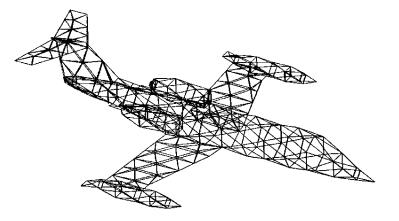
GPU Acceleration of <u>BIG</u> Matrix Algebra

HP-CAST

Dr. Ronald Young November 10, 2012

Why do you want to solve a BIG problem?

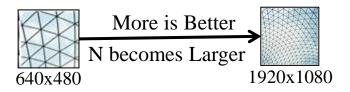
Scientific community Fluid Mechanics Structural Analysis Heat Transfer Electromagnetics Diffusion (reservoir simulation) Acoustics Circuit design Economic modeling NVIDIA Tesla users



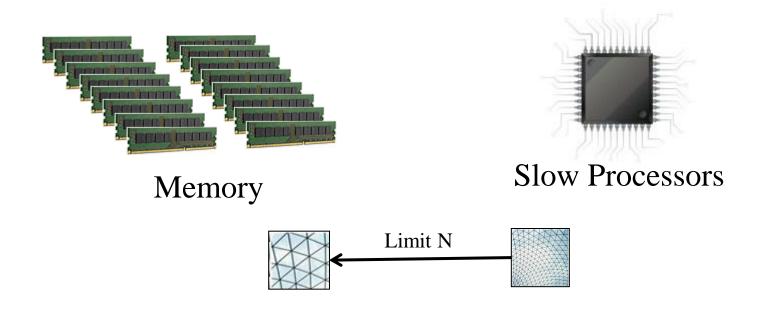
Discretized mesh model

A(N,N) is a model of the plane

 $[\mathbf{A}]\{\mathbf{X}\} = \{\mathbf{B}\}$



What limits the size of the problem (N)?



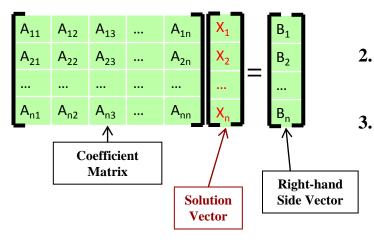
Goal: Make N as big as possible economically

$$A_{11}X_1 + A_{12}X_2 + A_{13}X_3 + \dots + A_{1n}X_n = B_1$$

$$A_{21}X_1 + A_{22}X_2 + A_{23}X_3 + \dots + A_{2n}X_n = B_2$$

$$\dots$$

$$A_{n1}X_1 + A_{n2}X_2 + A_{n3}X_3 + \dots + A_{nn}X_n = B_n$$



As N becomes large, it encounters three obstacles:

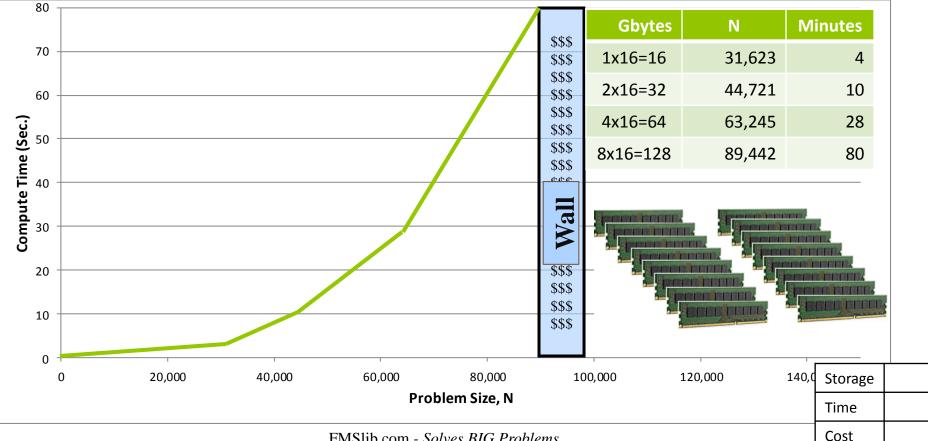
- **1.** <u>Storage</u> for matrix A(N,N) increases as <u>N</u>²
 - Limited by size of memory or disk
 - **<u>Computing time</u>** for $[A]{X}={B}$ increases as <u>N</u>³
 - Limited by processing power (CPUs, GPUs)

