General Purpose Computation on Graphics Processors (GPGPU)

Mike Houston, Stanford University

A little about me

- http://graphics.stanford.edu/~mhouston
- Education:
 - UC San Diego, Computer Science BS
 - Stanford University, Computer Science MS
 - Currently a PhD candidate at Stanford University
- Research
 - Parallel Rendering
 - High performance computing
 - Sequoia
 - Computation on graphics processors (GPGPU)
 - Brook, Folding@Home (GPU)

The world changed over the last year...

- Multiple GPGPU initiatives
 - Vendors without GPGPU talking about it
- A few big apps:
 - Game physics
 - Folding@Home
 - Video processing
 - Finance modeling
 - Biomedical
 - Real-time image processing

- Courses
 - UIUC ECE 498
 - Supercomputing 2006
 - SIGGRAPH 2006/2007
- Lots of academic research
- Actual GPGPU companies
 - PeakStream
 - RapidMind
 - Accelware
 - ...

What can you do on GPUs other than graphics?

- Large matrix/vector operations (BLAS)
- Protein Folding (Molecular Dynamics)
- Finance modeling
- FFT (SETI, signal processing)
- Ray Tracing
- Physics Simulation [cloth, fluid, collision,...]
- Sequence Matching (Hidden Markov Models)
- Speech/Image Recognition (Hidden Markov Models, Neural nets)
- Databases
- Sort/Search
- Medical Imaging (image segmentation, processing)
- And many, many, many more...

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Why use GPUs?

- Commodity off the shelf (COTS)
 - In every machine
- Performance
 - Intel Core2 Duo
 - 48 GFLOPS peak
 - 10 GB/s to main memory
 - AMD HD2900XT
 - 475 GFLOPS peak
 - 100 GB/s to video memory
- Lots of Perf/Watt and Perf/\$

Rolling your own GPGPU apps

- Lots of information on GPGPU.org
- Use graphics APIs (old way)
- High level languages and systems to make GPGPU easier
 - PeakStream
 - RapidMind
 - Brook
 - CUDA
- Mid-level
 - CTM (CAL)
- Low-level
 - Full control over hardware CTM (HAL)

GPGPU/Streaming Languages

- Why do you want them?
 - Make programming GPUs easier!
 - Don't need to know OpenGL, DirectX, or ATI/NV extensions
 - Simplify common operations
 - Focus on the algorithm, not on the implementation

PeakStream

http://www.peakstreaminc.com

RapidMind

Commercial follow-on to Sh http://www.rapidmind.net

Accelerator

- Microsoft Research
 http://research.microsoft.com/downloads
- Brook
 - Stanford University http://brook.sourceforge.net
- CUDA – NVIDIA
- CTM – AMD

BrookGPU

History

- Developed at Stanford University
- Goal: allow non-graphics users to use GPUs for computation
- Lots of GPGPU apps written in Brook
- Design
 - C based language with streaming extensions
 - Compiles kernels to DX9, OpenGL, CTM
 - Runtimes (DX9/OpenGL/CTM) handle GPU interaction
- Used for Folding@Home GPU code
 - Large, real application

Streams

Collection of records requiring similar computation

- particle positions, voxels, FEM cell, ...

```
Ray r<200>;
float3 velocityfield<100,100,100>;
```

- Similar to arrays, but...
 - index operations disallowed: position[i]
 - read/write stream operators

```
streamRead (r, r_ptr);
```

```
streamWrite (velocityfield, v_ptr);
```

Functions applied to streams

- similar to for_all construct
- no dependencies between stream elements

Kernel arguments

input/output streams

- Kernel arguments
 - input/output streams
 - gather streams

```
kernel void foo (..., float array[] ) {
    a = array[i];
}
```

- Kernel arguments
 - input/output streams
 - gather streams
 - constant parameters

```
kernel void foo (..., float c ) {
    a = c + b;
}
```

Brook for GPUs

Open source - Sourceforge

- CVS tree *much* more up to date (includes CTM support)
- Project Page
 - http://graphics.stanford.edu/projects/brook
- Source
 - http://www.sourceforge.net/projects/brook
- Paper:

Brook for GPUs: Stream Computing on Graphics Hardware

Ian Buck, Tim Foley, Daniel Horn, Jeremy Sugerman, Kayvon Fatahalian, Mike Houston, Pat

Hanrahan







Fly-fishing fly images from The English Fly Fishing Shop

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PeakStream

- http://www.peakstreaminc.com
- C/C++ library based
- Extended operators
 - Array constructs
- Just-in-time compilation/optimization
 - Really amazing vectorization and scheduling
- Fairly easy switch for Fortan/Matlab users
- GPUs and multi-core

RapidMind

http://www.rapidmind.net/

Embedded within Standard C++

- No new tools, compilers, preprocessors, etc.

