

GPGPU: General-Purpose Computation on GPUs

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EG 2004

Why GPGPU?



- The GPU has evolved into an extremely flexible and powerful processor
 - Programmability
 - Precision
 - Performance
- This talk addresses the basics of harnessing the GPU for general-purpose computation

Motivation: Computational Power



- GPUs are fast...

- 3 GHz Pentium 4 *theoretical*: 6 GFLOPS

- 5.96 GB/sec peak

- GeForce FX 5900 *observed*^{*}: 20 GFLOPs

- 25.6 GB/sec peak

- GeForce 6800 Ultra *observed*^{*}: 40 GFLOPs

- 35.2 GB/sec peak

^{*} Observed on a synthetic benchmark:

- A long pixel shader with nothing but MUL instructions

GPU: high performance growth



● CPU

- Annual growth $\sim 1.5\times \rightarrow$ decade growth $\sim 60\times$
- Moore's law

● GPU

- Annual growth $> 2.0\times \rightarrow$ decade growth $> 1000\times$
- Much faster than Moore's law

Why are GPUs getting faster so fast?



- Computational intensity
 - Specialized nature of GPUs makes it easier to use additional transistors for computation not cache
- Economics
 - Multi-billion dollar video game market is a pressure cooker that drives innovation

Motivation: Flexible and precise



- Modern GPUs are programmable
 - Programmable pixel and vertex engines
 - High-level language support
- Modern GPUs support high precision
 - 32-bit floating point throughout the pipeline
 - High enough for many (not all) applications

Motivation: The Potential of GPGPU



- The performance and flexibility of GPUs makes them an attractive platform for general-purpose computation

● Example applications (from www.GPGPU.org)

- Advanced Rendering: Global Illumination, Image-based Modeling
- Computational Geometry
- Computer Vision
- Image And Volume Processing
- Scientific Computing: physically-based simulation, linear system solution, PDEs
- Stream Processing
- Database queries
- Monte Carlo Methods

The Problem: Difficult To Use



- GPUs are designed for and driven by graphics
 - Programming model is unusual & tied to graphics
 - Programming environment is tightly constrained
- Underlying architectures are:
 - Inherently parallel
 - Rapidly evolving (even in basic feature set!)
 - Largely secret
- Can't simply "port" code written for the CPU!

Mapping Computational Concepts to GPUs



- Remainder of the Talk:
- Data Parallelism and Stream Processing
- Computational Resources Inventory
- CPU-GPU Analogies
- Flow Control Techniques
- Examples and Future Directions

Importance of Data Parallelism



- GPUs are designed for graphics
 - Highly parallel tasks
- GPUs process *independent* vertices & fragments
 - Temporary registers are zeroed
 - No shared or static data
 - No read-modify-write buffers
- Data-parallel processing
 - GPU architecture is ALU-heavy
 - Multiple vertex & pixel pipelines, multiple ALUs per pipe
 - Hide memory latency (with more computation)

Arithmetic Intensity



- Arithmetic intensity = ops per word transferred
- “Classic” Graphics pipeline
 - Vertex
 - BW: 1 triangle = 32 bytes
 - OP: 100-500 f32-ops / triangle
 - Fragment
 - BW: 1 fragment = 10 bytes
 - OP: 300-1000 i8-ops/fragment

Courtesy of Pat Hanrahan

Data Streams & Kernels



● Streams

- Collection of records requiring similar computation
 - Vertex positions, Voxels, FEM cells, etc.
- Provide data parallelism

● Kernels

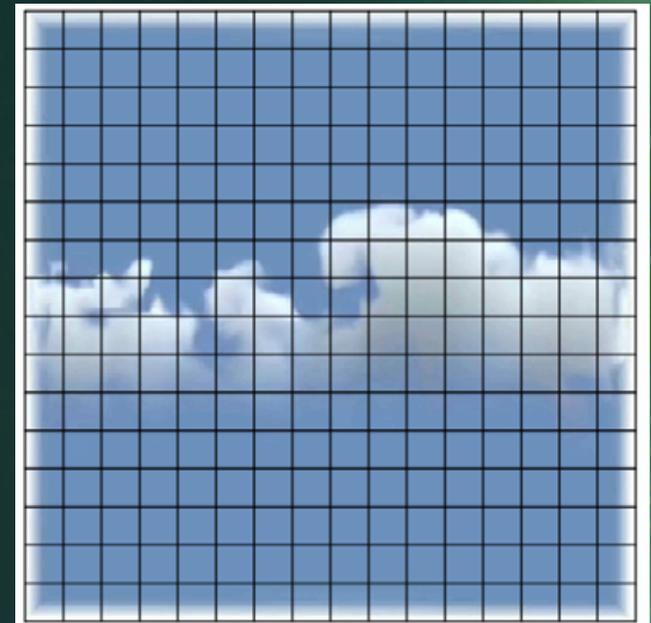
- Functions applied to each element in stream
 - transforms, PDE, ...
- Few dependencies between stream elements
 - Encourage high Arithmetic Intensity

Courtesy of Ian Buck

Example: Simulation Grid



- Common GPGPU computation style
 - Textures represent computational grids = streams
- Many computations map to grids
 - Matrix algebra
 - Image & Volume processing
 - Physical simulation
 - Global Illumination
 - ray tracing, photon mapping, radiosity
- Non-grid streams can be mapped to grids