Cost/Performance

• \$/Gflop increases with performance

Storage	
Time	
Cost	

Storage: first stop-server memory



FMSlib.com - Solves BIG Problems

Three Ways to Overcome the Storage Obstacle:

1. Use larger very expensive memory modules

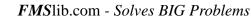
Gbytes	N	Minutes	GPU \$
16x16=256	126,491	255	12X
32x16=512	178,885	636	40X

2. Build a cluster

- A. Replicate server
- B. Reprogram in MPI
- C. Scales storage (N^2) and compute (N^3) equally
- D. Only extends the "Wall"

3. Store data on disk

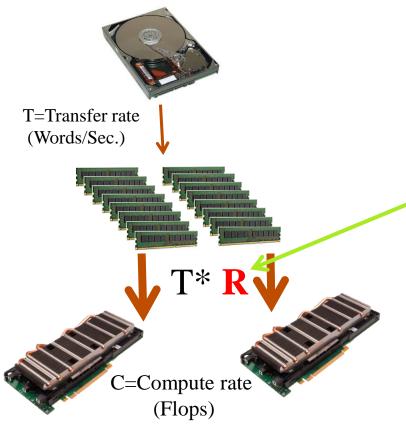
- A. Inexpensive (100 x less than memory)
- B. Practically unlimited size; easily added
- C. Independent scaling of storage and compute
- D. No "Wall"





Storage	
Time	
Cost	

Are Disks Fast Enough?

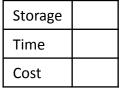


Yes, because of *Reuse*

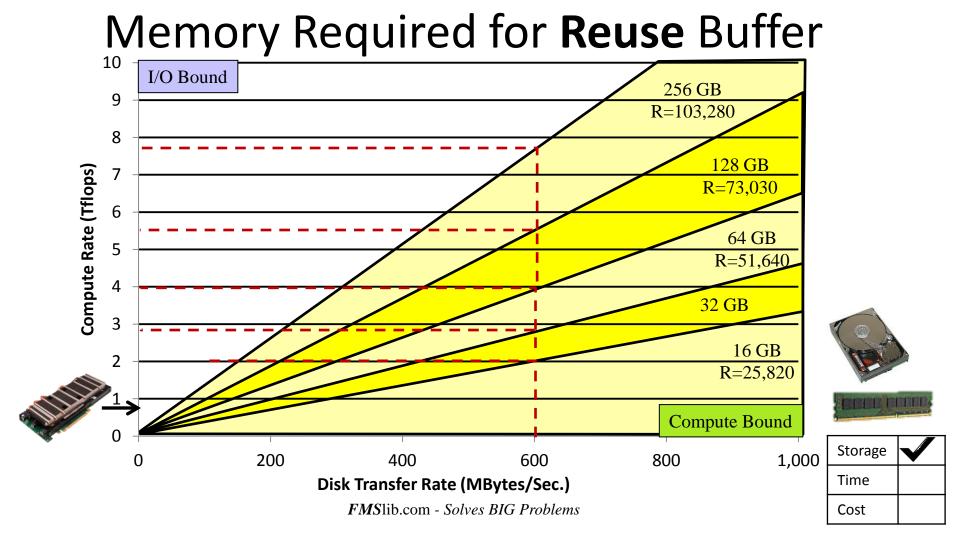
	Real	Complex
Compute $[C] = \sum [A_i] [B_i]$ (Flops)	2N ³	8N ³
Read next [A _i] and [B _i] (Words)	2N ²	4N ²
"Reuse" R	N	2N

