Data parallelism and graphics

Interactivity

- Interactive rendering applications
 - Video game, architectural modeling, 3D visualization (as opposed to offline rendering, e.g., movies)
- Requirement: Must be interactive
- How fast is enough?
- Monitors refresh at 60 Hertz (60 Hz)
 - Renders 60 images in each second
- Display 6 frames per second (6 FPS)
 - Each image will be repeated 10 times!
 - Testing shows this is the bare minimum
 http://www.realtimerendering.com/udacity/?load=demo/unit1-fps.js

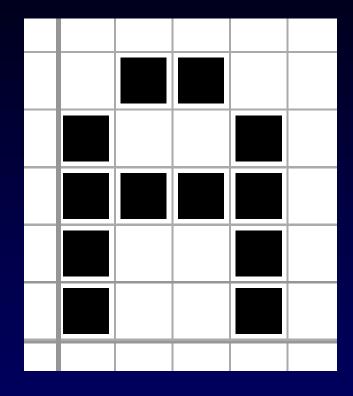
Real-time graphics

- Demands are often much higher
- 30-60 FPS are common thresholds
 - Tied to the 60Hz refresh rate
 Each image being displayed once or twice
 - 60 FPS is upper limit
- If game takes 100ms to generate the frame
 - 10 FPS result
 - Display will still refresh at 60Hz but it will overwrite with same image

Screen resolution

- Pixel: A point in your screen
- Holds an RGB color value

Higher screen resolution
 leads to larger number of pixels
 leads to better image
 leads to lower higher processing cost



leads to lower FPS!

Pixel throughout

- A screen resolution of 1024 by 768 (modest!)
- A frame rate of 60 FPS
- How many pixels are we drawing per second?
 - A: 47,185,920
 - In each second the GPU has to calculate the color of a pixel over 47 million times!
 - A lot of computation
 - And that's a lower bound
 - Objects can overlap and be drawn on top of others

How can GPUs be so fast?!

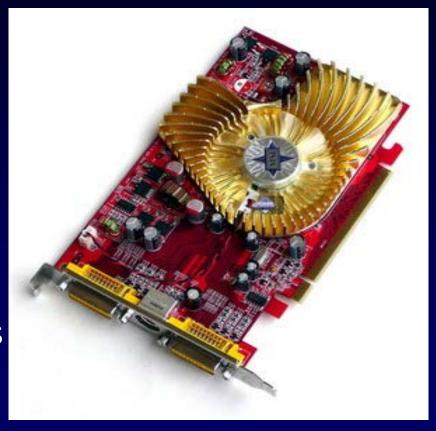
A: Parallel processing

Data Parallelism for Graphics

- GPUs are designed for graphics
 - Most popular consumer need for highly parallel computation
- GPUs process independent vertices & pixels
 - Temporary registers are zeroed
 - No shared or static data (not in graphics APIs)
 - No read-modify-write buffers
- Data-parallel processing
 - GPUs architecture is ALU-heavy
 - Multiple vertex & pixel pipelines

Today's GPUs

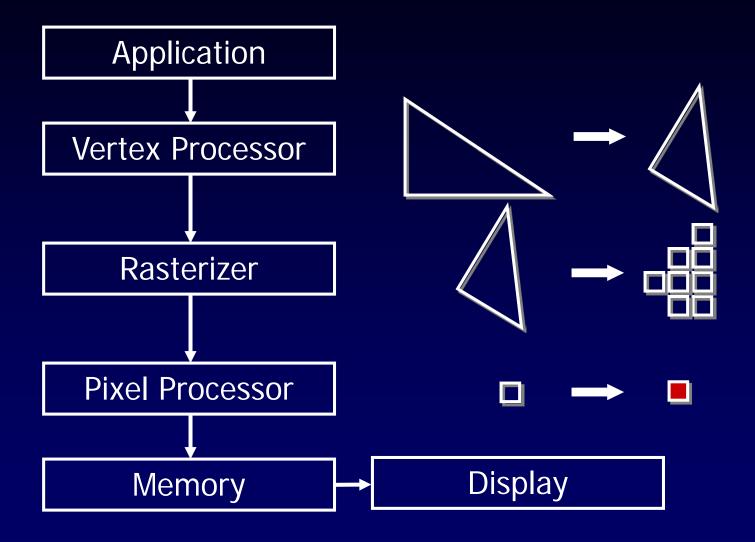
- Efficient architecture
- 1000+ parallel processors!
 - Working on vertices
 - Working on pixels
- SIMD processing
 - Must execute same code
 - Ok for graphics applications



