Lecture 23

Code Profiling and Debugging

(Some material adapted from Chris Simmons)

Tuesday, November 24th 2020

Performance Analysis

- Want a solution quickly
 - Some simulations are making predictions about time-sensitive things (hurricanes, disaster relief, etc).
- Want to resolve currently intractable scientific problems
 - Scientific discovery
- Want efficient code
 - This has a monetary impact
- Predictive science
 - Machine learning, uncertainty quantification, optimization, ...
 - These require running many code executions

Why Profile Code?

- You want your program to run *faster*
- Usually, a program has one or more bottlenecks leading to slow execution time
- You could diagnose these manually by inserting timers into different parts of your code
 - This would be a nightmare!
- It would be awfully nice if there were a way to automatically diagnose how fast
 / slow different parts of the code are running
 - This is where code profilers come in

What are the Options?

- For common compiled languages (C, C++, Fortran), have a look at <u>gprof</u>
- In Python, you have <u>cProfile</u> and profile
- cProfile is a good choice as a default option with low overhead



import cProfile

def convert(sentence):
 return sentence.split()

cProfile.run('convert("A very fine sentence.")')

5 function calls in 0.000 seconds

Ordered by: standard name

ncalls	tottime	percall	cumtime	percall	filename:lineno(function)
1	0.000	0.000	0.000	0.000	<stdin>:1(convert)</stdin>
1	0.000	0.000	0.000	0.000	<string>:1(<module>)</module></string>
1	0.000	0.000	0.000	0.000	{built-in method builtins.exec}
1	0.000	0.000	0.000	0.000	<pre>{method 'disable' of '_lsprof.Profiler' objects}</pre>
1	0.000	0.000	0.000	0.000	<pre>{method 'split' of 'str' objects}</pre>

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1	0.000	0.000	0.000	0.000	{built-in method builtins.exec}
1	0.000	0.000	0.000	0.000	<pre>{method 'disable' of '_lsprof.Profiler' objects}</pre>
1	0.000	0.000	0.000	0.000	<pre>{method 'split' of 'str' objects}</pre>

There were five total function calls. It took 0.000 seconds (but not really).

5 function calls in 0.000 seconds

Ordered by: standard name

ncalls	tottime	percall	cumtime	percall	<pre>filename:lineno(function)</pre>
1	0.000	0.000	0.000	0.000	<stdin>:1(convert)</stdin>
1	0.000	0.000	0.000	0.000	<string>:1(<module>)</module></string>
1	0.000	0.000	0.000	0.000	<pre>{built-in method builtins.exec}</pre>
1	0.000	0.000	0.000	0.000	<pre>{method 'disable' of '_lsprof.Profiler' objects}</pre>
1	0.000	0.000	0.000	0.000	<pre>{method 'split' of 'str' objects}</pre>

Output is sorted by text string in the last column

5 function calls in 0.000 seconds

Ordered by: standard name

<mark>ncalls</mark>	tottime	percall	cumtime	percall	filename:lineno(function)
1	0.000	0.000	0.000	0.000	<stdin>:1(convert)</stdin>
1	0.000	0.000	0.000	0.000	<string>:1(<module>)</module></string>
1	0.000	0.000	0.000	0.000	{built-in method builtins.exec}
1	0.000	0.000	0.000	0.000	<pre>{method 'disable' of '_lsprof.Profiler' objects}</pre>
1	0.000	0.000	0.000	0.000	<pre>{method 'split' of 'str' objects}</pre>

ncalls: The number of times that function was called.

5 function calls in 0.000 seconds

Ordered by: standard name

ncalls	<mark>tottime</mark>	percall	cumtime	percall	filename:lineno(function)
1	0.000	0.000	0.000	0.000	<stdin>:1(convert)</stdin>
1	0.000	0.000	0.000	0.000	<string>:1(<module>)</module></string>
1	0.000	0.000	0.000	0.000	{built-in method builtins.exec}
1	0.000	0.000	0.000	0.000	<pre>{method 'disable' of '_lsprof.Profiler' objects}</pre>
1	0.000	0.000	0.000	0.000	<pre>{method 'split' of 'str' objects}</pre>

tottime: The total time spent in that function.