IO/Compute Ratio, X=(T*R)/C

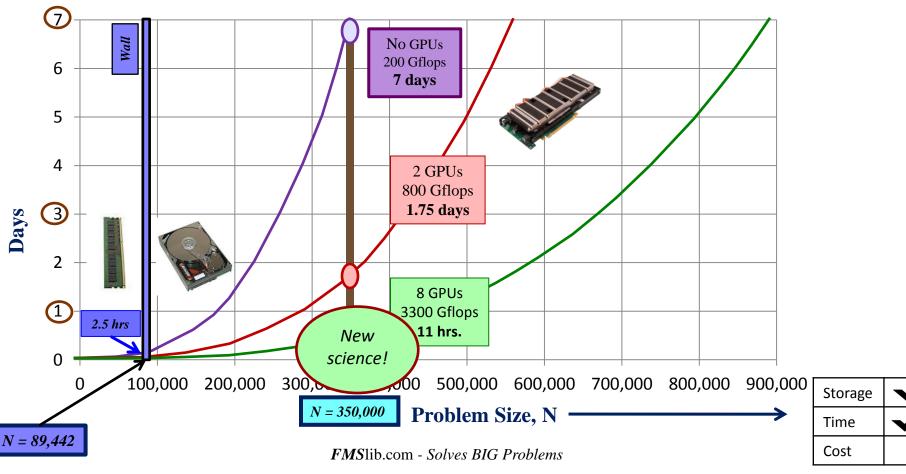
- If X > 1, Compute bound:
 - Increase processing power by X
 - If **X** < **1**, <u>I/O bound:</u>
 - 1. Increase disk transfer rate to C/R
 - 2. Increase reuse to C/T by increasing memory
 - 3. Some combination of (1) and (2)



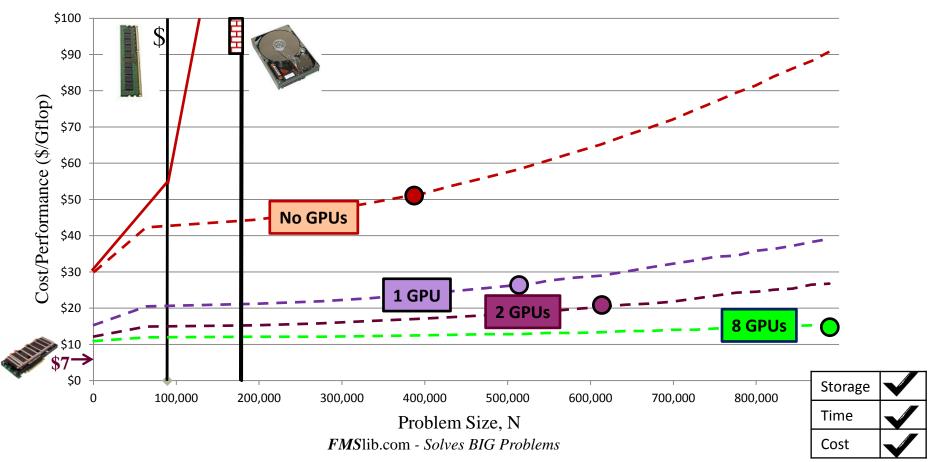
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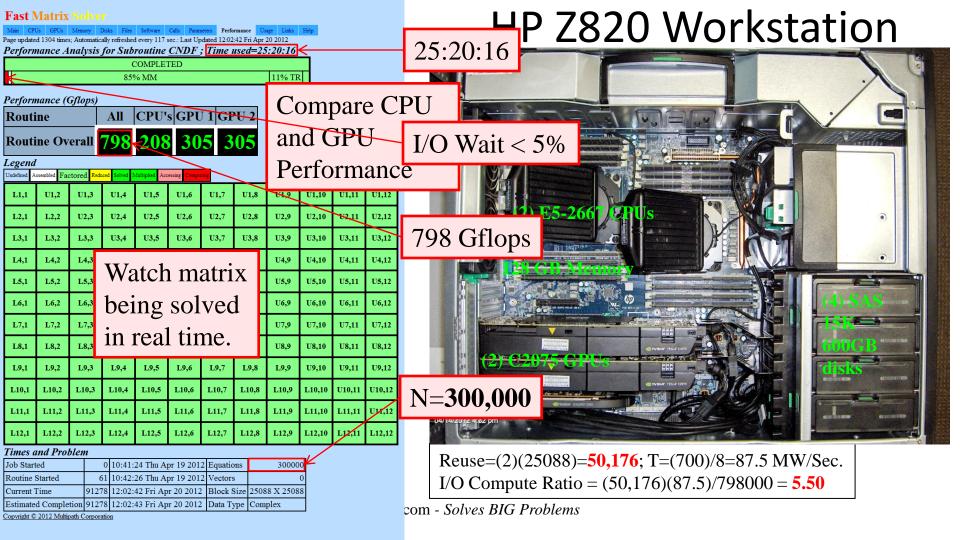


