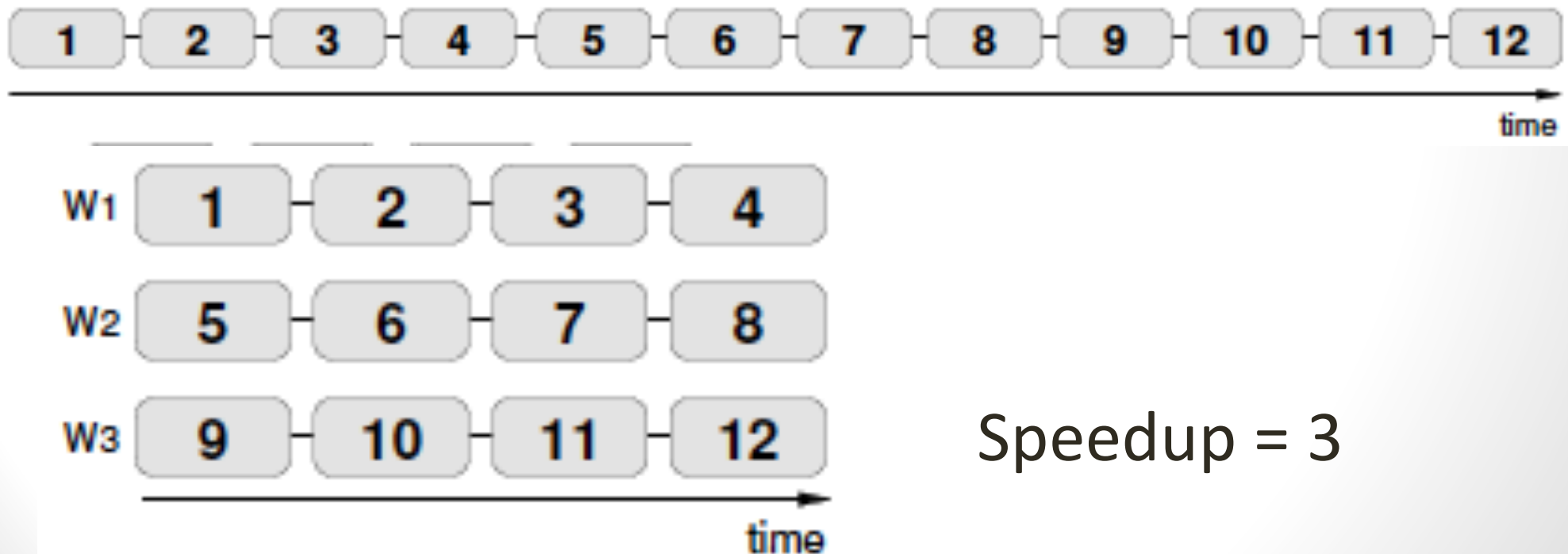


# Profiling and Timing your code

# Speedup Formula

$$\text{Speedup} = \frac{\text{Sequential execution time}}{\text{Parallel execution time}}$$



$$\text{Speedup} = 3$$

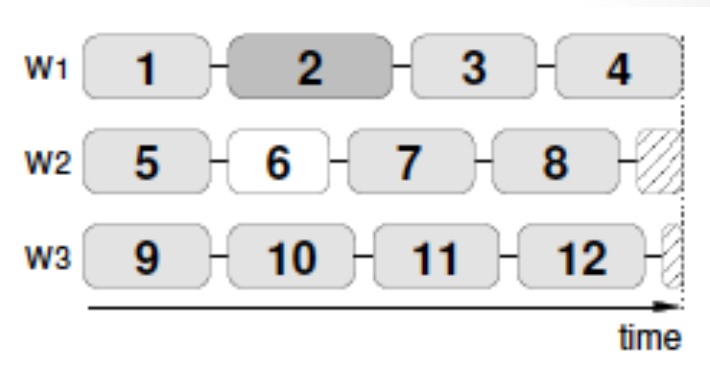
# Execution Time Components

- Inherently sequential computations:  $s(n)$
- Potentially parallel computations:  $p(n)$
- Communication operations:  $c(n, p)$
- Speedup expression:

$$S \leq \frac{s+p}{s(n)+p/N+c}$$

# Parallel Overhead

- Overhead because of
  - Startup time
  - Synchronizations
  - Communication
  - Overhead by libraries, compilers
  - Termination time
- Other barriers to perfect speedup
  - Not perfectly load balanced