Stream Computation



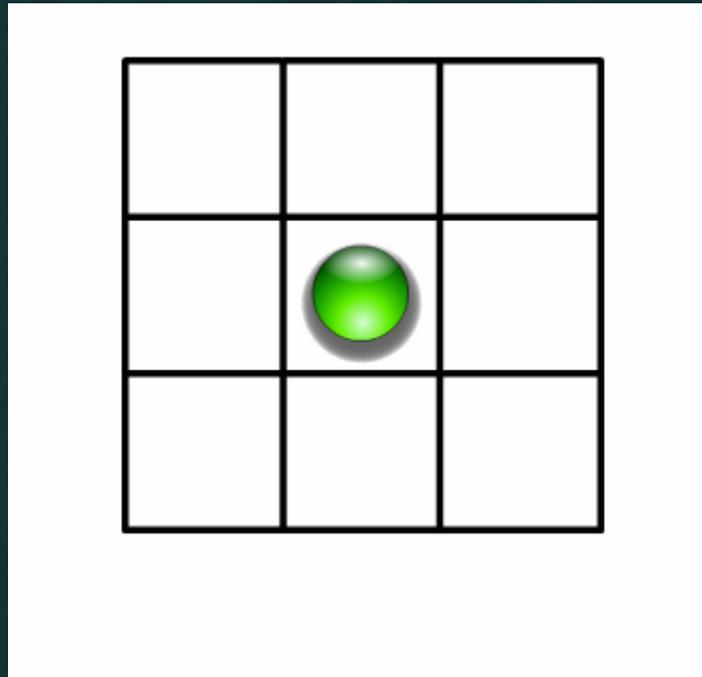
| Algorithm |
|--------------|
| advect |
| accelerate |
| water/thermo |
| divergence |
| jacobi |
| jacobi |
| jacobi |
| jacobi |
| ⋮ |
| jacobi |
| u-grad(p) |

- Grid Simulation algorithm
 - Made up of steps
 - Each step updates entire grid
 - Must complete before next step can begin
- Grid is a stream, steps are kernels
 - Kernel applied to each stream element

Scatter vs. Gather



- Grid communication (a necessary evil)
 - Grid cells share information
 - Two ways:



Computational Resources Inventory



- Programmable parallel processors
 - Vertex & Fragment pipelines
- Rasterizer
 - Mostly useful for interpolating addresses (texture coordinates) and per-vertex constants
- Texture unit
 - Read-only memory interface
- Render to texture
 - Write-only memory interface

Vertex Processor



- Fully programmable (SIMD / MIMD)
- Processes 4-vectors (RGBA / XYZW)
- Capable of scatter but not gather
 - Can change the location of current vertex (scatter)
 - Cannot read info from other vertices (gather)
 - Small constant memory
- New GeForce 6 Series features:
 - Pseudo-gather: read textures in the vertex program
 - MIMD: independent per-vertex branching, early exit

Fragment Processor



- Fully programmable (SIMD)
- Processes 4-vectors (RGBA / XYZW)
- Capable of gather but not scatter
 - Random access memory read (textures)
 - Output address fixed to a specific pixel
- Typically more useful than vertex processor
 - More fragment pipelines than vertex pipelines
 - Gather / RAM read
 - Direct output
- GeForce 6 Series adds SIMD branching
 - GeForce FX only has conditional writes

CPU-GPU Analogies



- CPU programming is (assumed) familiar
 - GPU programming is graphics-centric

- Analogies can aid understanding

CPU-GPU Analogies



GPU Simulation Overview



| Algorithm |
|--------------|
| advect |
| accelerate |
| water/thermo |
| divergence |
| jacobi |
| jacobi |
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| ⋮ |
| jacobi |
| u-grad(p) |

- Analogies lead to implementation
 - Algorithm steps are fragment programs
 - Computational *kernels*
 - Current state variables stored in textures
 - Data *streams*
 - Feedback via render to texture
- One question:
 - How do we invoke computation?

Invoking Computation



- Must invoke computation at each pixel
 - Just draw geometry!
 - Most common GPGPU invocation is a full-screen quad

Standard “Grid” Computation



- Initialize “view” (so that pixels:texels::1:1)

```
glMatrixMode(GL_MODELVIEW);  
glLoadIdentity();  
glMatrixMode(GL_PROJECTION);  
glLoadIdentity();  
glOrtho(0, 1, 0, 1, 0, 1);  
glViewport(0, 0, gridResX, gridResY);
```

- For each algorithm step:
 - Activate render-to-texture
 - Setup input textures, fragment program
 - Draw a full-screen quad (1 unit x 1 unit)