Portable core

- Exposes platform specific functionality to also allow tuning for specific platforms
- Integrates with existing programming models
 - Wrap your compute heavy code in extensions
 - Leave rest of the code alone
- GPUs, multi-core, Cell

CUDA - NVIDIA

- C with extensions/limitations
- Similar to Brook, but better matched to G80 hardware
- Complete toolchain
 - Compiler/profiler/debugger
- Standard provided libraries
 - BLAS, FFT
- Support for G80+
- Programmer's guide and lots of hardware information available

- Exposed hardware traits
 - Shared memory
 - Multiple memory levels
 - Register limits
- Data-parallel + multi-threading
 - Exposes the steaming model as data-parallel at the top
 - Exposes multi-threading at the bottom
 - Threads/Warps/Blocks

AMD CTM

- HAL low level
 - Streaming processor ISA
 - Command buffer control
 - Full memory layout
 - Tiling
 - Addressing
 - Full processor scheduling

- CAL mid level
 - Stream processor compilers
 - HLSL/PS3/PS4
 - Command buffer assistance
 - Memory management
 - Allocation
 - Layout
 - Multi-GPU handling

- Basically a low level driver interface
- Similar to DX/GL GPGPU programming but semantics to match computation and without the overheads of graphics APIs

AMD + Stream Processing Future

Brook

- AMD actively supporting CTM Brook backend for R5XX/R6XX
- CTM CAL is open platform
 - http://sourceforge.net/projects/amdctm
 - Sits atop HAL for each target
- Streaming extensions to current programming languages
 - Brook-like at first, then expanding
- Multiple platforms
 - GPUs and multi-core with same abstraction through CAL
- Support multiple abstraction levels
 - High-level Brook support and streaming extensions to standard languages
 - Mid-level (CAL) memory management, command buffers, etc
 - Low-level (HAL) direct access to hardware, including ISA

Sequoia – Programming the memory system

Hierarchical stream programming

- Similar to multi-level Brook
- Performance is about programming the memory system
 - Compute is "free"
 - Fast application have data locality at all levels
- We can run on "difficult" systems
 - Cell/PS3
 - Clusters
 - Multi-core/SMP
 - Cluster of Cells in SMP
- http://sequoia.stanford.edu

GPGPU and the HD 2900XT

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GPGPU on the HD 2900XT

- 32-bit floating point
 - Tighter IEEE conformance compared to last generation
 - $\frac{1}{2}$ ULP on MUL/ADD,
 - 1 ULP on MAD
 - Denorm/underflow treated as 0
 - No exception support
- Integer support
 - Much more natural stream addressing
- Scalar floating-point engines
 - 320 of them

- Long programs
 - No instruction limits (up to available memory)
- Branching and Looping
 - Low overhead branching and looping
 - Fine branch granularity:
 - ~64 fragments

GPGPU on the HD 2900XT, cont.

- Advanced memory controller
 - Latency hiding for streaming reads and writes to memory
 - With enough math ops you can hide all memory access!
 - Large bandwidth improvement over previous generation
 - Fetch4 performance without the pain

- Large memory addressing
 - 40-bit address space
- Faster upload/download
 - DMA engine

GPGPU on the HD 2900XT, cont.