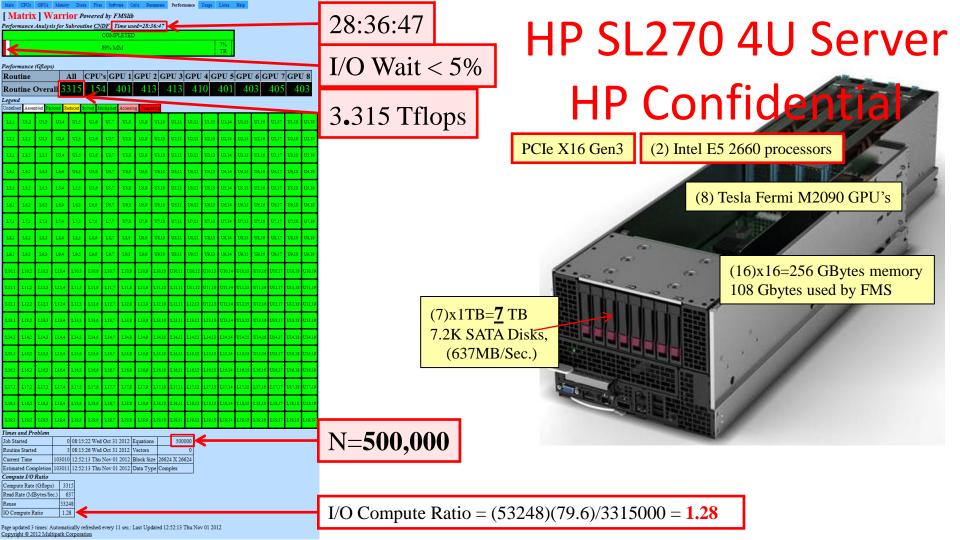
Computing Time: *What can these GPUs really DO?*



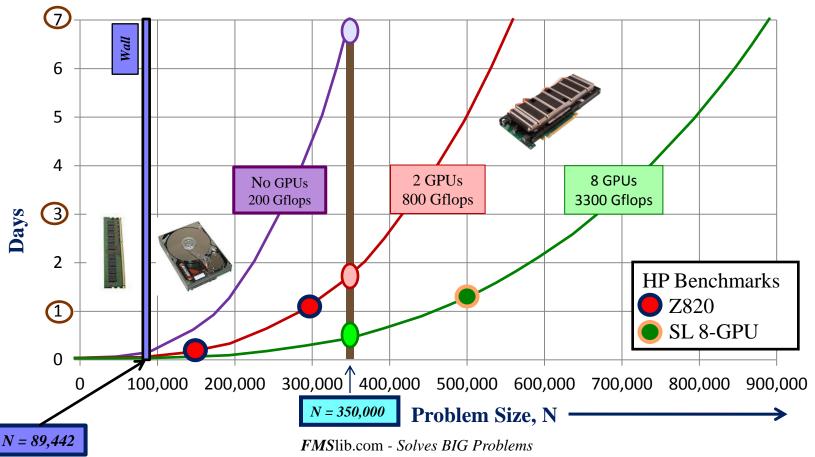
Cost Performance (Processing Efficiency)