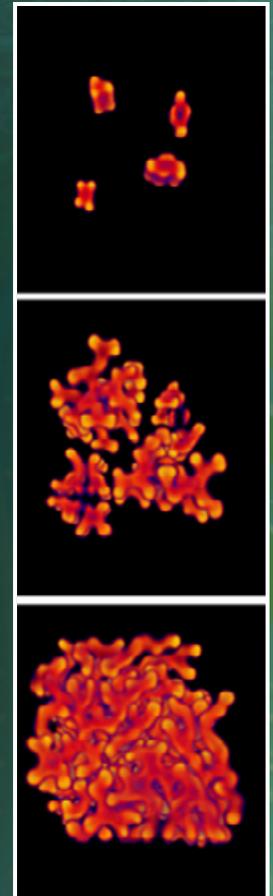
Reaction-Diffusion



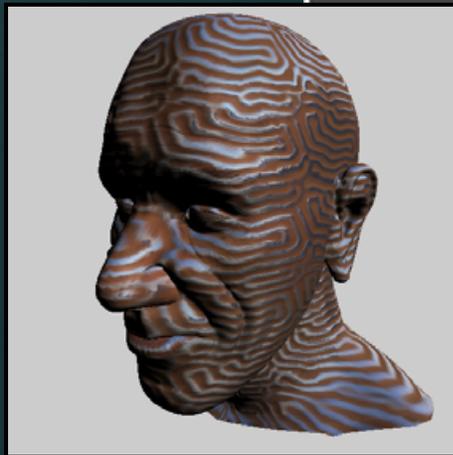
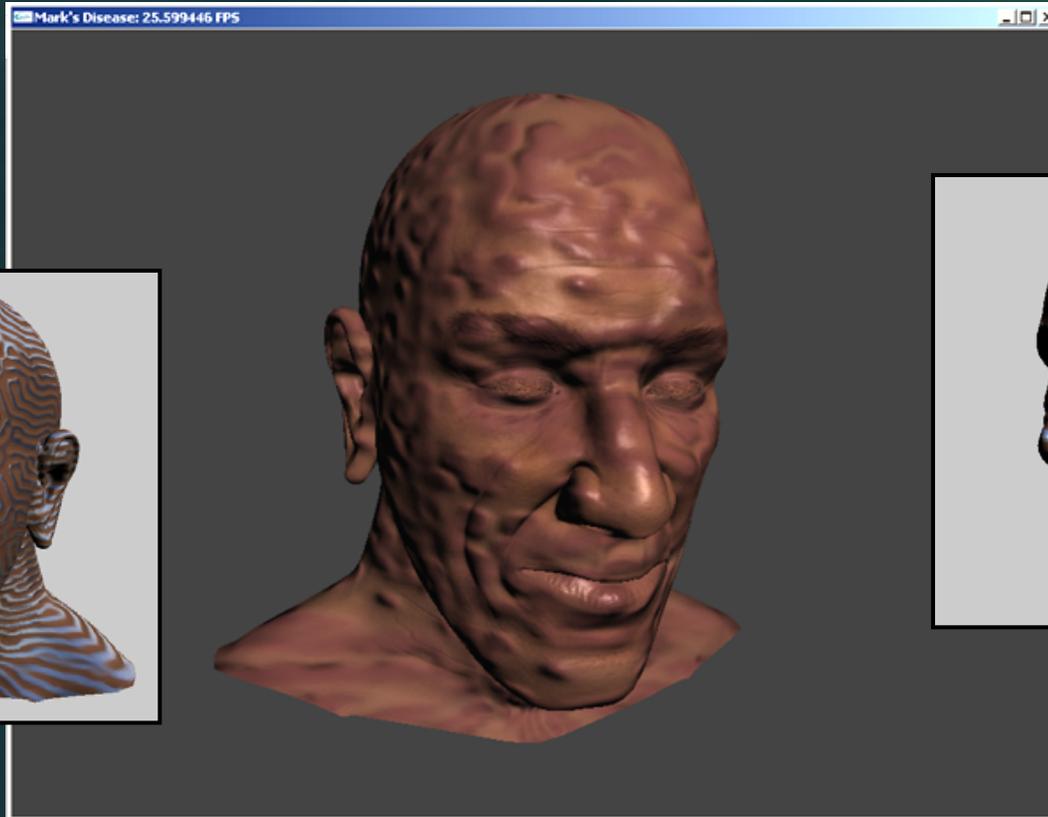
- Gray-Scott reaction-diffusion model [Pearson 1993]
- Streams = two scalar chemical concentrations
- Kernel: just **Diffusion** and **Reaction** ops

$$\frac{\partial U}{\partial t} = D_u \nabla^2 U - UV^2 + F(1 - U),$$
$$\frac{\partial V}{\partial t} = D_v \nabla^2 V + UV^2 - (F + k)V$$

U, V are chemical concentrations,
 F, k, D_u, D_v are constants



Demo: "Disease"



Available in NVIDIA SDK: <http://developer.nvidia.com>

"Physically-based visual simulation on the GPU",
Harris et al., Graphics Hardware 2002

Per-Fragment Flow Control



- No true branching on GeForce FX
 - Simulated with conditional writes: every instruction is executed, even in branches not taken
- GeForce 6 Series has SIMD branching
 - Lots of deep pixel pipelines → many pixels in flight
 - Coherent branching = likely performance win
 - Incoherent branching = likely performance loss

Fragment Flow Control Techniques



- Try to move decisions up the pipeline
 - Replace with math
 - Occlusion Query
 - Domain decomposition
 - Z-cull
 - Pre-computation



