Agenda

- Numpy
- Pandas
- Lab

Introduction



Create a new notebook for your code-along:

If necessary, from our lab03 submission directory type

ipython notebook

from the IPython Dashboard open a new notebook. Change the title to "Numpy and Pandas"

Introduction to Numpy

- Overview
- ndarray
- · Indexing and Slicing

More info: http://wiki.scipy.org/Tentative NumPy Tutorial)

Numpy Overview

- Why Python for Data? Numpy brings decades of C math into Python
- Numpy provides wrapper for extensive C/C++/Fortran codebases for data analysis and analytical functionality
- NDAarray allows easy vectorized math and broadcasting (functions on vector elements of different shapes)
- · The computational foundation fro

```
In []: from numpy import * #Load all the numpy packages
```

Note: "import as"

import * loads all sub module and is wasteful of memory when incorporated into deployed code. We use it here by example -- and its fine to use for learning purposes, legibility, etc.

As we'll see later, the the convention is to use:

```
import numpy as np
```

And then to specifically call needed methods:

```
In [2]: import numpy as np

In [5]: zeros?

Object `zeros` not found.

In [6]: np.zeros?

Type: builtin_function_or_method
String Form: <built-in function zeros>
Docstring:
zeros(shape, dtype=float, order='C')

Return a new array of given shape and type, filled with zeros.

Parameters

shape: int or sequence of ints
```

Creating ndarrays

An array object represents a multidimensional, homogeneous array of fixed-size items.

```
In []: # Creating arrays
    a = zeros((3))
    b = ones((2,3))
    c = random.randint(1,10,(2,3,4))
    d = arange(0,11,1)
```

What are these functions?

In []: | >>> c = a-b | >>> c

```
arange?
```

```
In []: # Note the way each array is printed:
    a,b,c,d

In []: ## Arithmetic in arrays is element wise

In []: >>> a = array( [20,30,40,50] )
    >>> b = arange( 4 )
    >>> b
```

```
In []: >>> b**2
```

Indexing, Slicing and Iterating

```
In []: # one-dimensional arrays work like lists:
    a = arange(10)**2

In []: a

In []: a[2:5]

In []: # Multidimensional arrays use tuples with commas for indexing
    # with (row, column) conventions beginning, as always in Python, from 0

In []: b = random.randint(1,100,(4,4))

In []: b

In []: # Guess the output
    print(b[2,3])
    print(b[0,0])

In []: b[0:3,1],b[:,1]
In []: b[1:3,:]
```

Pandas

- · Object Creation
- · Viewing data
- Selection
- · Missing data
- Grouping
- Reshaping
- · Time series
- Plotting
- i/o

Pandas Overview

Source: pandas.pydata.org (http://pandas.pydata.org/pandas-docs/stable/10min.html)

```
In []:
       import pandas as pd
       import numpy as np
       import matplotlib.pyplot as plt
In []: | dates = pd.date_range('20140101',periods=6)
In []: df = pd.DataFrame(np.random.randn(6,4),index=dates,columns=list('ABCD'))
In []: # Index, columns, underlying numpy data
       df
In []: | df2 = pd.DataFrame({ 'A' : 1.,
                                 'B' : pd.Timestamp('20130102'),
                                 'C' : pd.Series(1,index=list(range(4)),dtype='float3
       2'),
                                 'D' : np.array([3] * 4,dtype='int32'),
                                 'E' : 'foo' })
       df2
In []: # With specific dtypes
       df2.dtypes
```

Viewing Data

```
In []: df.head()
In []: df.tail()
In []: df.index
In []: df.describe()
In []: df.sort(columns='B')
```

Selection

In []: |df

```
In []: |df['A']
   In []: |df[0:3]
   In []:
          # By label
          df.loc[dates[0]]
  In []: | # multi-axis by label
          df.loc[:,['A','B']]
   In []: # Date Range
          df.loc['20140102':'20140104',['B']]
   In []: # Fast access to scalar
          df.at[dates[1],'B']
   In []: \# iloc provides integer locations similar to np style
Boolean Indexing
   In []: |df[df.A < 0] # Basically a 'where' operation</pre>
Setting
   In []: df_posA = df.copy() # Without "copy" it would act on the dataset
          df_posA[df_posA.A < 0] = -1*df_posA
   In []:
          df posA
  In []: | #Setting new column aligns data by index
          s1 = pd.Series([1,2,3,4,5,6],index=pd.date_range('20140102',periods=6))
   In []: s1
  In []: |df['F'] = s1
```

Missing Data

Operations

```
In []: df.describe()
In []: df.mean(),df.mean(1) # Operation on two different axes
```

Applying functions

```
In []: df
In []: df.apply(np.cumsum)
In []: df.apply(lambda x: x.max() - x.min())
In []: # Built in string methods
s = pd.Series(['A', 'B', 'C', 'Aaba', 'Baca', np.nan, 'CABA', 'dog', 'cat'])
s.str.lower()
```

Merge

```
In []: random.randn(10,4)
In []: #Concatenating pandas objects together
```

```
df = pd.DataFrame(np.random.randn(10,4))
df

In []: # Break it into pieces
    pieces = [df[:3], df[3:7], df[7:]]

In []: pd.concat(pieces)

In []: # Also can "Join" and "Append"
```

Grouping

Reshaping

```
In []: # You can also stack or unstack levels
In []: a = df.groupby(['A','B']).sum()
In []: # Pivot Tables
    pd.pivot_table(df,values=['C','D'],rows=['A'],cols=['B'])
```

Time Series

```
In []: import pandas as pd
import numpy as np
In []: # 100 Seconds starting on January 1st
```

```
rng = pd.date_range('1/1/2014', periods=100, freq='S')

In []: # Give each second a random value
    ts = pd.Series(np.random.randint(0, 500, len(rng)), index=rng)

In []: ts

In []: # Built in resampling
    ts.resample('1Min',how='mean') # Resample secondly to 1Minutely

In []: # Many additional time series features
    ts. #use tab
```

Plotting

```
In []: |ts.plot()
In []:
       def randwalk(startdate,points):
           ts = pd.Series(np.random.randn(points), index=pd.date_range(startdate, pe
       riods=points))
           ts=ts.cumsum()
           ts.plot()
           return(ts)
In []: # Using pandas to make a simple random walker by repeatedly running:
       a=randwalk('1/1/2012',1000)
In []:
       # Pandas plot function will print with labels as default
In []: | df = pd.DataFrame(np.random.randn(100, 4), index=ts.index,columns=['A', 'B',
       'C', 'D'])
       df = df.cumsum()
       plt.figure();df.plot();plt.legend(loc='best') #
```

1/0

I/O is straightforward with, for example, pd.read_csv or df.to_csv

The benefits of open source:

Let's look under x's in plt modules

Lab

Next Steps

Recommended Resources

Name	Description
Official Pandas Tutorials (http://pandas.pydata.org/pandas- docs/stable/tutorials.html)	Wes & Company's selection of tutorials and lectures
Julia Evans Pandas Cookbook (https://github.com/jvns/pandas-cookbook)	Great resource with examples from weather, bikes and 311 calls
Learn Pandas Tutorials (https://bitbucket.org/hrojas/learn-pandas)	A great series of Pandas tutorials from Dave Rojas
Research Computing Python Data PYNBs (https://github.com/ResearchComputing/Meetup-Fall- 2013/tree/master/python)	A super awesome set of python notebooks from a meetup-based course exclusively devoted to pandas