- Read/Write cache
 - 8KB of on-chip storage
 - Spill to memory
 - Lots of interesting uses
 - Reduction buffer
 - Unlimited writes
 - Register spill
 - • •

Flexibility

- Unlimited texture reads
- Scalar engines
- Lots of latency hiding
- Heavy register usage without penalty
 - Fewer threads in flight, but you should have more math

Performance basics for GPGPU – HD 2900XT

- Compute (DX/CTM)
 - 470 GFLOPS (MAD observed)
 - No games, just a giant MAD kernel
- Offload to GPU (DX)
 - Readback (GPU → CPU):
 1.4 GB/s
 - Download (CPU→ GPU):
 2.4GB/s
- Branch granularity (DX/CTM)
 - ~64 threads

- Memory (CTM)
 - 180 GB/s cache bandwidth
 - 60 GB/s streaming bandwidth
 - 8 GB/s random access
 - 1 cycle latency for a float4 fetch (cache hit)
 - 4 cycle latency for a float4 fetch (streaming)

1 cycle = 5 four-way MADs

Folding@Home

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What is protein folding

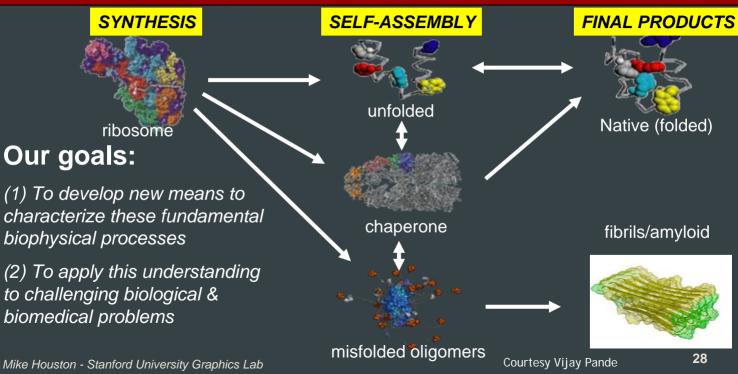
Protein folding: a fundamental biophysical problem



- What are proteins and why do they "fold"?
 - Proteins are molecules in the body which carries out many important functions, such as enzymes and antibodies
 - Before proteins can carryout their function, they must first assemble themselves or "fold"
- Potential impact
 - Design of novel proteins & protein-like heteropolymers
 - Understanding protein misfolding related diseases

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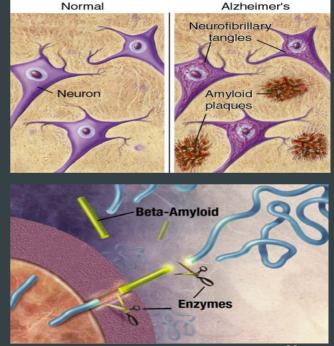
Protein folding in the cell



Molecular nature of Alzheimer's Disease

Molecular nature

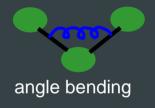
- Brain tissue contains deposits of beta-Amyloid peptide (Aβ)
- amphipathic 39-42 residue fragment of a membrane protein cleaved by secretases
- Aggregate first into oligomers, then fibrils
- Intrinsically unstructured protein
 - Important new class of proteins
 - Common issue in many protein misfolding related diseases
 - Very different paradigm for structural study



The 25+ year old dream: use the laws of physics to study biomolecules

- Biomolecules obey the laws of physics
 - Use physical chemistry theory and simulation techniques to study biomolecules
 - All atom model with physical interactions
- Physics-based model: Molecular Dynamics
 - Numerically integrate Newton's equations
 - Choose δt to match timescale ($\delta t \sim 10^{-15}$ sec)
 - Classical model, with parameters to match experiment and/or quantum mechanics calculation
- Comparison to informatic approaches
 - Not limited by protein experiment: examine structures not in PDB (eg aggregates)
 - Possibility to include non-protein molecules (eg small molecules)
 - BUT, very, very computationally demanding



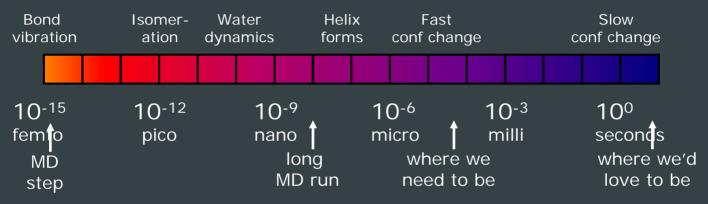




van der Waals



electrostatics



- Fundamental problem for simulation
 - Proteins fold in micro- to milliseconds
 - Computers can simulate nanoseconds
 - How can we break this impasse?