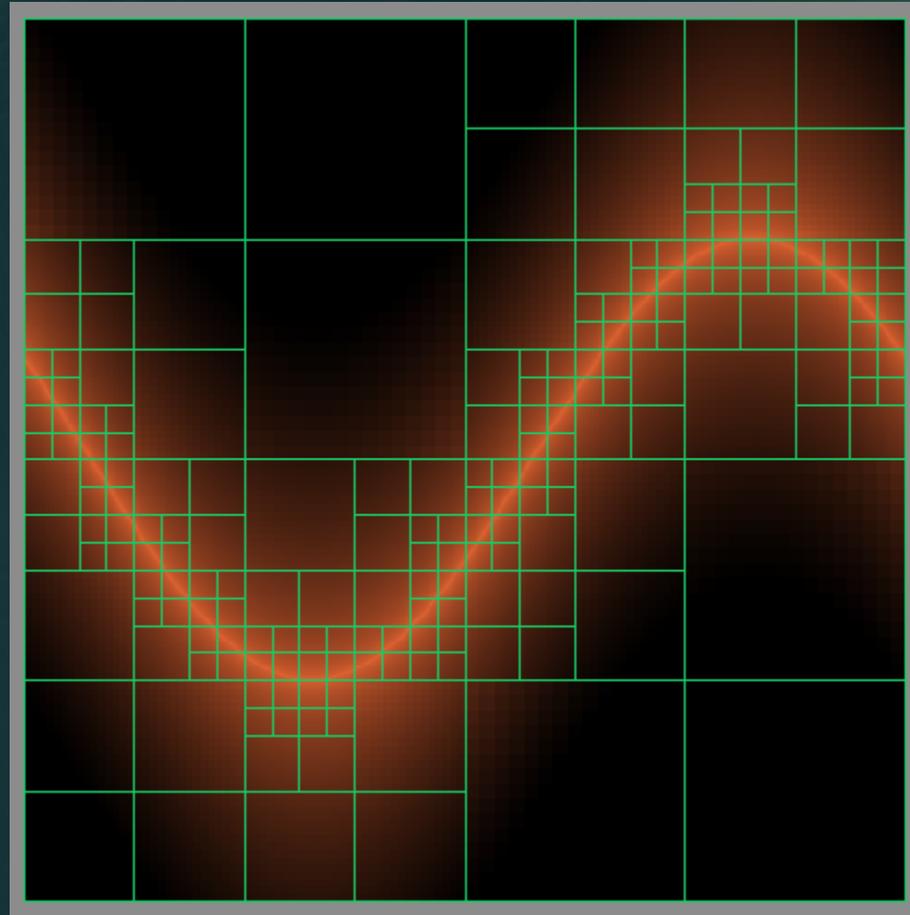
Branching with Occlusion Query

- OQ counts the number of fragments written
 - Use it for iteration termination

```
Do { // outer loop on CPU
    BeginOcclusionQuery {
        // Render with fragment program
        // that discards fragments that
        // satisfy termination criteria
    } EndQuery
} While query returns > 0
```

- Can be used for subdivision techniques

Example: OQ-based Subdivision



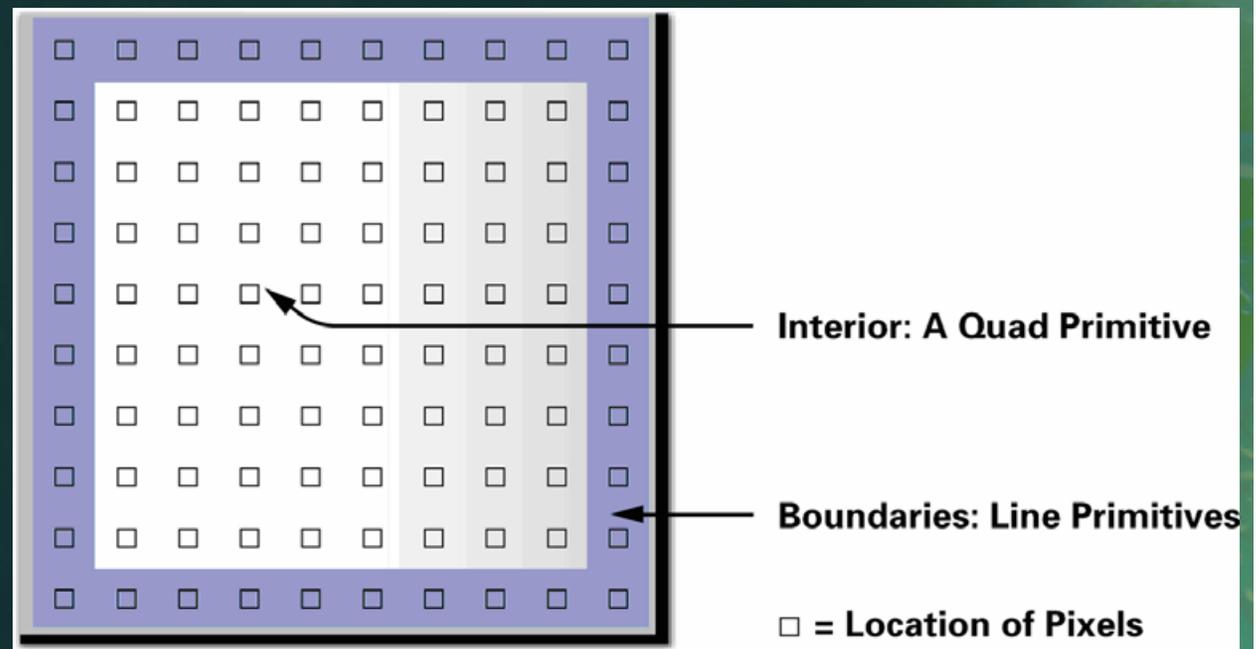
Used in Coombe et al., "Radiosity on Graphics Hardware"



Static Branch Resolution

- Avoid branches where outcome is fixed
 - One region is always true, another false
 - Separate FP for each region, no branches

