

# High Performance Python

Week 1: Profiling

# The Strengths of Python

- Fast to Write
- Easy to Read
- Batteries Included

# The Weaknesses of Python

- Slow to Execute
- Global Interpreter Lock (GIL)

# How to Make Python Fast

- Array Computations
- Offloading Functions to C or Fortran
- Multiple Processors

**Warning:** Don't optimize before the code works

All of these techniques will make your code more difficult for yourself and others to read and understand.

We also should be (but aren't) unit testing

# The Problem!

- How do we go about finding the slow parts of our code so we can speed them up?
- Once we find a slow part of the code, how do we make sure we change it so that it is faster?

# The Answer

- Time functions using the built in time module

(good)

- Time functions using ipython's magic timeit functionality (or using the built in timeit module)

(better)

- Profile the codes execution time using a line profiler

(best)

# Example

```
def smooth1( ary, M ):
    #done in place
    for i in range(M):
        ary[1:-1] += ary[:-2] + ary[2:]
        ary /= 3
        ary[0] = 2*ary[0] + ary[1]/3
        ary[-1] = 2*ary[-1] + ary[-2]/3
```

```
def smooth2( ary, M ):
    #done in place
    for i in range(M):
        ary += np.append( ary[0] ,ary[:-1]) + \
               np.append( ary[1:],ary[-1] )
        ary /= 3
```

# Using time.time

```
def smooth1( ary,M ):
    #done in place
    startTime = time.time()
    for i in range(M):
        ary[1:-1] += ary[:-2] + ary[2:]
        ary /= 3
        ary[0] = 2*ary[0] + ary[1]/3
        ary[-1] = 2*ary[-1] + ary[-2]/3
    print time.time()-startTime
```

```
def smooth2( ary, M):
    #done in place
    startTime = time.time()
    for i in range(M):
        ary += np.append( ary[0] ,ary[:-1]) + \
               np.append( ary[1:],ary[-1] )
        ary /= 3
    print time.time()-startTime
```



# iPython Shell

# Using timeit in ipython

```
In [93]: %timeit( smooth.smooth1(x.copy()),M) )
```

1000 loops, best of 3: 1.78 ms per loop

```
In [94]: %timeit( smooth.smooth2(x.copy()),M) )
```

100 loops, best of 3: 2.18 ms per loop

# Profiling

- For more complex problems, `time.time` and `timeit` quickly become limiting.
- This is where using a code profiler comes in.

# cProfile

- Python has cProfile built in

```
$ python -m cProfile -s cumulative smooth.py | less
```

```
0.000411987304688
```

```
0.000672101974487
```

```
11157 function calls (10968 primitive calls) in 0.068 seconds
```

```
Ordered by: cumulative time
```

ncalls	tottime	percall	cumtime	percall	filename:lineno(function)
1	0.000	0.000	0.069	0.069	smooth.py:2(<module>)
1	0.001	0.001	0.067	0.067	__init__.py:106(<module>)
1	0.000	0.000	0.035	0.035	add_newdocs.py:9(<module>)
2	0.002	0.001	0.028	0.014	__init__.py:1(<module>)
1	0.001	0.001	0.023	0.023	__init__.py:15(<module>)
2	0.004	0.002	0.019	0.010	__init__.py:2(<module>)

More info can be found at <http://docs.python.org/2/library/profile.html>

# cProfile Switches

Argument	Meaning
'calls'	Call count
'cumulative', 'cumtime'	Cumulative time
'file', 'filename', 'module'	File name
'ncalls'	Call count
'pcalls'	Primitive call count
'line'	Line number
'name'	Function name
'nfl'	Name/file/name
'stdname'	Standard name
'time', 'tottime'	Internal time

# A Better Way (line\_profiler)

Step 1: Add decorators

```
@profile  
def smooth1( ary,M ):
```

Step 2: Profile

```
$ kernprof.py -l smooth.py  
Wrote profile results to smooth.py.lprof  
$ python -m line_profiler smooth.py.lprof | less
```

# A Better Way (line\_profiler)

Timer unit: 1e-06 s

File: smooth.py

Function: smooth1 at line 5

Total time: 0.00058 s

Line #	Hits	Time	Per Hit	% Time	Line Contents
5					@profile
6					def smooth1( ary,M ):
7					#done in place
8	1	7	7.0	1.2	startTime = time
9	17	15	0.9	2.6	for i in range(M
10	16	184	11.5	31.7	ary[1:-1
11	16	188	11.8	32.4	ary
12	16	86	5.4	14.8	ary[0]
13	16	58	3.6	10.0	ary[-1]
14	1	42	42.0	7.2	print time.time(

# Exercise, cluster

- cluster.py works, but is slow
- Run line\_profiler on cluster.py
- Find any bottle necks, and fix them

cluster\_solution.py is my solution to the problem

We'll use this script again to further optimize it