Lecture 2.2
Speedup and Efficiency

EN 600.320/420

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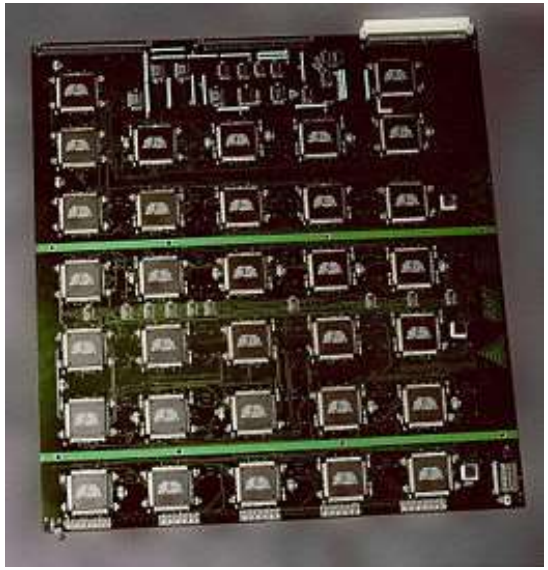
The What of Parallelism?

- Solve bigger problems (weak scaling)
  - Largest direct numerical solutions have evolved from meshes of 1024^3 to 8096^3 in the last ten years
  - Reveals fundamental structure in the nature of turbulent flow
- Solve the same problems faster (strong scaling)
  - Brute force a 56-bit DES key has evolved from
    - $20M (est) 1976
    - 1998, 4.5 days, $250,000
    - 1999 22 hours, $250,000
    - 2006, 9 days, $10,000
    - 2008, 1 days, $10,000



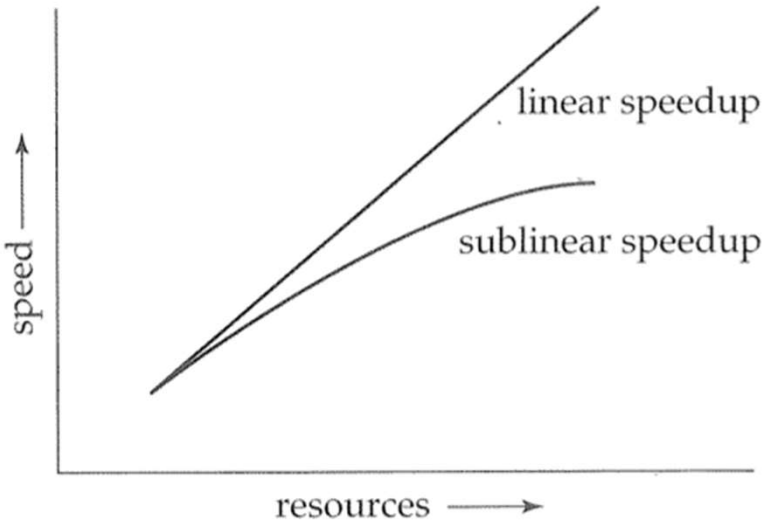
The Why of Parallelism?

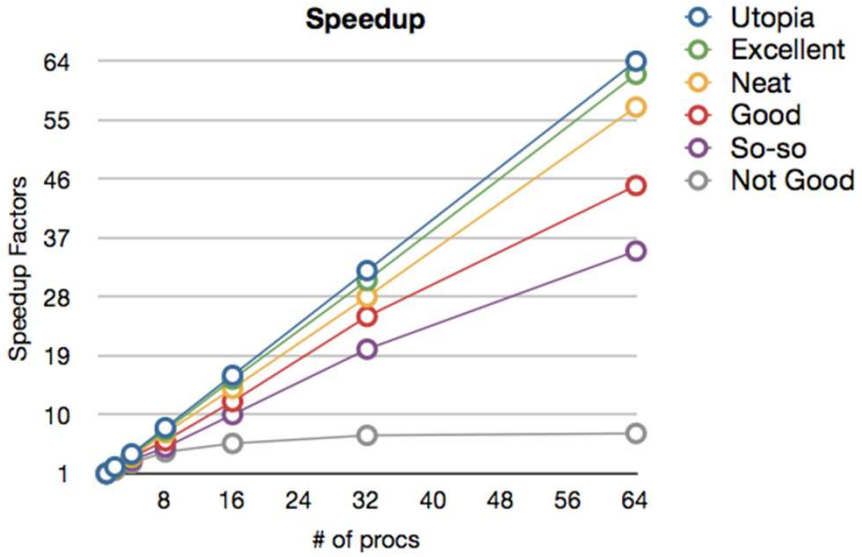
- Solve bigger problems
- Solve the same problems faster or cheaper

- Architectures demand it
  - Multi-core
  - GPUs
- Minimize energy consumption
- Maximize investment

