> in concatenate(*args, **kwargs)

ValueError: all the input array dimensions for the concatenation axis must match exactly, but along dimension 1, the array at index 0
has size 2 and the array at index 1 has size 3
1 np.hstack((a,b))

array([[2., 8., 6., 7.],
       [2., 7., 1., 7.]])
```

```
1 np.hstack((b,c))
```

```
-----
ValueError                                Traceback (most recent call last)
<ipython-input-27-71093adb15f8> in <module>()
----> 1 np.hstack((b,c))

<_array_function__ internals> in hstack(*args, **kwargs)

/usr/local/lib/python3.7/dist-packages/numpy/core/shape_base.py in hstack(tup)
    344     return _nx.concatenate(arrs, 0)
    345     else:
--> 346     return _nx.concatenate(arrs, 1)
    347
    348

<_array_function__ internals> in concatenate(*args, **kwargs)

ValueError: all the input array dimensions for the concatenation axis must match exactly, but along dimension 0, the array at index 0
has size 2 and the array at index 1 has size 3
```

SEARCH STACK OVERFLOW

▼ Splitting arrays

```
1 a = np.floor(10*np.random.random((2,12)))
2 a

array([[9., 5., 7., 2., 3., 3., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]])
```

```
1 np.hsplit(a,6)

[array([[9., 5.],
       [0., 6.]])], array([[7., 2.],
       [9., 5.]])], array([[3., 3.],
       [6., 4.]])], array([[6., 4.],
       [8., 7.]])], array([[0., 4.],
       [4., 7.]])], array([[8., 1.],
       [7., 1.]])]
```

```
1 np.hsplit(a,(2,5,7)) # Split a after the third and the fourth column

[array([[9., 5.],
       [0., 6.]])], array([[7., 2., 3.],
       [9., 5., 6.]])], array([[3., 6.],
       [4., 8.]])], array([[4., 0., 4., 8., 1.],
       [7., 4., 7., 7., 1.]])]
```

```
1 np.vsplit(a,2) # Split a vertically

[array([[9., 5., 7., 2., 3., 3., 6., 4., 0., 4., 8., 1.]])],
array([[0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]])]
```

▼ Copy and view

```
1 c = a.copy() # Creates a disjoint copy of the array
2 d = a.view() # Creates another link to the same array
```

```
1 a, c, d

(array([[9., 5., 7., 2., 3., 3., 6., 4., 0., 4., 8., 1.]])
```

```

    [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]],
array([[9., 5., 7., 2., 3., 3., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]],
array([[9., 5., 7., 2., 3., 3., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]])

```

```
1 c[0,4] = 88
```

```
1 a, c, d
```

```

(array([[9., 5., 7., 2., 3., 3., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]]),
array([[9., 5., 7., 2., 88., 3., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]]),
array([[9., 5., 7., 2., 3., 3., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]])

```

```
1 d[0,5] = 66
```

```
1 a, c, d
```

```

(array([[9., 5., 7., 2., 3., 66., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]]),
array([[9., 5., 7., 2., 88., 3., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]]),
array([[9., 5., 7., 2., 3., 66., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]])

```

Use base to check whether two arrays have same underlying elements

```
1 c.base is a, d.base is a
```

```
(False, True)
```

Can reshape arrays with same base without affecting the shape of the other

```
1 d.shape = 4,6
```

```
1 a,c,d
```

```

(array([[9., 5., 7., 2., 3., 66., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]]),
array([[9., 5., 7., 2., 88., 3., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]]),
array([[9., 5., 7., 2., 3., 66.],
       [6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4.],
       [8., 7., 4., 7., 7., 1.]])

```

```
1 d[2,4] = 99
```

```
1 a, c, d
```

```

(array([[9., 5., 7., 2., 3., 66., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 99., 4., 8., 7., 4., 7., 7., 1.]]),
array([[9., 5., 7., 2., 88., 3., 6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 6., 4., 8., 7., 4., 7., 7., 1.]]),
array([[9., 5., 7., 2., 3., 66.],
       [6., 4., 0., 4., 8., 1.],
       [0., 6., 9., 5., 99., 4.],
       [8., 7., 4., 7., 7., 1.]])

```