Improving at faster rate than CPUs



GPU processing at a glance

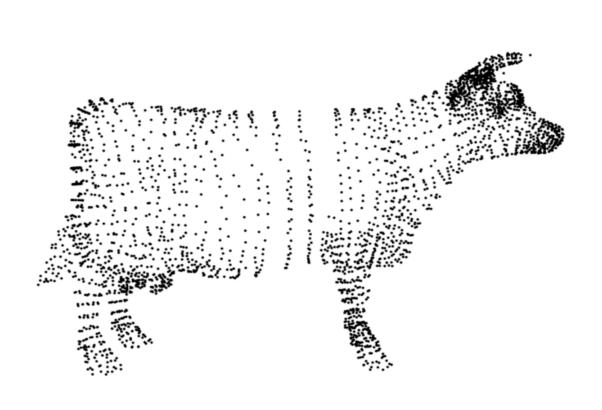


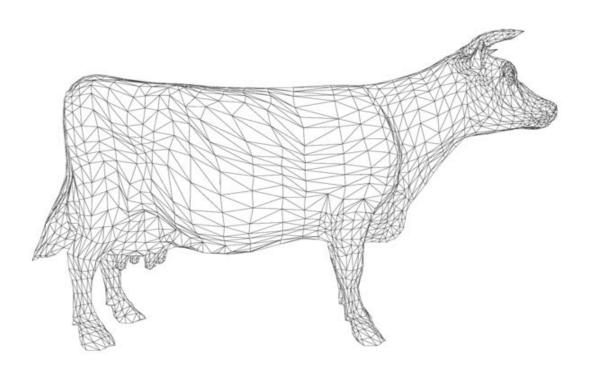
Vertex Processor

From object coordinates to screen coordinates

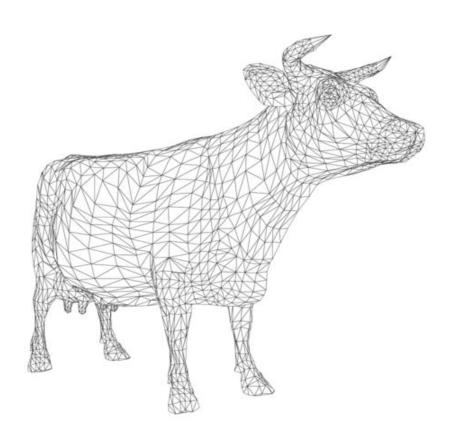
Geometry

- Composed of two parts
 - Vertex data in a vertex array/buffer
 - Primitives in an index/element array/buffer
- Vertices must include a position
 - Color and other optional parameters
- Primitives reference one or more vertices
 - Triangles, lines, points
- Geometry transformations act on vertices
 - And as a consequence change primitives too