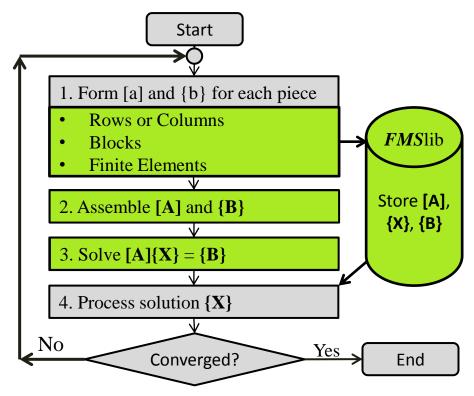




HP Benchmarks



Application Program Interface: **Disk-Based**



Why **FMSlib**?

- 1. FMSlib is based on an in-depth understanding of mathematics and computer architecture, incorporating no shortcuts. Performance is obtained by exploiting all hardware features.
- **2. FMSlib** was the first linear algebra library, initially introduced in 1982 by Floating Point Systems to accelerate their array processors.
- 3. FMSlib includes only those routines that have proven commercial value.

FMSlib Solvers

iile	PROFILE SOLVER : Accounts for the sparsity of matrix [A] on an equation by equation basis
	BLOCK SOLVER : Divides the matrix [A] into square blocks, accounting for sparsity on a block by block basis
ock	SLAB SOLVER: Extends

PROFILE SOLVER : Accounts for the sparsity	Data Type	8-byte Real 16-byte Complex
of matrix [A] on an equation by equation basis BLOCK SOLVER : Divides the matrix [A] into square blocks, accounting for sparsity on a block by block basis SLAB SOLVER : Extends the functionality of the Block Solver by providing full column partial pivoting for full nonsymmetric matrices.	Matrix Symmetry	Symmetric Nonsymmetric Hermitian
	Direct	No iteration Predictable performance
	Dense	No indirect addressing Maximum GPU performance
	Out-of- core	Option to use disk for data storage
	Multiple Solutions	Efficiently solves for multiple {X}
	GPUs	Plug-and-play
	OS	Linux, Windows





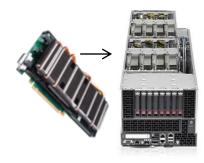
Slab

FMSlib Performance History 1978-present



Machine	Year	Flops	N*	\$/Gflop
DEC VAX	1978	97,000	1,465	2,000,000,000
FPS 164	1982	11,000,000	7,090	50,000,000
FPS 164-MAX	1985	341,000,000	22,272	2,500,000

* Factor full complex nonsymmetric matrix in 1 day



	Y			
(8) NVIDIA GPUs	2012	3,300,000,000,000	474,627	15
		ion times on times		

Demonstrate your GPU Performance with Matrix Warrior

- 1. Demonstrate new computer technology
- 2. Benchmark performance studies (CPUs, GPUs, Memory, Disks)
- 3. Assess existing machine performance (your laptop to GPU server).

Free download from FMSlib .com

Conclusions

- 1. Larger matrices result from more detailed analysis.
- 2. Matrix Algebra scales differently than most applications:
 - Storage as N2
 - Processing as N3
- 3. High <u>**Reuse**</u> in matrix algebra allows efficient use of multilevel memory systems:
 - Inexpensive disks can be used for storage
 - Overlapped transfers from Disk \rightarrow Memory \rightarrow (GPUs) \rightarrow Cache \rightarrow Registers
 - Processors continuously operate at near peak speed
- 4. GPUs have an ideal architecture for matrix algebra:
 - High performance
 - Lower capital and operational costs
- 5. [Matrix]Warrior may be used for
 - Machine benchmarking
 - Demonstrating performance
 - Machine burn in

It's a new day in scientific computing **FMSlib.com** [Matrix] Warrior



3 + Tflops Sustained!, Low \$/Gflop

Solves BIGGERGProblems