```
1 a = np.array([[1,2],[3,4]])
```

```
2 b = np.array([[5,6],[7,8]])
```

```
1 a,b
```

```

(array([[1, 2],
       [3, 4]]), array([[5, 6],
       [7, 8]]))

```

```
1 a+b # Pointwise addition
```

```

array([[6, 8],
       [10, 12]])

```

```
1 a*b # Pointwise multiplication
```

```

array([[5, 12],
       [21, 32]])

```

```

1 np.matmul(a,b) # Normal matrix multiplication
   array([[19, 22],
          [43, 50]])

1 a.T # Transpose
   array([[1, 3],
          [2, 4]])

1 np.linalg.inv(a) # Matrix inverse
   array([[ -2. ,  1. ],
          [ 1.5, -0.5]])

1 np.matmul(a,np.linalg.inv(a)) # a a^-1
   array([[1.0000000e+00, 0.0000000e+00],
          [8.8817842e-16, 1.0000000e+00]])

```

▼ Pandas (Python and data analysis)

- Built on top of numpy

Series and data frames

- Numpy defines homogeneous n-dimensional arrays
- Data science works with tables: 2-dimensional arrays
- Pandas has two fundamental data structures
 - Series : A column of data
 - Data Frame : A table of data

Key difference

- Numpy indices are always [0..n-1] in each dimension
- Pandas allows more flexible "named" indices for rows and columns
 - Dictionary vs list

▼ Load pandas

- Don't need to import numpy unless one is separately using numpy arrays

```
1 import pandas as pd
```

▼ Create a series

- Convert a sequence into a series (column)

```

1 h = ('AA', '2012-02-01', 100, 10.2)
2 s = pd.Series(h)
3 type(s)

```

```
pandas.core.series.Series
```

```
1 s
```

```

0      AA
1  2012-02-01
2      100
3      10.2
dtype: object

```

- Convert a dictionary to a series
- Keys become "row indices"

```

1 d = {'name' : 'IBM', 'date' : '2010-09-08', 'shares' : 100, 'price' : 10.2}
2 ds = pd.Series(d)
3 type(ds)

```

```
pandas.core.series.Series
```

```
1 ds
```

```
name      IBM
date      2010-09-08
shares    100
price     10.2
dtype: object
```

▼ Creating an index

```
1 f = ['FB', '2001-08-02', 90, 3.2]
2 fs = pd.Series(f, index = ['name', 'date', 'shares', 'price'])
```

```
1 fs
```

```
name      FB
date      2001-08-02
shares    90
price     3.2
dtype: object
```

▼ Accessing elements

- Use named index, or position
- Use slices, sublists

```
1 fs['shares']
```

```
90
```

```
1 fs[0]
```

```
'FB'
```

```
1 fs[0:2]
```

```
name      FB
date      2001-08-02
dtype: object
```

```
1 fs[[0,2]]
```

```
name      FB
shares    90
dtype: object
```

```
1 fs['name':'price']
```

```
name      FB
date      2001-08-02
shares    90
price     3.2
dtype: object
```

▼ Data frames

- A table is a sequence of columns
- A data frame is a sequence of series

```
1 data1 = [ ['AA', 'IBM', 'GOOG'],
2           ['2001-12-01', '2012-02-10', '2010-04-09'],
3           [100, 30, 90],
4           [12.3, 10.3, 32.2]
5           ]
6 df1 = pd.DataFrame(data1)
```

```
1 df1
```

	0	1	2
0	AA	IBM	GOOG
1	2001-12-01	2012-02-10	2010-04-09
2	100	30	90
3	12.3	10.3	32.2

```
1 data2 = {'name' : ['AA', 'IBM', 'GOOG'],
```

```
2     'date' : ['2001-12-01', '2012-02-10', '2010-04-09'],
3     'shares' : [100, 30, 90],
4     'price' : [12.3, 10.3, 32.2]
5 }
6 df2 = pd.DataFrame(data2)
```

```
1 df2
```

	name	date	shares	price
0	AA	2001-12-01	100	12.3
1	IBM	2012-02-10	30	10.3
2	GOOG	2010-04-09	90	32.2