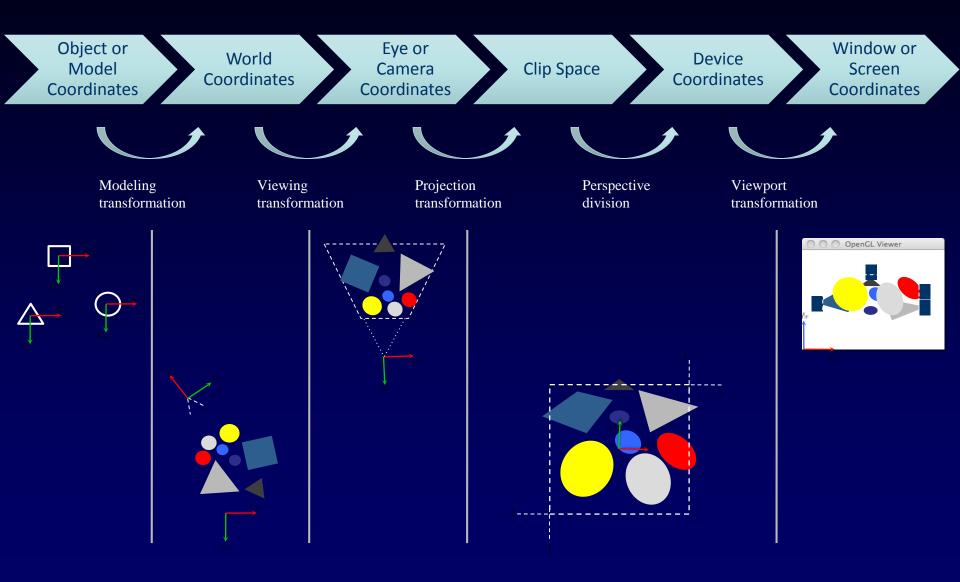




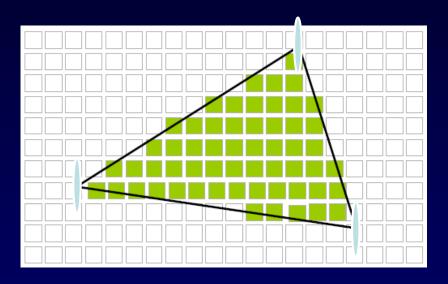




Vertex transformation



Rasterizer



From vertices and primitives to pixels

Pixel processor

Coloring the geometry
Use a popular illumination model



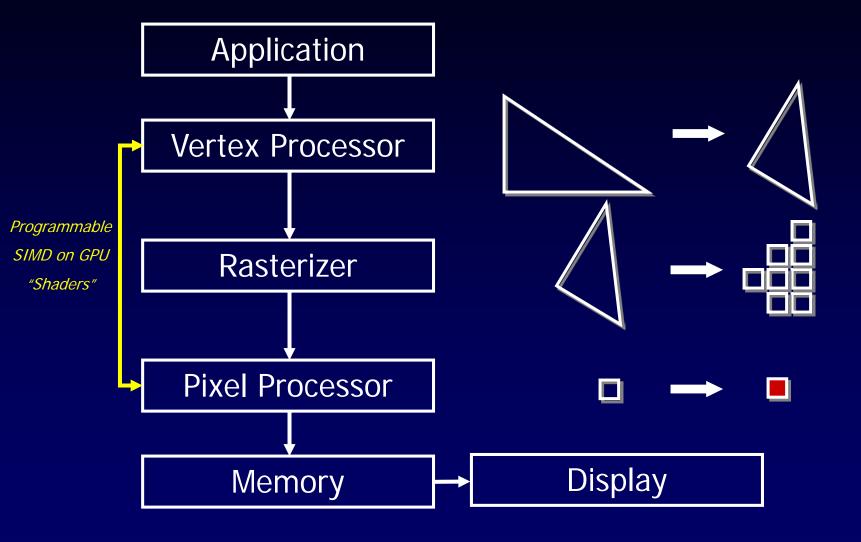
Pixel processor

Applying texture images





Programmability



Unified rendering pipeline

- No longer has dedicated vertex and pixel processors
- All programmable computation shares the same set of processors
- A complex scheduler determines their allocation when rendering a frame

GPU "shaders" processes vertices and pixels

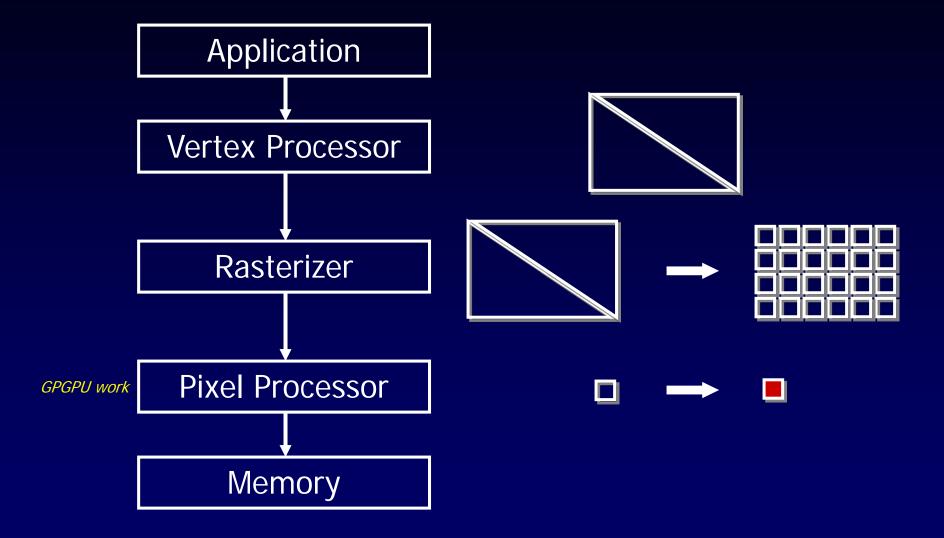
GPU Shading languages

- GLSL (OpenGL standard library)
- HLSL (Microsoft DirectX)
- CG (Nvidia proprietary)

