



# 40.317 Lecture 4

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Slido Event #N290

# Agenda

- Speeding up Python code

# Speeding up Python Code: Motivation

Why are we studying this topic?

- To understand the steps we should follow to speed up code in *any* language.
- To present performance pitfalls and speedup techniques unique to Python.

# Speeding up Python Code

Best overall references:

- <http://pypy.org/performance.html>
- <http://earthpy.org/speed.html>

Next, we describe a general eight step approach.

# Speeding up Code in 8 Steps

1. Design, part 1: Ask yourself what you are really building (!).
  - E.g. if what you are building is a message-passing system, then pass messages already!

# Speeding up Code in 8 Steps

## 2. Design, part 2: Select your algorithms and data structures before you start coding.

- Examples:
  - The famous story from Gauss's childhood
  - The [3x3 magic square problem](#) on hackerrank.com
- An understanding of complexity (c.f. Lecture 2) will really help you here
- Using less space often translates into taking less time

# Speeding up Code in 8 Steps

3. Do not attempt to speed up your code unless you have a clear justification.
  - “Premature optimisation is the root of all evil.” —Donald Knuth
4. Before attempting any speedups, write a regression test which is so complete you can modify your code with confidence.
  - Reference and recommended packages for writing tests: <http://docs.python-guide.org/en/latest/writing/tests/>

# Speeding up Code in 8 Steps

5. Use a [\(statistical\) profiler](#) to determine exactly where the actual performance problem resides.
  - It must be very non-intrusive to avoid distorting the statistics it's gathering.
  - The owners of PyPy recommend [vmprof](#):
    - Supports MacOS X, Windows, and Linux
    - Supports multi-threaded applications
    - Lets you profile only a portion of your code
    - Install via `pip install vmprof` (on Windows, requires Microsoft Visual Studio Build Tools)



# Speeding up Code in 8 Steps

6. Look for opportunities to apply concurrent programming (discussed next week): can you divide up the task, run the sub-tasks on separate cores / PCs, and then quickly combine the results?

There are several popular packages to consider for this purpose, such as:

- [multiprocessing](#)
- [Dask](#)
- [Disco](#)

# Speeding up Code in 8 Steps

7. Look for opportunities to apply dynamic programming, often implemented via memoisation.

- A [second example](#) from hackerrank.com
- An example from Finance: translating one type of security ID to another
  - c.f. “FIGIs”, <https://www.openfigi.com>

# Speeding up Code in 8 Steps

8. Focus your remaining efforts on speeding up the insides of loops, e.g.:
  - Perform all validations / assertion checks in an earlier step
  - Minimise *explicit* branching
    - Make use of polymorphic class methods (discussed in about two weeks)
    - Create a dictionary which maps possible values (known only at runtime) to functions

# Python-Specific Tips

First, two meta-comments:

- “Python” is a *specification*. Its default / reference implementation, [CPython](#), happens to be the slowest of the most popular implementations.
- What Python speeds up the most is [your productivity](#) as a developer.
  - Example: `pythonic_radix_sort.py`

# Python-Specific Tips

- Python code run in CPython is [much slower](#) than a corresponding C version. My suggestions, from most general to least general:
  - Run your slowest programs in [PyPy](#), a “just-in-time” compiler, or push speed-critical code into C via [Cython](#)
  - Use [Numba](#), designed with the [NumPy](#) / [SciPy](#) stack in mind
  - Use [NumExpr](#), specifically for speeding up NumPy operations

# Python-Specific Tips

- Do not write your own versions of existing functions
- Do not use global variables inside loops
  - Python accesses global variables slowly compared to local variables
  - Either eliminate them or make local copies
- Avoid the use of “dots” inside a loop
  - Python resolves function addresses slowly
  - Use local variables to eliminate function resolution inside loops
- Lessen memory footprint with [generators](#)
  - Not the same thing as using less space

# Python-Specific Tips

- Eliminate as many explicit loops as possible. Replace them with either:
  - comprehensions, e.g.  
`{w.capitalize(): len(w) for w in words}`
  - or set operations, e.g. replace  
`in_common = set()`  
`for x in a:`  
    `for y in b:`  
        `if x == y:`  
            `in_common.add(x)`
  - with  
`in_common = set(a) & set(b)`

# Thank you

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