This does not include time in subfunctions.

5 function calls in 0.000 seconds

Ordered by: standard name

ncalls	tottime	<mark>percall</mark>	cumtime	percall	filename:lineno(function)
1	0.000	0.000	0.000	0.000	<stdin>:1(convert)</stdin>
1	0.000	0.000	0.000	0.000	<string>:1(<module>)</module></string>
1	0.000	0.000	0.000	0.000	{built-in method builtins.exec}
1	0.000	0.000	0.000	0.000	<pre>{method 'disable' of '_lsprof.Profiler' objects}</pre>
1	0.000	0.000	0.000	0.000	<pre>{method 'split' of 'str' objects}</pre>

percall: Time per call (tottime / ncalls).

5 function calls in 0.000 seconds

Ordered by: standard name

ncalls	tottime	percall	<mark>cumtime</mark>	percall	filename:lineno(function)
1	0.000	0.000	0.000	0.000	<stdin>:1(convert)</stdin>
1	0.000	0.000	0.000	0.000	<string>:1(<module>)</module></string>
1	0.000	0.000	0.000	0.000	{built-in method builtins.exec}
1	0.000	0.000	0.000	0.000	<pre>{method 'disable' of '_lsprof.Profiler' objects}</pre>
1	0.000	0.000	0.000	0.000	<pre>{method 'split' of 'str' objects}</pre>

cumtime: Cumulative time spent in this function and all of its subfunctions.

5 function calls in 0.000 seconds

Ordered by: standard name

ncalls	tottime	percall	cumtime	<mark>percall</mark>	filename:lineno(function)
1	0.000	0.000	0.000	0.000	<stdin>:1(convert)</stdin>
1	0.000	0.000	0.000	0.000	<string>:1(<module>)</module></string>
1	0.000	0.000	0.000	0.000	{built-in method builtins.exec}
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1	0.000	0.000	0.000	0.000	<pre>{method 'split' of 'str' objects}</pre>

percall: cumtime / primitive calls.

A primitive call is a function call that was not induced by recursion. In our example, everything is a primitive call.

5 function calls in 0.000 seconds

Ordered by: standard name

ncalls	tottime	percall	cumtime	percall	<pre>filename:lineno(function)</pre>
1	0.000	0.000	0.000	0.000	<stdin>:1(convert)</stdin>
1	0.000	0.000	0.000	0.000	<string>:1(<module>)</module></string>
1	0.000	0.000	0.000	0.000	{built-in method builtins.exec}
1	0.000	0.000	0.000	0.000	<pre>{method 'disable' of '_lsprof.Profiler' objects}</pre>
1	0.000	0.000	0.000	0.000	<pre>{method 'split' of 'str' objects}</pre>

filename:lineno(function): The data of the function.

Writing and reading performance stats

import pstats

cProfile.run('convert("A very fine sentence.")', 'pstats')

```
p = pstats.Stats('pstats')
```

p.print_stats()

```
Demo 2
```

import numpy as np

```
import cProfile
def f(x):
    return x - np.exp(-2.0 * np.sin(4.0*x) * np.sin(4.0*x))
def dfdx(x):
    return 1.0 + 16.0 * np.exp(-2.0 * np.sin(4.0*x) * np.sin(4.0*x)) * np.sin(4.0*x) * np.cos(4.0*x)
def dfdx_h(x, epsilon):
    return (f(x + epsilon) - f(x)) / epsilon
def main():
    # Start Newton algorithm
    xk = 0.0 # Initial guess
    tol = 1.0e-08 # Some tolerance
    max_it = 100 # Just stop if a root isn't found after 100 iterations
    h = 1.0e-09
    root = None # Initialize root
    for k in range(max_it):
        delta_xk = -f(xk) / dfdx(xk) # Update Delta x_{k}
        if (abs(delta_xk) <= tol): # Stop iteration if solution found
            root = xk + delta_xk
            print("Found root at x = \{0:17.16f\} after \{1\} iterations.".format(root, k+1))
            break
        print("At iteration {0}, Delta x = {1:17.16f}".format(k+1. delta_xk))
        xk += delta_xk # Update xk
```

cProfile.run('main()')