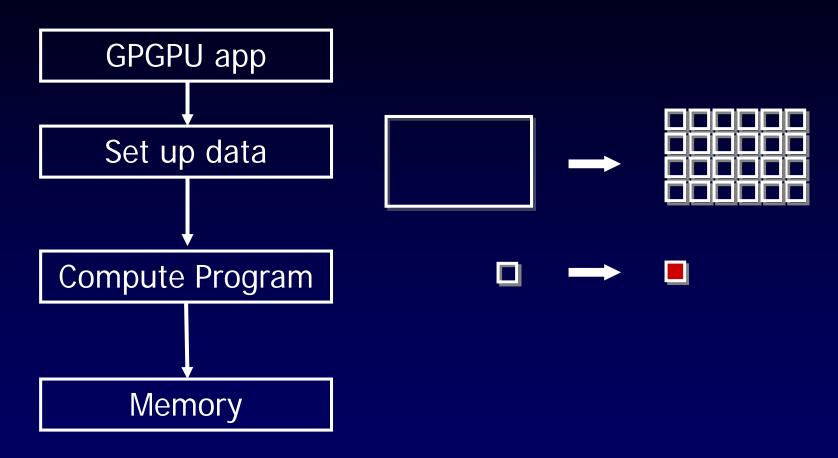
- All C-like languages
- 2D-4D vector and matrix data types
 - Enough to represent geometry and color



GPGPU usage scenario

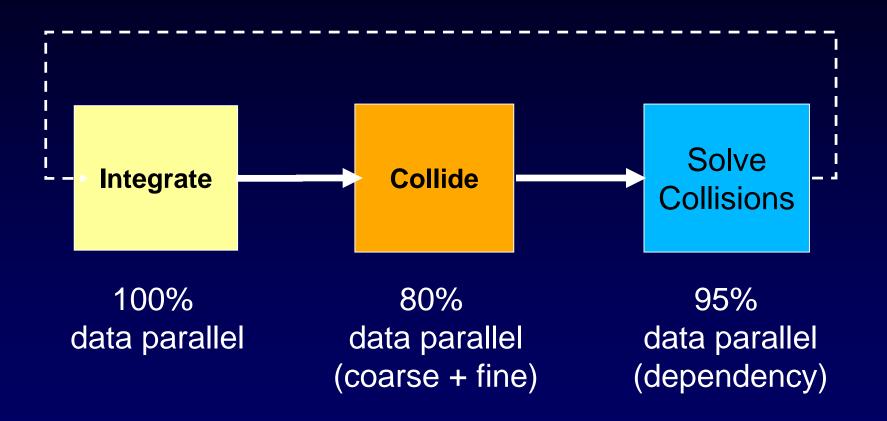


GPGPU Programming Model



Physical simulations on the GPU

Physics is a data parallel task



Mapping concepts to GPU

- Properties:
 - Algorithm must be data-parallel
 - Computations should ideally be local

Now, how do we map it?

Data Streams & Kernels

Streams

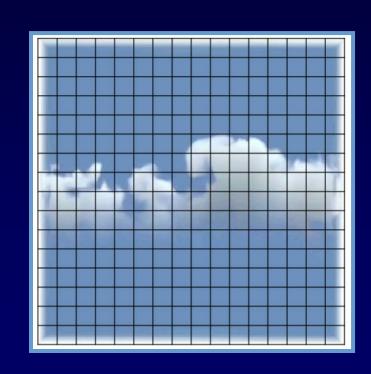
- Collection of records requiring similar computation
 - Vertex positions, Voxels, etc.
- Provide data parallelism

