

# Hands-on Session Examples

## ProDy Examples

```
In [39]: from prody import parsePDB # we imported only a function that we need for
```

```
In [40]: parsePDB ?
```

Let's parse some structures:

```
In [41]: p38 = parsePDB('1p38')
```

```
@> PDB file is found in the local mirror (.\1p38.pdb).  
DEBUG:.prody:PDB file is found in the local mirror (.\1p38.pdb).  
@> 2962 atoms and 1 coordinate set(s) were parsed in 0.04s.  
DEBUG:.prody:2962 atoms and 1 coordinate set(s) were parsed in 0.04s.
```

```
In [42]: p38
```

```
Out[42]: <AtomGroup: 1p38 (2962 atoms)>
```

p38 is an AtomGroup object. You can think of it as a list of atoms, and indexing will work in the same way.

```
In [43]: l = range(10)
```

```
In [44]: l[0] # to get the first item in the list
```

```
Out[44]: 0
```

```
In [45]: p38[0] # to get the first atom in the atom group
```

```
Out[45]: <Atom: N from 1p38 (index 0)>
```

```
In [46]: n = p38[0]
```

```
In [47]: n
```

```
Out[47]: <Atom: N from 1p38 (index 0)>
```

## Dictionaries

There are also similarities between AtomGroup objects and dictionaries. Dictionaries map keys to values. In the following we will map some first names to last names:

```
In [48]: d = {}
```

```
In [49]: d ?
```

```
In [50]: d['ahmet']
```

```
-----
--
KeyError                                Traceback (most recent call
last)
<ipython-input-50-94b25eed165b> in <module>()
----> 1 d['ahmet']

KeyError: 'ahmet'
```

```
In [51]: d['ahmet'] = 'bakan'
```

```
In [52]: d
```

```
Out[52]: {'ahmet': 'bakan'}
```

```
In [53]: d['tim'] = 'lezon'
```

```
In [54]: d
```

```
Out[54]: {'ahmet': 'bakan', 'tim': 'lezon'}
```

```
In [55]: d['ahmet']
```

```
Out[55]: 'bakan'
```

## AtomGroups - Chains

AtomGroup objects map one letter chain identifiers, e.g. 'A' to chain objects.

```
In [56]: p38.numAtoms()
```

```
Out[56]: 2962
```

```
In [57]: p38.numChains()
```

```
Out[57]: 1
```

```
In [58]: p38['A']
```

```
Out[58]: <Chain: A from lp38 (480 residues, 2962 atoms)>
```

```
In [59]: chA = p38['A']
```

```
In [60]: chA
```

```
Out[60]: <Chain: A from lp38 (480 residues, 2962 atoms)>
```

```
In [61]: len(p38) # length of AtomGroup is number atoms, since AtomGroup is like a list
```

```
Out[61]: 2962
```

```
In [62]: len(chA) # whereas length of a chain object is number of residues, since chain
```

```
Out[62]: 480
```

## Some calculations

```
In [63]: from prody import * # we now import everything
```

```
In [64]: calcGyradius(p38) # we calculated radius of gyration for the AtomGroup
```

```
Out[64]: 22.057752024921747
```

```
In [65]: calcGyradius(chA) # Chain object can also be an input to the same function
```

```
Out[65]: 22.057752024921747
```

```
In [66]: chA
```

```
Out[66]: <Chain: A from lp38 (480 residues, 2962 atoms)>
```

```
In [67]: p38
```

```
Out[67]: <AtomGroup: 1p38 (2962 atoms)>
```

```
In [68]: p38.numAtoms('water') # this gives us number of water atoms in the AtomGro
```

```
Out[68]: 129
```

```
In [69]: p38.numAtoms('protein') # this give number of protein atoms
```

```
Out[69]: 2833
```

```
In [70]: p38.select('protein') # we select protein atoms
```

```
Out[70]: <Selection: 'protein' from 1p38 (2833 atoms)>
```

```
In [71]: p38.protein # same selection simplified like this
```

```
Out[71]: <Selection: 'protein' from 1p38 (2833 atoms)>
```

```
In [72]: p38.protein == p38.select('protein') # they are the same thing
```

```
Out[72]: True
```

```
In [73]: calcGyradius(p38.protein) # this is what we would do if we wanted to perfo
```

```
Out[73]: 21.960840914729118
```

## Residues

```
In [74]: chA[10] # get the residue number 10
```

```
Out[74]: <Residue: ARG 10 from Chain A from 1p38 (11 atoms)>
```

```
In [75]: calcPsi(chA[10]) # a residue can be an input to calcPsi function
```

```
Out[75]: 147.49025666398765
```