Breakout Room (10 minutes)

- Change the script to use the finite difference derivative in the Newton solver
- Run the profiler
- Did anything change?
 - Results?
 - Timing?
 - Function calls?

Debugging

- There is no bug-free code
 - You will introduce bugs
 - Other members of the community will introduce bugs
 - Bugs even live in commercial codes
- Bugs can:
 - Prevent a code from running at all
 - Prevent a code from running well
 - Lead to incorrect results and predictions
- Good debugging skills will make you a more efficient and confident programmer

Defensive Programming

- Check function return codes for errors (more useful in C, C++, Fortran)
- Check input values
 - We discussed this in our testing unit
 - Check "impossible" values
- Write out the control parameters to a file
 - This helps you keep track of your runs
 - Also helps for experimental repeatability
- Check for "non-physical" results
 - This applies beyond the physical sciences
 - For example, for negative chemical concentrations
 - Or check for negative number of people
- Employ the techniques we've discussed throughout the semester
 - Write regression tests, use version control, write modular code, document, error checking

Some common indicators of bugs

- Build errors (for compiled code) 😴
- Improper memory reads / writes (again, usually met in C / C++) (2)
- Illegal operations (division by zero, etc)
- Infinite loops 😬
- I/O errors 😳
- Algorithmic errors 🙀
- Poor performance 😭

The Debugging Process

- Start with defensive programming
 - You will still get bugs, just not nearly as many
- Basic steps:
 - Determine that there is a bug somewhere
 - Isolate the source of the bug (a.k.a. find the bug)
 - Identify the cause of the bug (how did this happen?)
 - Determine a fix for the bug (how do we remove this bug?)
 - Fix the bug and test
- These steps are not always so easy
- A debugger can help

Debuggers

- Command line debuggers are tools to aid in diagnosing problems
- The GNU Debugger Project
 - Very powerful
 - Works with many languages (not Python)
 - "allows you to see what is going on `inside' another program while it executes"
- pdb The Python Debugger
- These debuggers are a front-end for the application
 - Step though the code and examine: variables, arrays, functions, etc
- Have the opportunity to investigate the run-time behavior of the application

Debugging: Important Commands and Concepts

- Show program backtraces
 - The calling history up to the current point
- Set breakpoints
- Display values of individual names
- Set new values
- Step through the program

Breakpoints

- A breakpoint is a pseudo-instruction that you can insert at any place during a debugging session
- The debugger will interpret the breakpoint
- The program execution hits a breakpoint, the debugger will pause the program so you can:
 - Inspect names
 - Set and / or clear breakpoints
 - Continue execution
- There are also conditional breakpoints
 - Program pauses only if the breakpoint's condition holds
 - e.g. an expression is true, the breakpoint has been crossed "N" times, an expression changed its value

pdb **demo**

>>> import pdb # import pdb

- >>> import script # import our module
- >>> pdb.run('script.main()') # attach debugger

Getting Familiar

>>> pdb.run('script.main()') > <string>(1)<module>() (Pdb) continue

- Attach debugger to main script
- Debugger enters module and pauses by default
- We type continue to just run through everything

Stepping Through

- (Pdb) s
- --Call--
- > script.py(14)main()
 -> def main():
- (Pdb) s
- > script.py(16)main()
- -> xk = 0.0 # Initial guess

- Take a step to the next line
- We're at the main function
- Take a step into the main function
- We're at the initial guess

Exploring

-> xk = 0.0 # Initial guess (Pdb) s

```
> script.py(17)main()
```

```
-> tol = 1.0e-08 # Some tolerance
(Pdb) p xk
0.0
```