Kernels

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

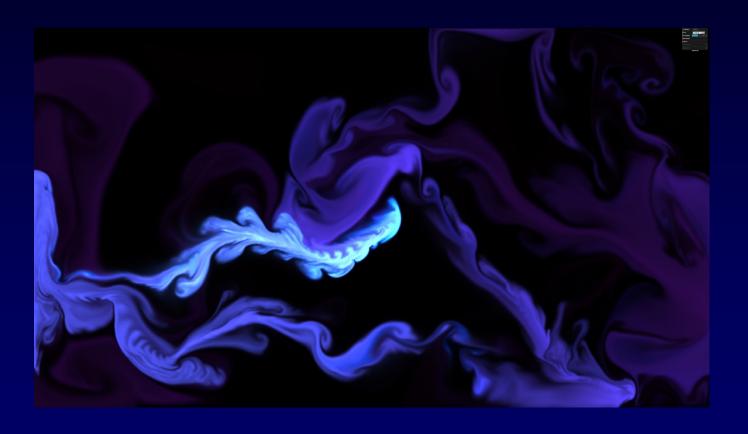
Example: Fluid simulation

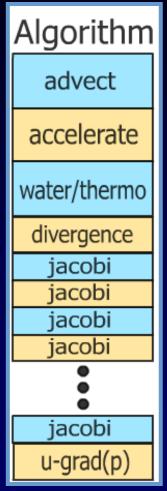
- Uses a simulation grid
- Common GPGPU rendering pipeline computation style
 - Textures represent computational grids = streams
- Many computations map to grids
 - Matrix algebra
 - Image & Volume processing
 - Global Illumination
 - Physically-based simulation
- Non-grid streams can be mapped to grids



Fluid simulation

Navier stokes equations for fluid simulation





Kernels

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));

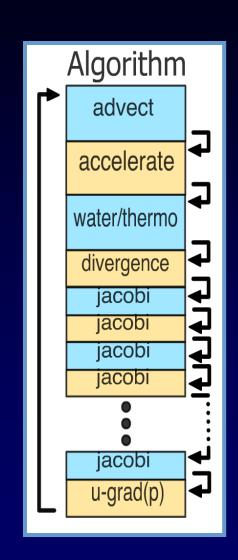
        grid (x,y) = grid bilerp (x, y);
}
</pre>
```

Kernel / loop body / algorithm step = Fragment Program

Feedback

 Each algorithm step depends on the results of previous steps

 Each time step depends on the results of the previous time step



GPU Simulation

- Algorithm steps are pixel shaders
 - Computational kernels
- Current state is stored in textures
- Feedback via render to texture



Invoking Computation

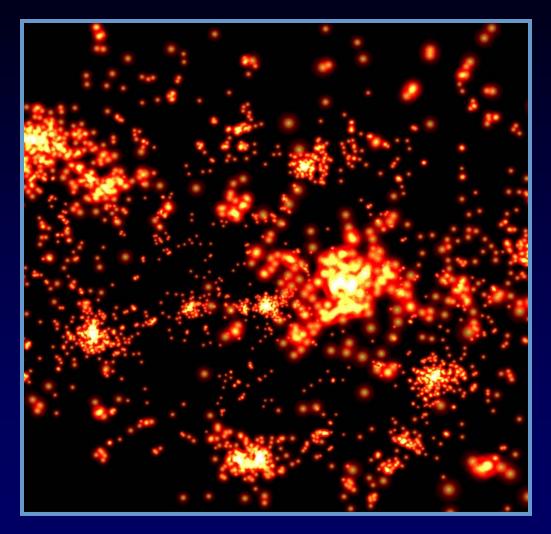
- Full screen quadrilateral invokes computation
- So, rasterization = kernel invocation

Examples:

https://www.youtube.com/watch?v=7EOjAdmURY4

https://haxiomic.github.io/GPU-Fluid-Experiments/html5/?q=Medium

Example: N-Body Simulation

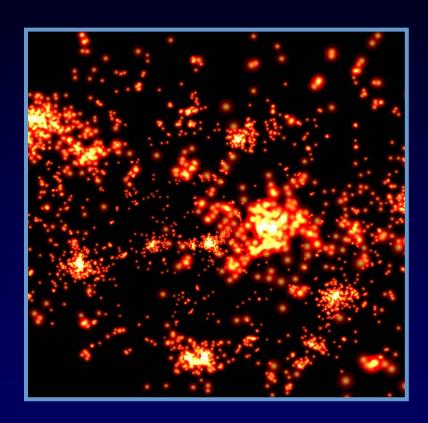


https://www.youtube.com/watch?v=CPuVfiWLIHI

GPU computation

- Brute force 😊
- N = 8192 bodies
- N² gravity computations

- 64M force comps. / frame
- ~25 flops per force

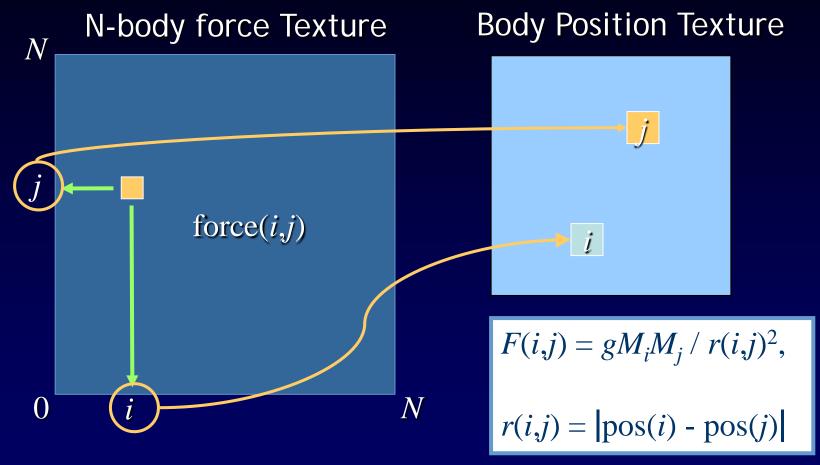


Nyland, Harris, Prins, GP² 2004 poster

Computing Gravitational Forces

- Each body attracts all other bodies
 - N bodies, so N² forces
- Draw into an NxN buffer
 - Pixel (i,j) computes force between i and j
 - Very simple fragment program
 - More than 8192 bodies makes it trickier
 - Limited by max buffer size...

Computing Gravitational Forces



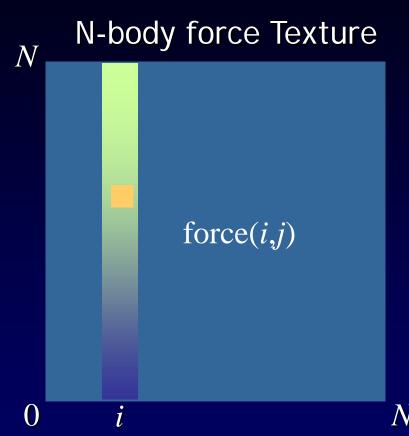
Force is proportional to the inverse square of the distance between bodies

Computing Gravitational Forces