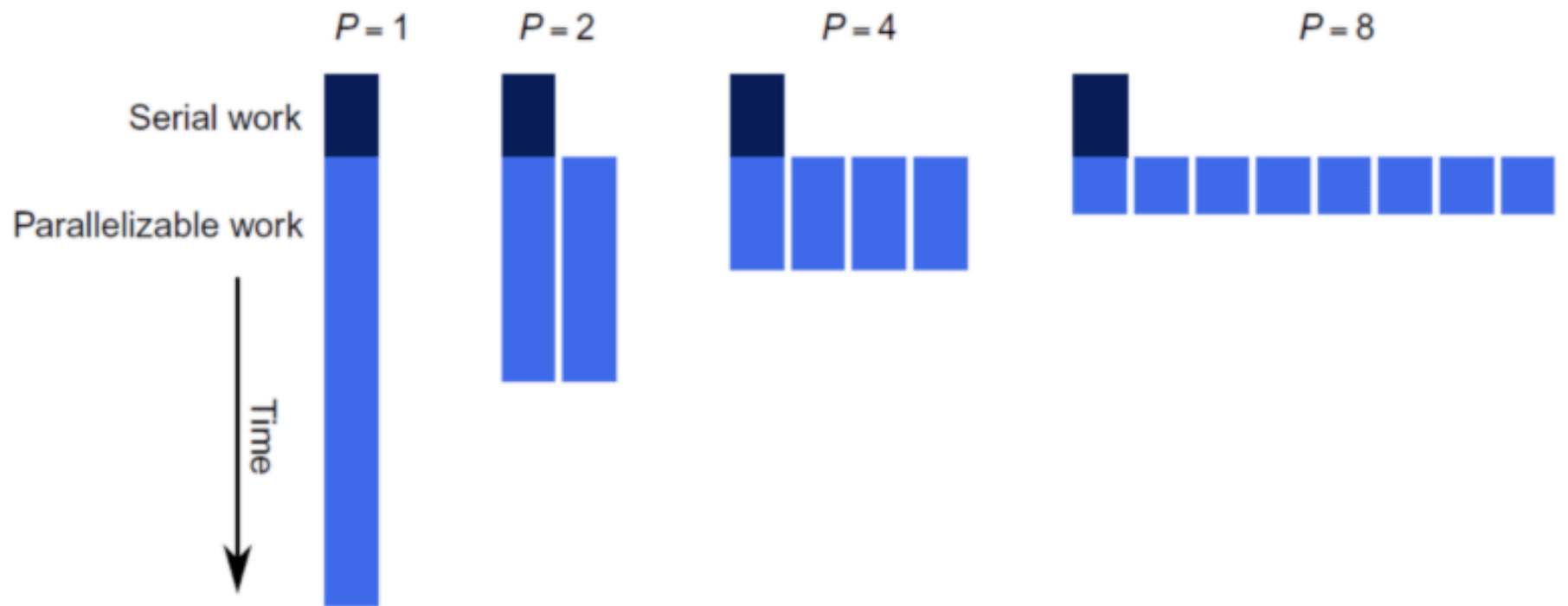
# Efficiency

$$\text{Efficiency} = \frac{\text{Sequential execution time}}{\text{Processors} \times \text{Parallel execution time}}$$