Natural idea: Use multiple processors

- Natural first idea: use multiple CPU's
 - Typical calculation needed requires 1,000,000 days on 1 fast processor
 - How about running on 100,000 CPU's?
 - "Linear scaling" would suggest this would take just 10 days with 100,000 CPU's
- Life isn't that easy
 - Most codes can't scale this way
 - In 2000, we suggested a method to solve this problem
 - Newest version (2005), allows for a very efficient use of 100K CPU's

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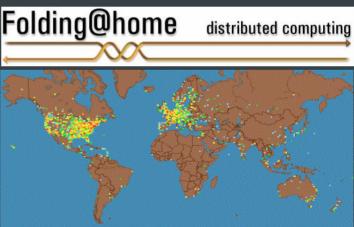


Can 2000 grad students complete a PhD's worth of research in 1 day?

Most likely not!One must design novelways to utilize large scale resources in
efficient ways.32

Folding@Home: Computational resources from the future, realized today

- Very powerful computational resource
 - ~<u>700 Teraflops sustained performance</u>
 - >2,000,000 total processors; ~250,000 active
- Big impact
 - Break the micro- to millisecond barrier
 - What used to take 1M days, now takes a week or two
 - Enables key research previously impossible
 - These new methods are also starting to be used by other groups for folding study



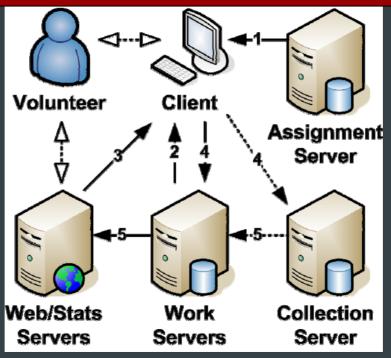
>250,000 active CPUs over the world jay Pande (CPU locations from IP address)

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Courtesy Vijay Pande

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How does this all work?



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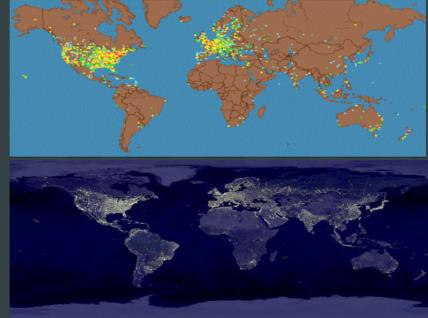
Who's Folding?

Folding@Home

~100,000 CPUs over the world (CPU locations from IP address)

Earth@Night

Electricity as distributed in the world (from NASA satellite data)

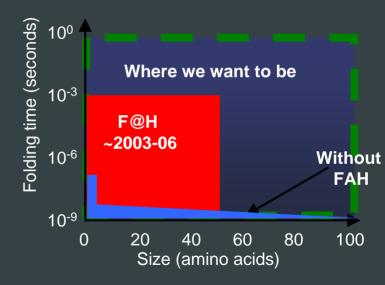


Folding@Home communities – small subset



Next steps: new challenges

- What we can do
 - Somewhat long timescales (microseconds to milliseconds)
 - Somewhat large peptides & proteins (30-50 residues)
 - Allows us to directly test the relevant chemical detail in these biophysical problems
- Where we want to be
 - Longer timescales (seconds?)
 - Larger systems (~100 residues)
 - Yet not sacrificing details



Streaming processors on FAH

New processors

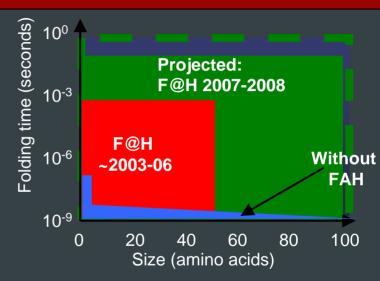
- Originally special purpose (eg for graphics, games)
- High flops: ~500GFLOPS peak programmable, 32-bit FLOPs

Utilize for scientific computing

- We have ported our MD code to GPUs & Cell (PS3)
- Partnership with Sony and AMD
- 20x to 40x speed increase
- Impact on calculations
 - Critical for MSM's: longer trajectories, not just more of them
 - Tackle more complex systems

Folding@Home's next steps

- New methods
 - New processors and new advances in implicit solvation models allow for a ~20x to 500x speedup
 - Achieve simulations on the second timescale
- Impact: sampling may soon not be a problem
 - "@Home" not needed to simulate sub-millisecond timescale
 - More complex problems would be in reach with distributed computing
 - Turn our attention to further improving models (experiment key for these next steps)



With these new capabilities, we would be able to simulate the folding of a significant fraction of protein domains

GPUs, SMPs, PS3s (oh my!)