```
float4 force(float2 ij : WPOS,
     uniform sampler2D pos) : COLOR0
  // Pos texture is 2D, not 1D, so we need to
  // convert body index into 2D coords for pos tex
  float4 iCoords = getBodyCoords(ij);
  float4 iPosMass = texture2D(pos, iCoords.xy);
  float4 jPosMass = texture2D(pos, iCoords.zw);
  float3 dir = iPos.xyz - jPos.xyz;
  float r2 = dot(dir, dir);
  dir = normalize(dir);
  return dir * q * iPosMass.w * jPosMass.w / r2;
```

Computing Total Force

- Have: array of (i,j) forces
- Need: total force on each particle i
 - Sum of each column of the force array
- Can do all N columns in parallel



Parallel Reduction

Parallel Reductions

1D parallel reduction:



- add two Itogether
- repeated
- Unkil/we'r texels

texels //x(///2) //x(///4) //x1 le row of



Requires log₂ / / steps

Update Positions and Velocities

- Now we have a 1-D array of total forces
 - One per body
- Update Velocity
 - $u(i,t+dt) = u(i,t) + F_{total}(i) * dt$
 - Simple pixel shader reads previous velocity and force textures, creates new velocity texture
- Update Position
 - -x(i, t+dt) = x(i,t) + u(i,t) * dt
 - Simple pixel shader reads previous position and velocity textures, creates new position texture

PhysX

- Nvidia middle-ware SDK
- Works on any CUDA-ready GPU
- Supports rigid body dynamics, soft body dynamics, vehicle dynamics, particles, volumetric fluid simulation, and cloth simulation

https://www.youtube.com/watch?v=6vipmar3wS4

https://www.youtube.com/watch?v=pEX13W-luLA

Optimize by force: Dual and Triple GPU Physics

Second (or third) GPU can be used for graphics or physics simulation, or share the primary card