$$\text{Efficiency} = \frac{\text{Speedup}}{\text{Processors}}$$

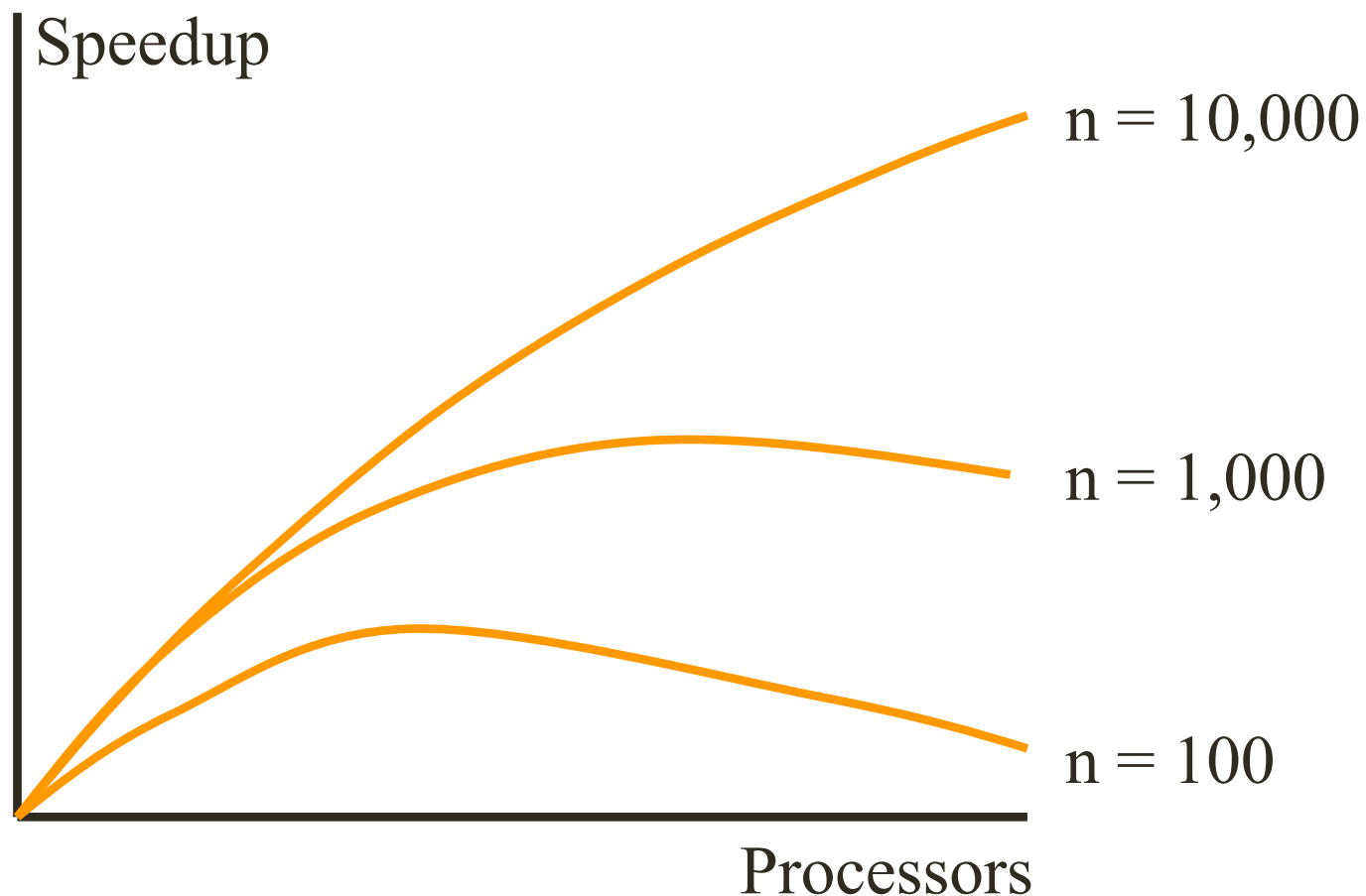
# Amdahl's law

- Speedup is limited by the non-parallelizable serial portion of the work



<http://www.drdoobs.com/parallel/amdahls-law-vs-gustafson-barsis-law/240162980?pgno=2>

# Illustration of Amdahl Effect



# Review of Amdahl's Law

- Treats problem size as a constant
- Shows how execution time decreases as number of processors increases
- **Strong scaling**
  - Problem size is fixed
  - Number of processor increases