- Execute line setting initial guess
- Go to next line
- Before executing, print out value of xk just to check

Getting our bearings

```
-> for k in range(max_it):
(Pdb) 1
17
             tol = 1.0e-08 # Some tolerance
18
             max_it = 100 # Just stop if a root isn't found after 100 iterations
19
             h = 1.0e-09
20
21
             root = None # Initialize root
22
     ->
               for k in range(max_it):
23
                 delta_xk = -f(xk) / dfdx(xk) # Update Delta x_{k}
24
                 if (abs(delta_xk) <= tol): # Stop iteration if solution found</pre>
25
                     root = xk + delta xk
                     print("Found root at x = \{0:17.16f\} after \{1\} iterations.".format(root, k+1))
26
27
                     break
```

- We stepped through a bunch of lines of code
- Let's list the lines around the current line
- We can see where we are and we can see a few lines above and below

Setting a breakpoint

- We're worried that our function fk isn't doing the right things
- Let's set a breakpoint at line 24
- Then we'll continue to that point
- Once we get to the breakpoint we can examine things more

```
(Pdb) b 24
Breakpoint 2 at script.py:24
(Pdb) continue
> script.py(24)main()
-> if (abs(delta_xk) <= tol): # Stop iteration if ...</pre>
```

More exploration

(Pdb) p delta_xk 1.0 (Pdb) p f(xk) -1.0 (Pdb) p dfdx(xk) 1.0

- Have a look at the step value
- Check the output of f ()
- Check the output of dfdx ()
- Things look fine so far!
- Let's put in another breakpoint

Going into the functions

(Pdb) b 23 Breakpoint 3 at script.py:23 (Pdb) continue At iteration 1, Delta x = 1.000000000000000 > script.py(23)main() -> delta_xk = -f(xk) / dfdx(xk) # Update Delta x_{k}

Inspecting the function

(Pdb) s --Call--> script.py(4)f() -> def f(x): (Pdb) s > script.py(5)f() -> return x - np.exp($-2.0 \times np.sin(4.0 \times x) \times np.sin(4.0 \times x)$) (Pdb) s --Return--> script.py(5)f()->0.6819351651504464 -> return x - np.exp($-2.0 \times np.sin(4.0 \times x) \times np.sin(4.0 \times x)$)

Other fun things: Exploring names

(Pdb) whatis f
Function f
(Pdb) whatis xk
<class 'numpy.float64'>

Other fun things: listing breakpoints

(Pdb) b Num Type Where Disp Enb breakpoint keep yes at script.py:24 breakpoint already hit 4 times breakpoint keep yes at script.py:24 2 3 breakpoint keep yes at script.py:23 breakpoint already hit 2 times

Other fun things: disabling breakpoints

(Pdb) disable 1 Disabled breakpoint 1 at script.py:24 (Pdb) b Disp Enb Num Type Where breakpoint keep <mark>no</mark> at script.py:24 1 breakpoint already hit 4 times breakpoint keep yes at script.py:24 2 breakpoint keep yes 3 at script.py:23 breakpoint already hit 2 times

Finishing up

(Pdb) continue At iteration 3, Delta x = 0.0782795784669513At iteration 4, Delta x = -0.0271256514726681At iteration 5, Delta x = -0.0012447280598220At iteration 6, Delta x = -0.0000049668585411Found root at x = 0.8560316824308374 after 7 iterations.

- In this example, we disable our breakpoints and just ran everything to the end
 - Easy because this is a little code
- Alternatively, when we were done, we could have just quit (q)

(Pdb) q >>>

Summary / Debrief

- The debugger can make debugging more fun!
- We only covered the most basic tasks today
- There are many other things you can do
- Here are some resources
 - pdb The Python Debugger
 - o <u>pdb</u> cheatsheet
 - <u>Python Call Graph</u> Pretty nifty way to visualize what your code is doing