Speedup

- The fundamental concept in parallelism
  - T(1) = time to execute task on a single resource
  - T(n) = time to execute task on n resources
  - Speedup = T(1)/T(n)





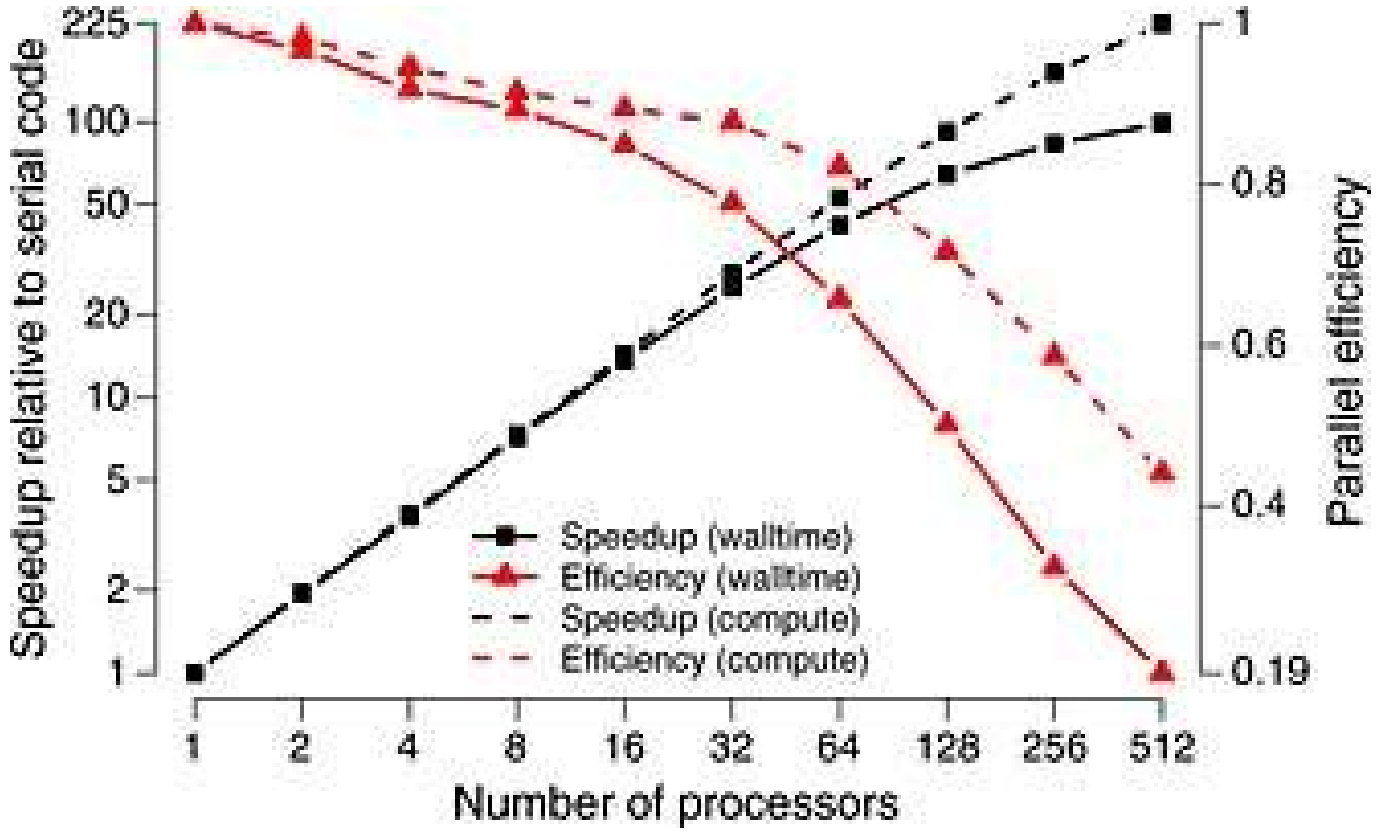
http://web.eecs.utk.edu/~huangj/hpc/hpc_intro.php

From Silberschatz, Korth, and Sudarshan. *Database Systems Concepts, 4th Ed.*

Parallel Efficiency

https://www.researchgate.net/figure/233885489_fig3_Figure-3-Speedup-black-squares-and-parallel-efficiency-red-triangles-for-the

- Companion concept to speedup
  - Efficiency = S(n)/n = T(1)/nT(n)
  - Informally: fraction of possible performance realized



Strong versus Weak Scaling

- Strong scaling: how the solution time varies with the number of processors for a fixed *total* problem size
  - Efficiency and speedup described for strong scaling
  - Amdahl's law addresses strong scaling

- Weak scaling: how the solution time varies with the number of processors for a fixed problem size *per processor*
  - There is another law—Gustavon's Law—that governs weak scaling
  - We'll do weak scaling another day

From Silberschatz, Korth, and Sudarshan. *Database Systems Concepts, 4th Ed.*