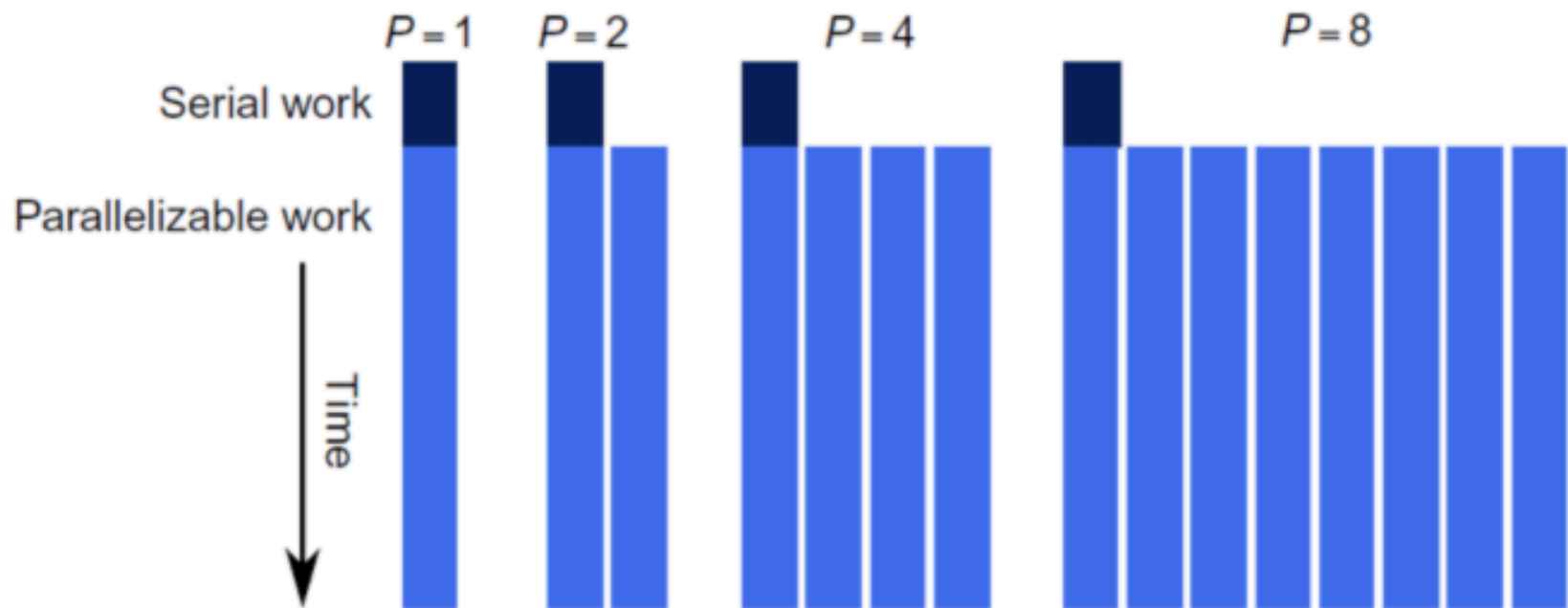
# Example

- Parallelizing Pi in python
- Benchmarking on one node
- Strong Scaling

# Another Perspective

- We often use faster computers to solve larger problem instances
- Let's treat time as a constant and allow problem size to increase with number of processors
- "...speedup should be measured by scaling the problem to the number of processors, not by fixing the problem size" – John Gustafson
- Weak scaling
  - Problem size per core stays constant
  - Overall program size increases with number of processors

# Gustafson-Barsis's Law



$$SS(N) = \frac{s+p*N}{s+p} = N + (1 - N)s$$

<http://www.drdobbs.com/parallel/amdahls-law-vs-gustafson-barsis-law/240162980?pgno=2>