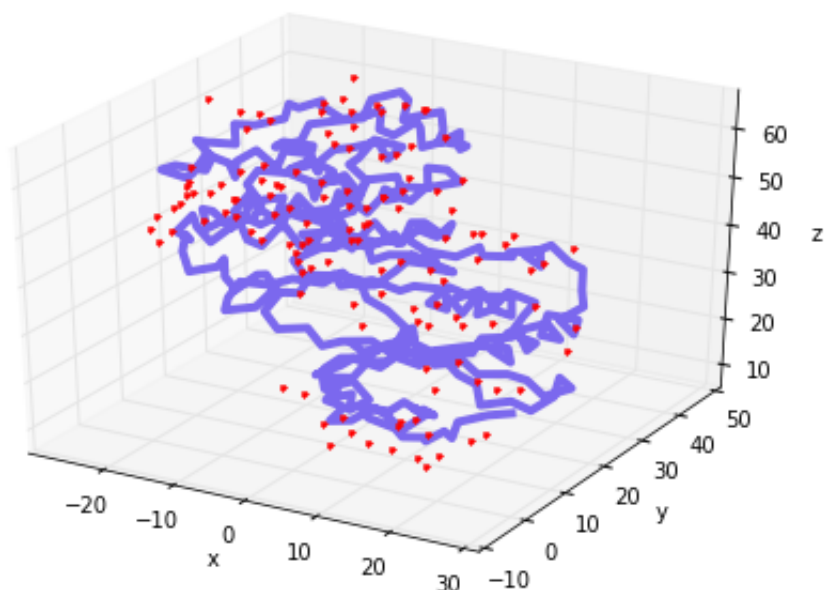
```
In [76]: calcPsi ?
```

## Show atoms

You can show atoms using the following function. For proteins only Calpha trace will be displayed. Heterogeneous atoms will be shown as points or dots

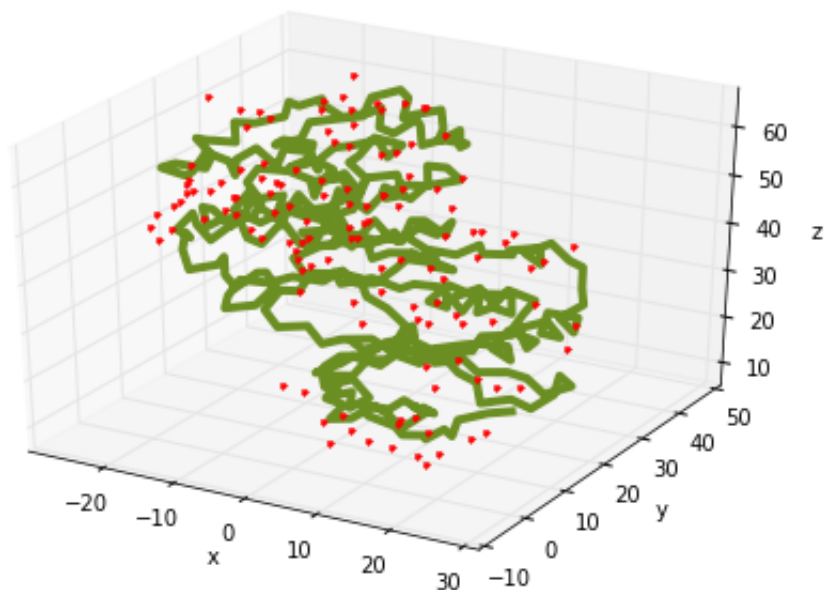
```
In [77]: showProtein(p38)
```

```
Out[77]: <mpl_toolkits.mplot3d.axes3d.Axes3D at 0x56a9ed0>
```



```
In [78]: showProtein(chA)
```

```
Out[78]: <mpl_toolkits.mplot3d.axes3d.Axes3D at 0x5b18170>
```



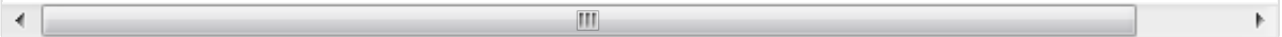
## Arrays and Efficiency

```
In [79]: l = range(10)
```

```
In [80]: array(l) # this give us an array
```

```
Out[80]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
In [81]: l = range(100000) # we will make a long list to show how efficient arrays
```



```
In [82]: len(l)
```

```
Out[82]: 100000
```

```
In [83]: a = array(l)
```

```
In [84]: len(a)
```

```
Out[84]: 100000
```

We will time the operation of summing up items in a list and array. For lists, we will use built-in Python function `sum`. For arrays, we will use array method `.sum`. Array methods are C code working on defined types, and they can be much more faster than pure Python equivalents.

```
In [85]: %timeit sum(l)
```

```
10 loops, best of 3: 23.4 ms per loop
```