- Example:
boundaries



Z-Cull



- In early pass, modify depth buffer
 - Clear Z to 1, enable depth test
 - Draw quad at $Z=0$
 - Discard pixels that should be modified in later passes
- Subsequent passes
 - Enable depth test (`GL_LESS`), disable depth write
 - Draw full-screen quad at $z=0.5$
 - Only pixels with previous $\text{depth}=1$ will be processed
- Can also use early stencil test on GeForce 6

Pre-computation



- Pre-compute anything that will not change every iteration!
- Example: arbitrary boundaries
 - When user draws boundaries, compute texture containing boundary info for cells
 - e.g. Offsets for applying PDE boundary conditions
 - Reuse that texture until boundaries modified
 - GeForce 6 Series: combine with Z-cull for higher performance!

GeForce 6 Series Branching



- True, SIMD branching
 - Lots of incoherent branching can hurt performance
 - Should have coherent regions of > 1000 pixels
 - That is only about 30×30 pixels, so still very useable!
- Don't ignore overhead of branch instructions
 - Branching over only a few instructions not worth it
- Use branching for early exit from loops
 - Save a lot of computation
- GeForce 6 vertex branching is fully MIMD
 - very small overhead and no penalty for divergent branching

Current GPGPU Limitations



- Programming is difficult
 - Limited memory interface
 - Usually “invert” algorithms (Scatter → Gather)
 - Not to mention that you have to use a graphics API...
- Limitations of communication from GPU to CPU
 - PCI-Express helps
 - GeForce 6 Quadro GPUs: 1.2 GB/s observed
 - Will improve in the near future
 - Frame buffer read can cause pipeline flush
 - Avoid frequent communication to CPU

Brook for GPUs



- A step in the right direction
 - Moving away from graphics APIs
- Stream programming model
 - enforce data parallel computing: streams
 - encourage arithmetic intensity: kernels
- C with stream extensions
 - Cross compiler compiles to HLSL and Cg
 - GPU becomes a streaming coprocessor
- See SIGGRAPH 2004 Paper and
 - <http://graphics.stanford.edu/projects/brook>
 - <http://www.sourceforge.net/projects/brook>



Examples

EG 2004

Example: Fluid Simulation



- Navier-Stokes fluid simulation on the GPU
 - Based on Stam's "Stable Fluids"
 - Vorticity Confinement step
 - [Fedkiw et al., 2001]
- Interior obstacles
 - Without branching
- Fast on latest GPUs
 - ~120 fps at 256x256 on GeForce 6800 Ultra
- Available in NVIDIA SDK 8.0



"Fast Fluid Dynamics Simulation on the GPU", Mark Harris. In *GPU Gems*.

Fluid Dynamics



- Solution of Navier-Stokes flow equations
 - Stable for arbitrary time steps
 - [Stam 1999], [Fedkiw et al. 2001]
- Fast on latest GPUs
 - 100+ fps at 256x256 on GeForce 6800 Ultra
- See “Fast Fluid Dynamics Simulation on the GPU”
 - Harris, *GPU Gems*, 2004