# Timing your code

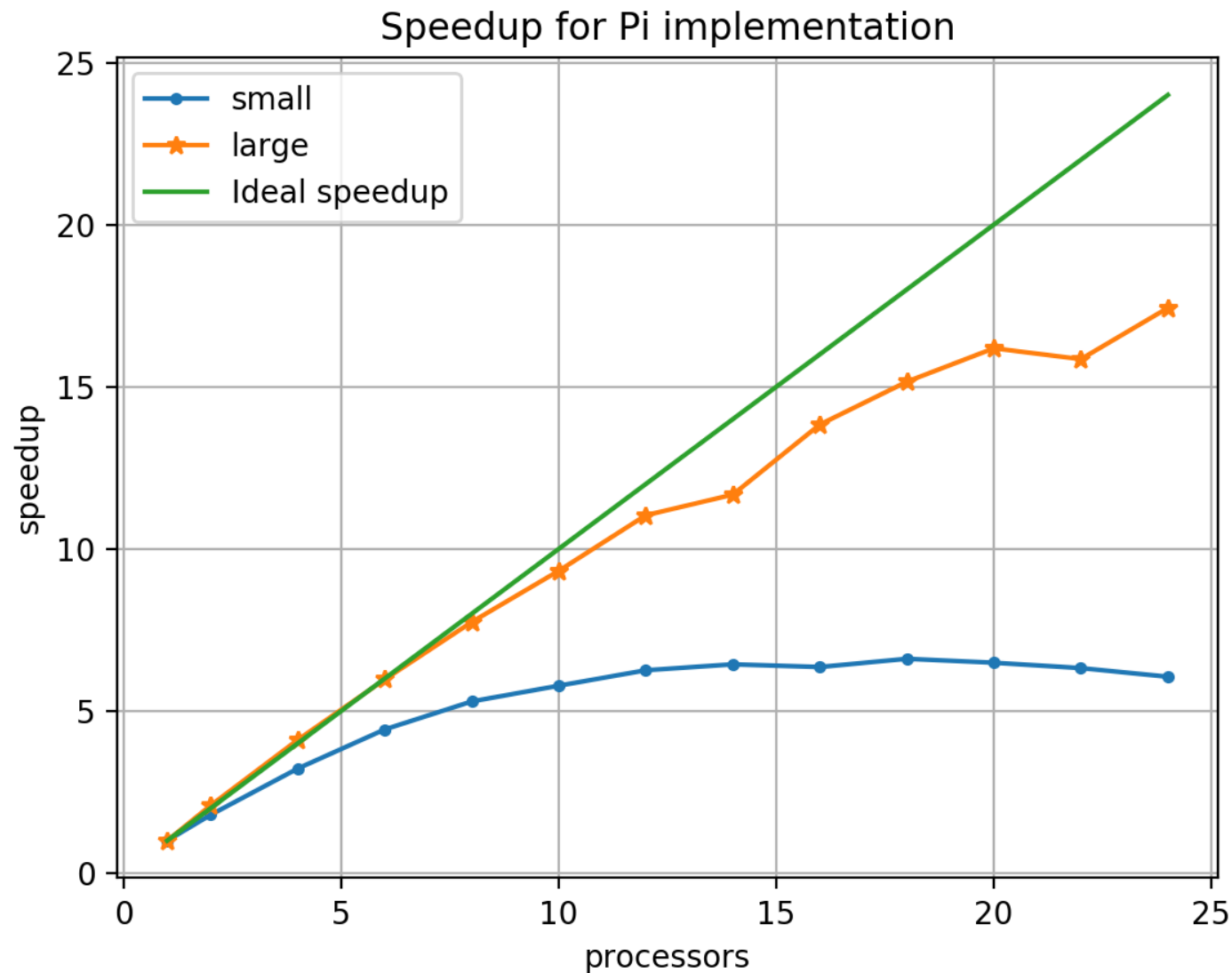
- Use the unix time command
- Provides time for
  - Real – wall-clock time
  - User
  - Sys
- Use real time / wall-clock time for parallel benchmarking
- User + Sys gives you CPU time
  - Can be larger than real time - multithreading