This operation took 23400 microseconds on my laptop.

```
In [86]: a.sum()
```

```
Out[86]: 704982704
```

```
In [87]: a.max()
```

```
Out[87]: 99999
```

```
In [88]: %timeit a.sum()
```

```
10000 loops, best of 3: 128 us per loop
```

This operation took 124 micro seconds

```
In [89]: 23400/125
```

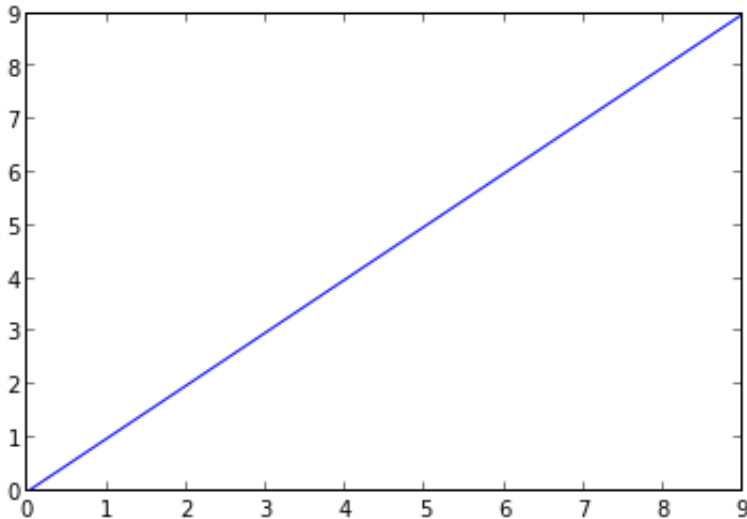
```
Out[89]: 187
```

Array summation is 187 times faster. ProDy uses NumPy arrays to store and modify data in structure file, so most operations will be as fast as it would be using C code developed for the same purpose.

## Plotting example

```
In [90]: plot( range(10) ) # simple plot
```

```
Out[90]: [<matplotlib.lines.Line2D at 0x5926ef0>]
```



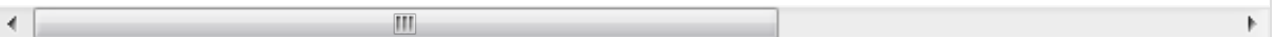
```
In [91]: random.random(10) # this returns random numbers from uniform distribution
```

```
Out[91]: array([ 0.64976994,  0.36475811,  0.01070796,  0.21786143,  0.91924073,
                0.27599671,  0.6306186 ,  0.59448841,  0.90071934,  0.83750927])
```

```
In [92]: random.random((3,3))
```

```
Out[92]: array([[ 0.4052089 ,  0.16822682,  0.18026583],
                [ 0.60029805,  0.57140157,  0.37823275],
                [ 0.32850192,  0.62418365,  0.1196399 ]])
```

```
In [93]: random_number = normal(10, 2, 1000) # will return 1000 random numbers from
```



```
In [1]: hist(random number) # this show a distribution of numbers
```

```

-----
--
NameError                                Traceback (most recent call
last)
<ipython-input-1-4fc3bba91e20> in <module>()
----> 1 hist(random_number) # this show a distribution of numbers

NameError: name 'random_number' is not defined

```

## A crazy example

In this example, we will download an image from the internet and show it using Matplotlib

```

In [2]: from urllib import urlopen # for reading online documents
        from matplotlib import image # for reading image data

```

```

In [3]: url_handle = urlopen('http://www.csb.pitt.edu/ProDy/_static/logo.png') # w

```

```

In [4]: img_data = image.imread(url_handle) # this will read image data and conver

```

```

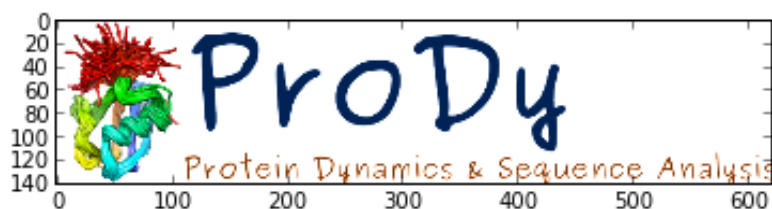
In [7]: imshow(img_data) # You see the image below

```

```

Out[7]: <matplotlib.image.AxesImage at 0x555e670>

```



You can do much more with Matplotlib. Take a look at its gallery of plots: <http://matplotlib.org/gallery.html>  
 You can pick something similar to that you want to plot and then get the code example that was used to plot it.