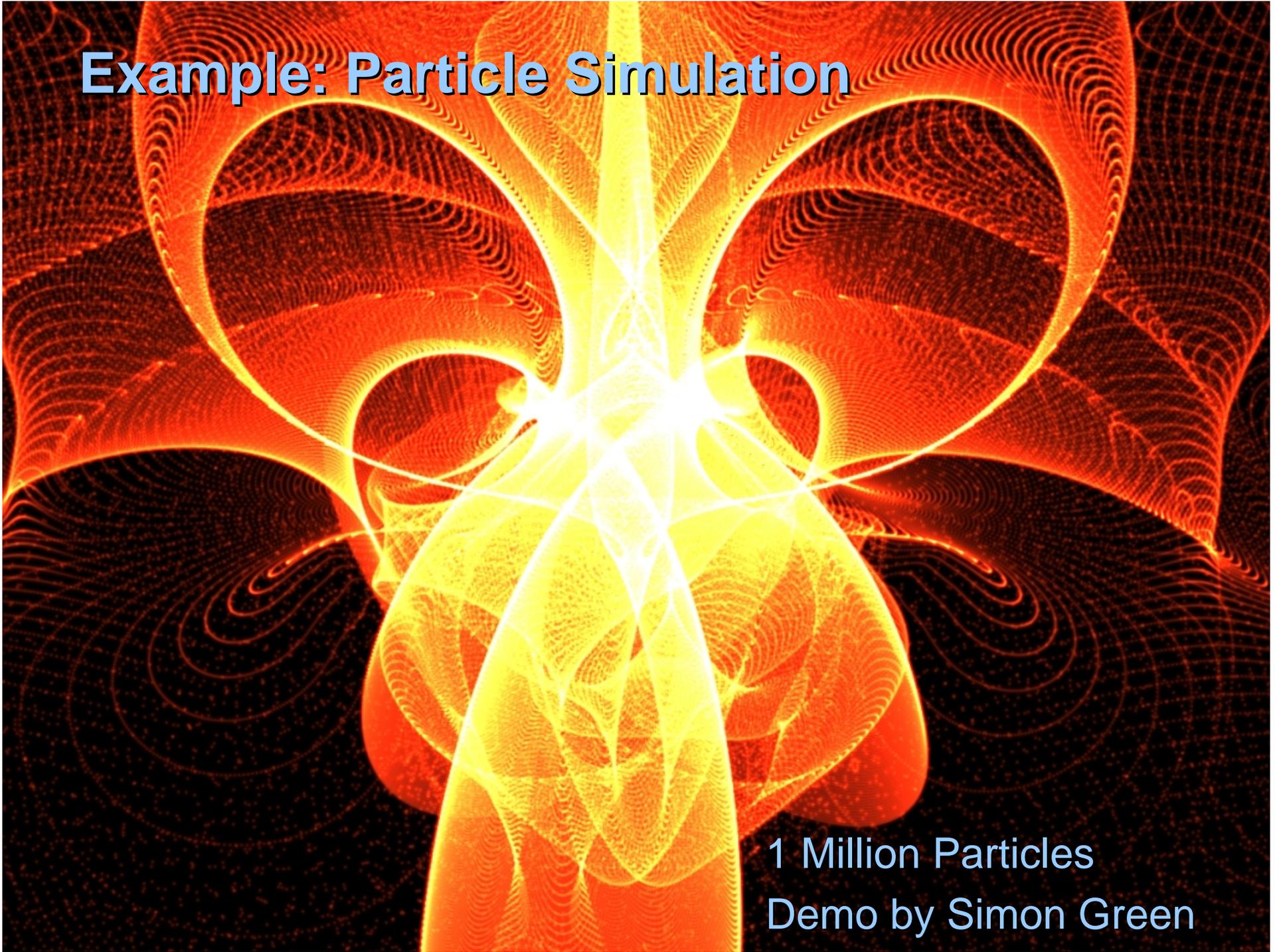
Fluid Simulator Demo



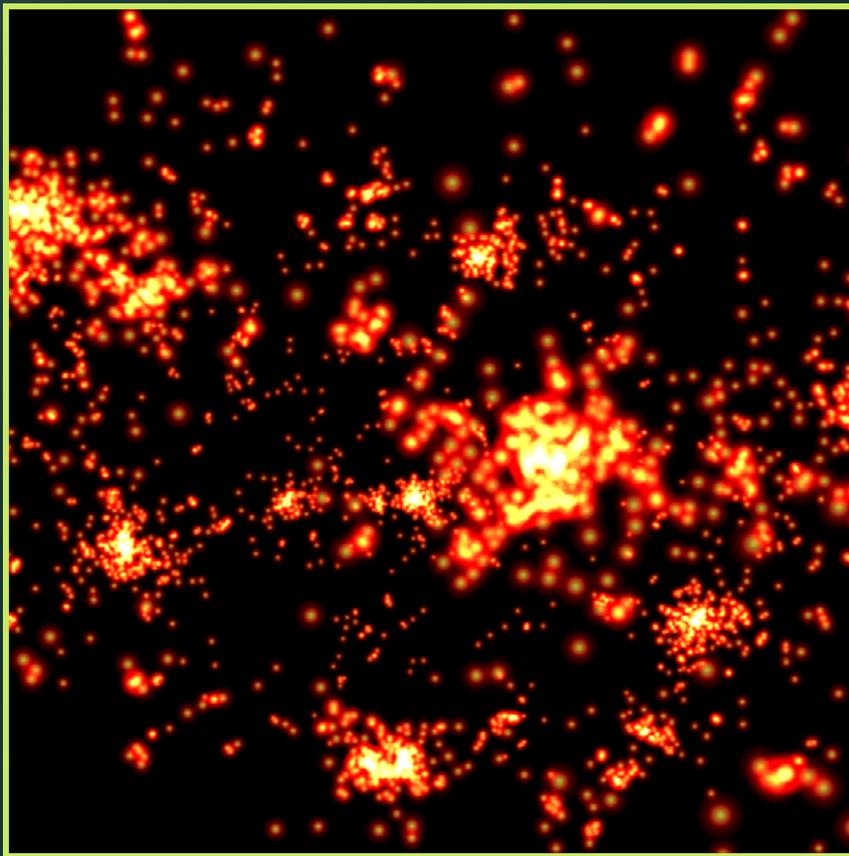
Available in NVIDIA SDK: <http://developer.nvidia.com>

Example: Particle Simulation

1 Million Particles
Demo by Simon Green



Example: N-Body Simulation



- Brute force ☹
- $N = 4096$ particles
- N^2 gravity computations
- 16M force comps. / frame
- ~25 flops per force
- 17+ fps
- 7+ GFLOPs sustained

Nyland et al., GP² poster

The Future



- Increasing flexibility
 - Always adding new features
 - Improved vertex, fragment languages
- Easier programming
 - Non-graphics APIs and languages?
 - Brook for GPUs
 - <http://graphics.stanford.edu/projects/brookgpu>

The Future



- Increasing performance
 - More vertex & fragment processors
 - More flexible with better branching
- GFLOPs, GFLOPs, GFLOPs!
 - Fast approaching TFLOPs!
 - Supercomputer on a chip
- Start planning ways to use it!