# Speedup and Parallel Efficiency

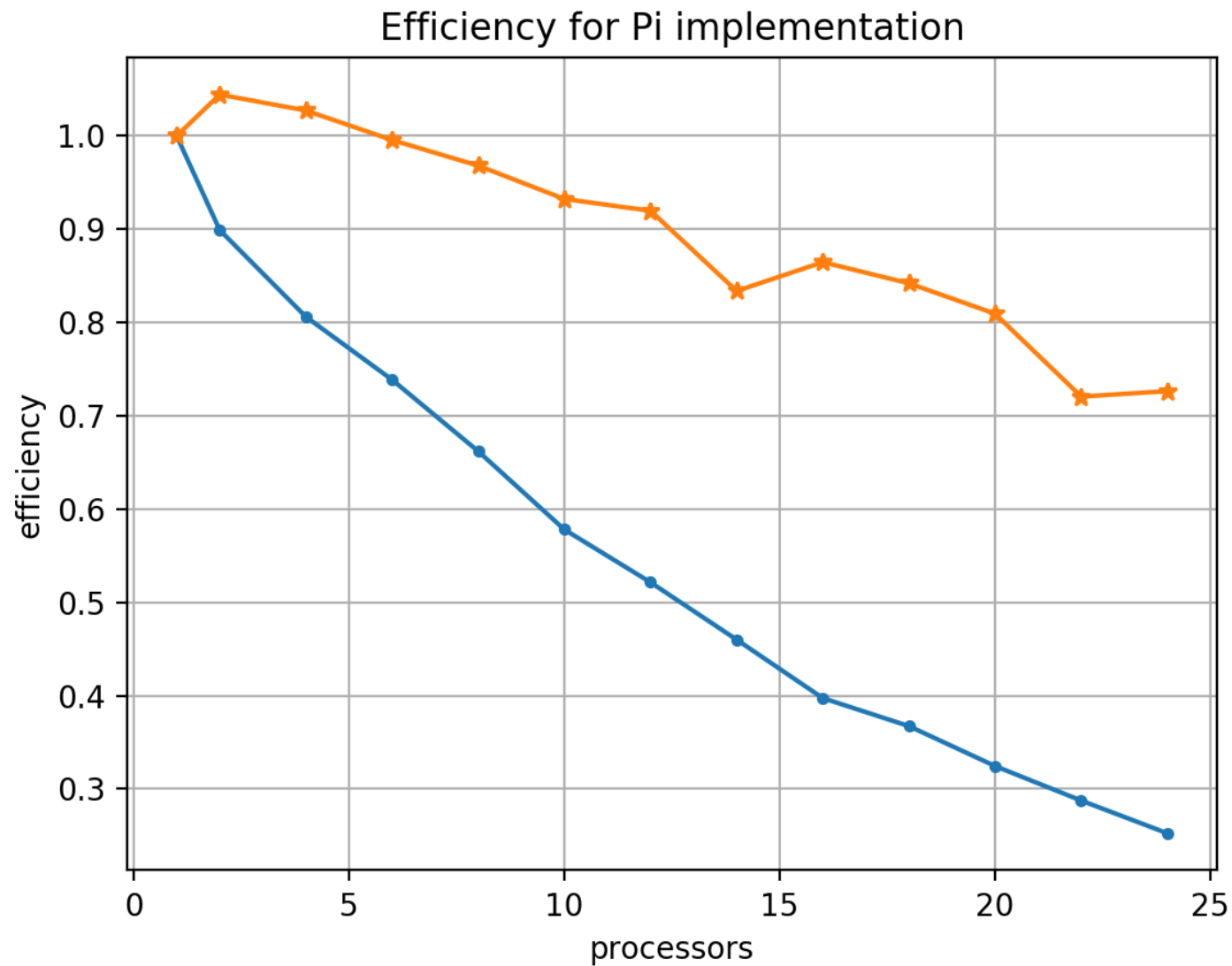
## Exercise

- Run the following job scripts on Summit
  - Day1/examples
    - 04-serial.sh
      - Creates timing for the serial time of the program
    - 04-strong\_scaling.sh
      - Creates timing for strong scaling
    - 04-weak\_scaling
      - Creates timing for weak scaling