Performance

- GPU > PS3 > SMP > Single-core
- Flexibility
 - Single-core > SMP > PS3 > GPU

GPUs are <u>really</u> fast

- 20-40x a single core P4
- But they can only run a subset of the research
 - Implicit water models

Points are awarded with this in consideration

Current status - April 20, 2007

http://fah-web.stanford.edu/cgi-bin/main.py?qtype=osstats

Client type	Current TFLOPS*	Active Processors
Windows	177	185870
Mac OS X/PPC	8	10394
Mac OS X/Intel	14	4600
Linux	47	27373
GPU	56	947
Playstation3	409	31217
Total	711	260401

*TFLOPS is actual flops from software cores, not peak values

For More Information

- Main site
 - http://folding.stanford.edu
- GPU information
 - http://folding.stanford.edu/FAQ-ATI.html
- Petaflop Initiative
 - Stream processors
 - GPUs
 - PS3 (Cell)
 - http://folding.stanford.edu/FAQ-FPI.html

Folding on HD 2900XT

- Support will ship with next release
- Force calculation 2.2X(!) faster than R580
- Overall application ~45% faster
 - Amdahl's law strikes again
 - Supporting kernels unoptimized
 - CPU overhead
 - Little tuning time so far on R6XX
- We can now explore more complex algorithms
 - Larger register requirements
 - Integer support random number generation
 - More bandwidth

Future Folding@Home GPGPU

- Improvements
 - Lower CPU overhead
 - Brook DX9 backend + core changes
 - CTM through Brook
 - Lower overhead/finer tuning
 - Linux support much easier
 - Better stability users can upgrade drivers for games without breaking anything
 - More GPUs supported
- New cores
 - Back-port some of what we learned from Cell ports
 - Several new models
 - Many more work units
 - Revisit other models running well on SMP/Cell

Acknowledgements – Folding@Home research group

Current members:

Adam Beberg **Relly Brandman** Kim Branson Jeremy England Dan Ensign Guha Jayachandran Rajdas Jaykumar Peter Kasson Nick Kelley **Del Lucent** Edgar Luttmann Sanghyun Park Paula Petrone Alex Robertson Nina Singhal Eric Sorin Vishal Vaidyananthan

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Making GPGPU easier

What GPGPU needs from vendors – what we got

- More information
 - Shader ISA 🛨
 - Latency information
 - GPGPU Programming guide (floating point) *
- Direct access to the hardware/Compute APIs +
- Fast transfer to and from GPU
 - Non-blocking
- Consistent graphics drivers *
 - Some optimizations for games hurt GPGPU performance

What GPGPU needs from vendors

More information

- Latency information
- Memory system information
- Fast transfer to and from GPU
 - Non-blocking
 - DMA driven
- Consistent drivers
 - Stable drivers for games AND computation
- High-level <u>and</u> low-level access
 - High-level: get more people using GPUs, make it easy to approach
 - Low-level: let us tune to our heart's content!
 - Remove shader compiler and game optimizations from breaking our code
- More software companies
 - PeakStream and RapidMind a good start

What GPGPU needs from the community

Data Parallel programming languages

- Need more exploration
- "GCC" for GPUs/streaming processors
 - We almost have enough information for this
- Parallel data structures
 - Still really hard...
- More applications
 - What will make the average user care about GPGPU?
 - Folding@Home! ;-)
 - What else can we make data parallel and run fast?

Questions?

- I'll also be around after the talk
- Email: mhouston@stanford.edu
- Web: http://graphics.stanford.edu/~mhouston

For lots of great GPGPU information:

- GPGPU.org (http://www.gpgpu.org)