More Information



- GPGPU news, research links and forums
 - www.GPGPU.org
- developer.nvidia.org
- Questions?
 - mharris@nvidia.com

New Functionality Overview



- Vertex Programs
 - Vertex Textures: gather
 - MIMD processing: full-speed branching
- Fragment Programs
 - Looping, branching, subroutines, indexed input arrays, explicit texture LOD, facing register
- Multiple Render Targets
 - More outputs from a single shader
 - Fewer passes, side effects

New Functionality Overview



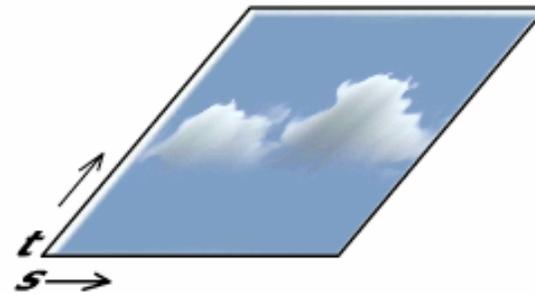
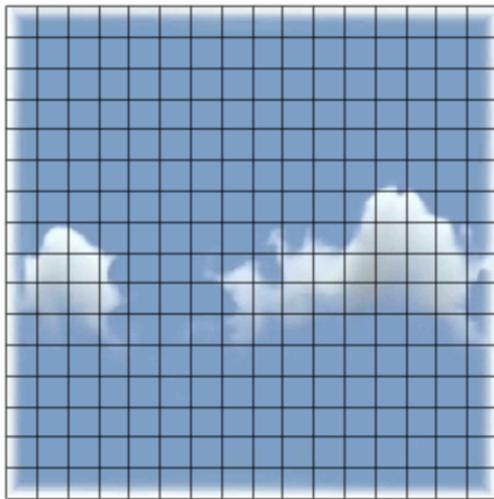
- VBO / PBO & Superbuffers
 - Feedback texture to vertex input
 - Render simulation output as geometry
 - Not as flexible as vertex textures
 - No random access, no filtering
 - Demos
- PCI-Express
 - Higher GPU \leftrightarrow CPU bandwidth

CPU-GPU Analogies



CPU

GPU



Stream / Data Array = Texture
Memory Read = Texture Sample

CPU-GPU Analogies



CPU

advect

GPU

```
for (int j = 1; j < height - 1; ++j)
{
    for (int i = 1; i < width - 1; ++i)
    {
        // get velocity at this cell
        Vec2f v = grid(x, y);

        // trace backwards along velocity field
        float x = (i - (v.x * timestep / dx));
        float y = (j - (v.y * timestep / dy));

        grid(x, y) = grid.bilerp(x, y);
    }
}
```

C++

```
void advect(float2 uv : WPOS,
           out float4 xNew : COLOR,

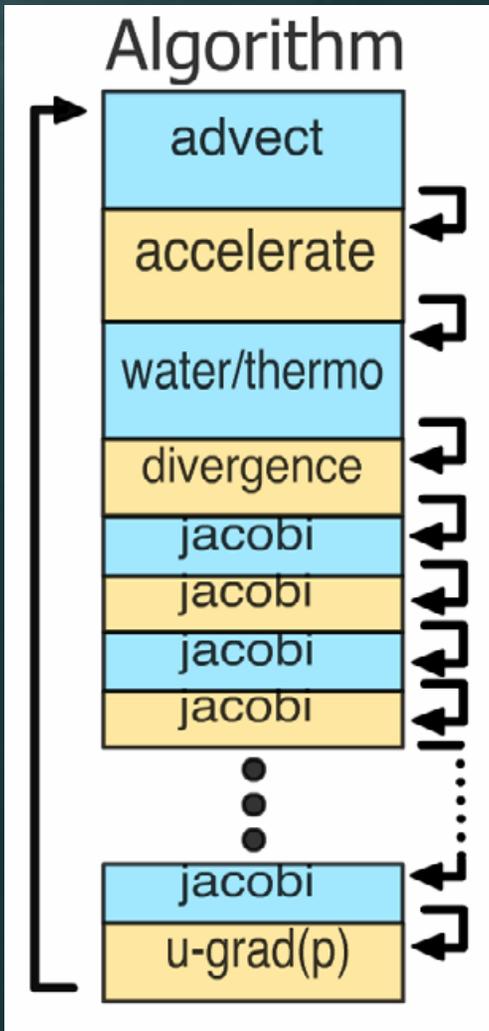
           uniform float dt, // timestep
           uniform float dx, // grid scale
           uniform samplerRECT u, // velocity
           uniform samplerRECT x) // state
{
    // trace backwards along velocity field
    float2 pos = uv - dt * f2texRECT(u, uv) / dx;

    xNew = f4texRECTbilerp(x, pos);
}
```

Cg

Loop body / kernel / algorithm step = Fragment Program

Feedback



- Each algorithm step depend on the results of previous steps
- Each time step depends on the results of the previous time step

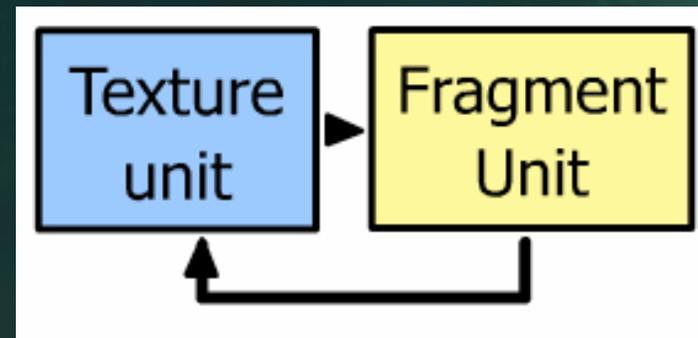
CPU-GPU Analogies



CPU

GPU

·
·
·
Grid[i][j] = x;
·
·
·



Array Write

=

Render to Texture

Navier-Stokes Equations



- Describe flow of an incompressible fluid

$$\frac{\partial \mathbf{u}}{\partial t} = -(\mathbf{u} \cdot \nabla) \mathbf{u} - \frac{1}{\rho} \nabla p - \nu \nabla^2 \mathbf{u} + \mathbf{f}$$