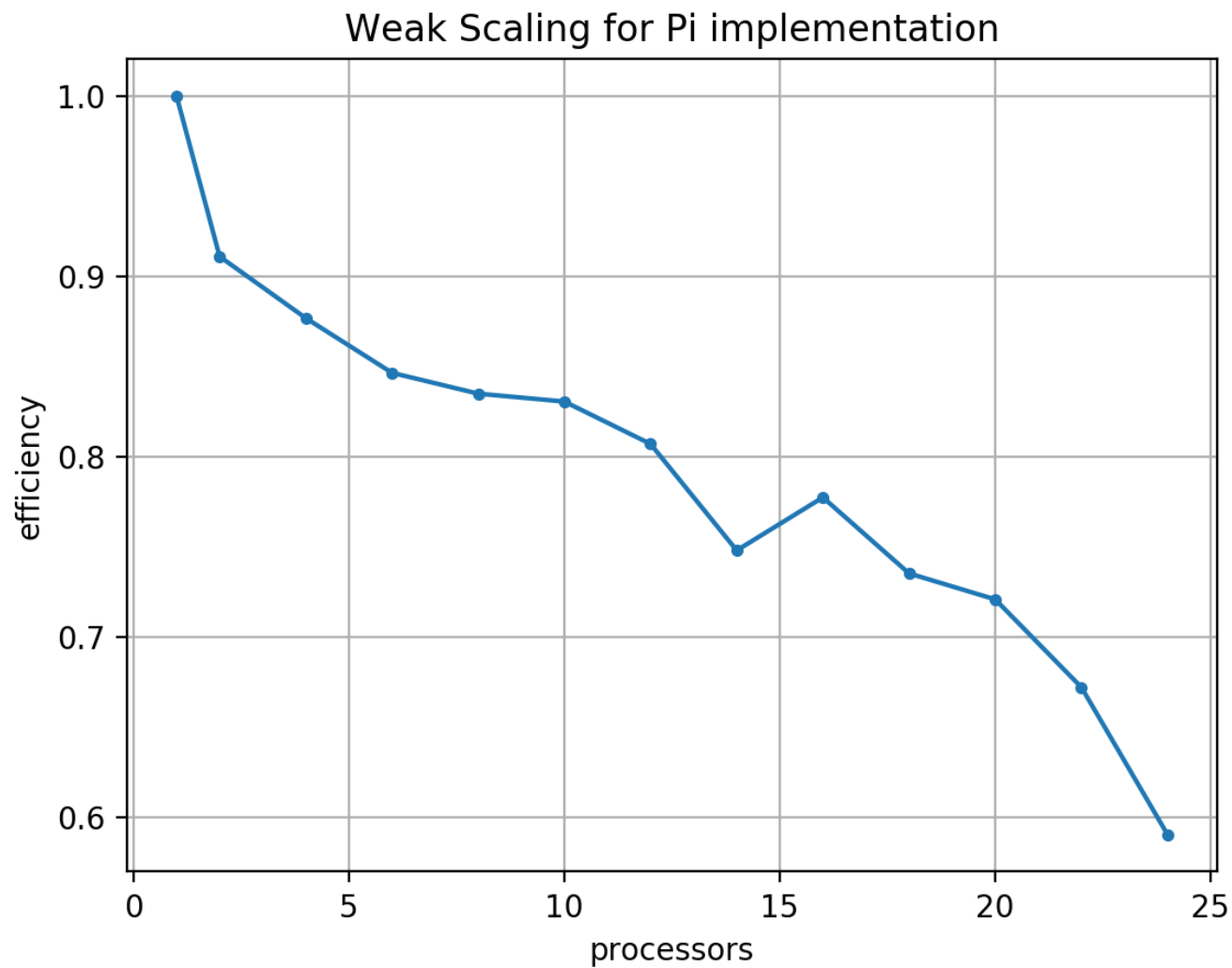
# Speedup



# Efficiency



# Weak Scaling





# Memory usage

- Massif as part of valgrind
- Valgrind help with memory related issues
  - `$ module load valgrind`
  - `$ valgrind --tool=massif python3 pi_serial.py 1000000`
  - `$ ms_print massif.out.88892`

# Massif Memory Usage Report