Advection

Pressure
Gradient

Diffusion
(viscosity)

External Force

$$\nabla \cdot \mathbf{u} = 0$$

← Velocity is divergence-free

Fluid Algorithm



● Break it down [Stam 1999]:

● Advect:

● Add forces:

● Solve for pressure:

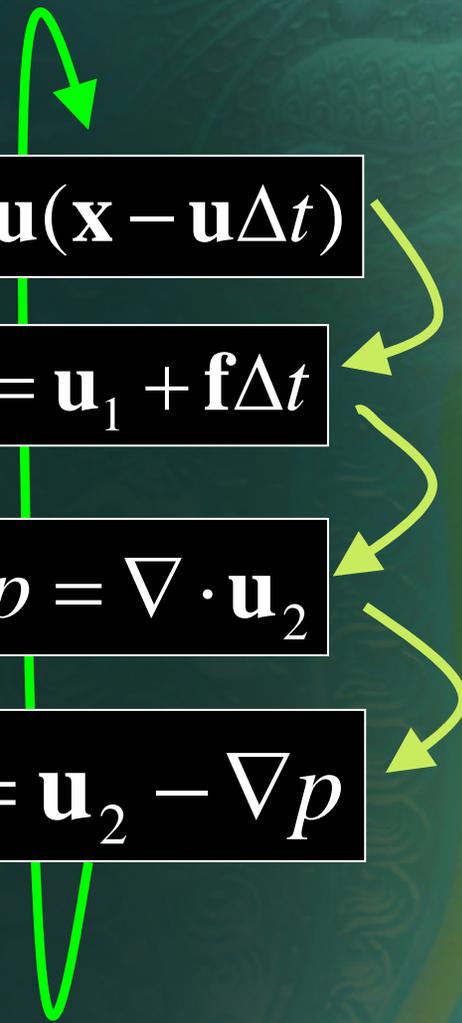
● Subtract pressure gradient:

$$\mathbf{u}_1 = \mathbf{u}(\mathbf{x} - \mathbf{u}\Delta t)$$

$$\mathbf{u}_2 = \mathbf{u}_1 + \mathbf{f}\Delta t$$

$$\nabla^2 p = \nabla \cdot \mathbf{u}_2$$

$$\mathbf{u}^* = \mathbf{u}_2 - \nabla p$$

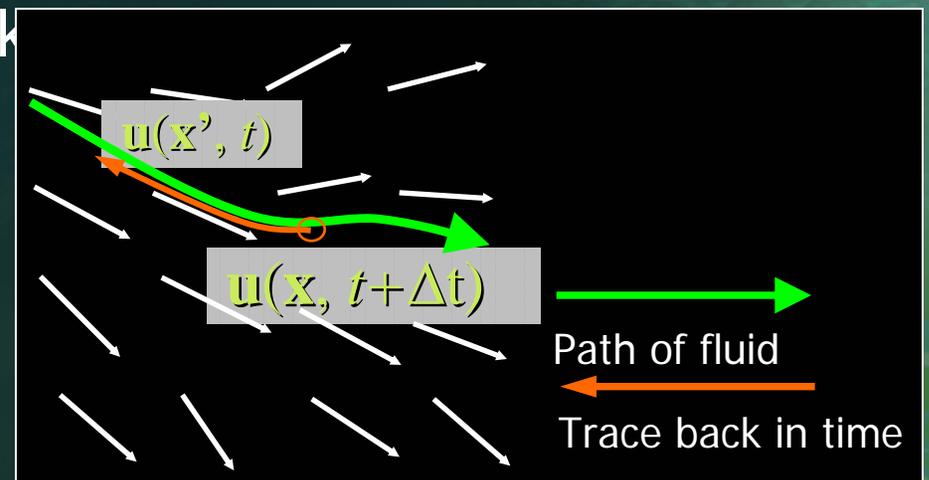


Advection



- Advection: quantities in a fluid are carried along by velocity
- Follow velocity field back to position

$$\mathbf{u}_1 = \mathbf{u}(\mathbf{x} - \mathbf{u}\Delta t)$$



```
float2 pos =  
    coords - delta_t * tex(u, coords);  
  
uNew = texBilerp(u, pos);
```

Poisson-Pressure Solution



$$\nabla^2 p = \nabla \cdot \mathbf{u}_2$$

- Discretize equation, solve using iterative solver
 - Jacobi, multigrid, conjugate gradient, etc.
 - Jacobi easy on GPU, but others possible too
 - Demo uses Jacobi iteration (50 iterations by default)
- Compute divergence field, then repeatedly

```
float pL = tex(pressure, coords + float2(-1, 0));
float pR = tex(pressure, coords + float2( 1, 0));
float pB = tex(pressure, coords + float2( 0,-1));
float pT = tex(pressure, coords + float2( 0, 1));

float div = tex(divergence, coords);

pNew = 0.25 * (pL + pR + pB + pT - delta2 * div);
```