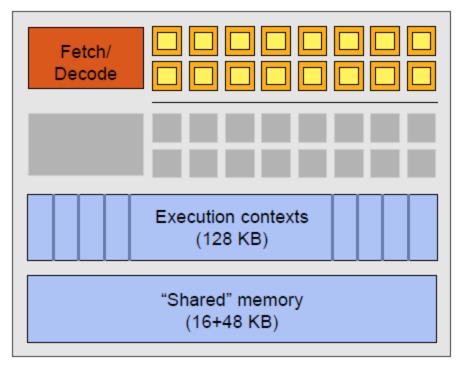
Topic for today:

Graphical Processing Units (GPUs)

Graphical Processing Unit

A GPU is a heterogeneous chip multi-processor (highly tuned for graphics)



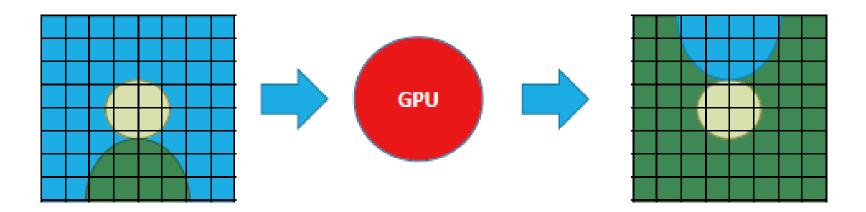
Source: Fermi Compute Architecture Whitepaper CUDA Programming Guide 3.1, Appendix G

GPU-accelerated computing

Use of a graphics processing unit (GPU) together with a CPU to accelerate scientific, analytics, engineering, consumer, and enterprise applications.

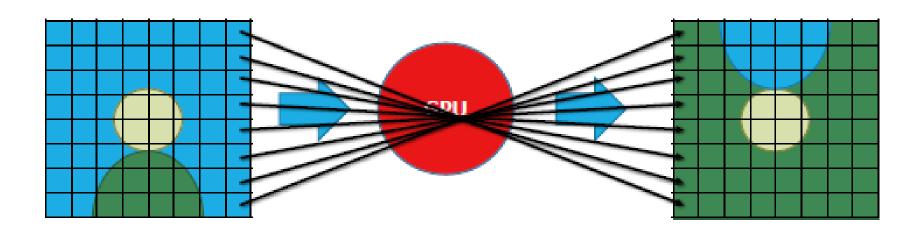
Graphics Workload

Streaming computation on pixels



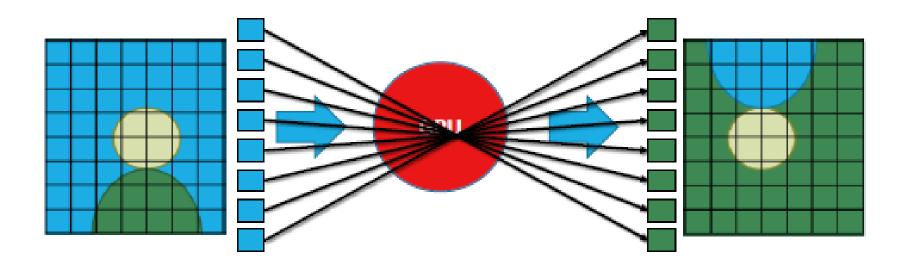
Graphics Workload

Identical, Streaming computation on pixels



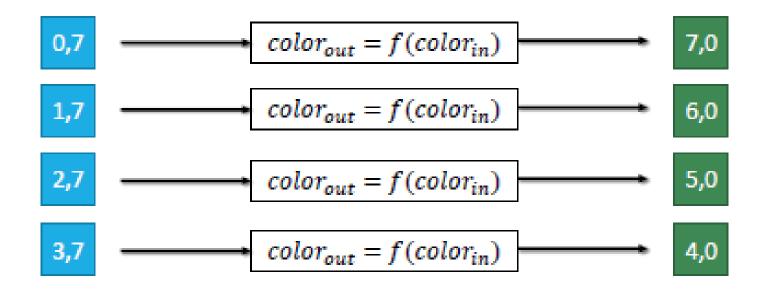
Graphics Workload

Identical, Independent, Streaming computation on pixels



Generalize: Data Parallel Workloads

Identical, Independent computation on multiple data inputs



Programming the GPU (NVIDIA)

- Heterogeneous execution model
 - CPU is the *host*, GPU is the *device*
- Develop a C-like programming language for GPU
- Unify all forms of GPU parallelism as CUDA thread
- Programming model is "Single Instruction Multiple Thread"

Threads and Blocks

- A thread is associated with each data element
- Threads are organized into blocks
- Blocks are organized into a grid

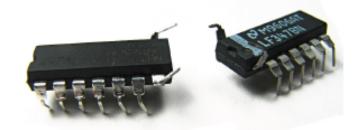
 GPU hardware handles thread management, not applications or OS

NVIDIA GPU Architecture

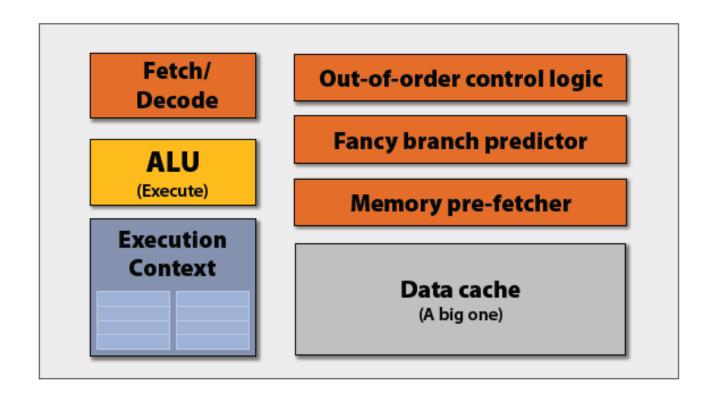
- Similarities to vector machines:
 - Works well with data-level parallel problems
 - Scatter-gather transfers
 - Mask registers
 - Large register files
- Differences:
 - No scalar processor
 - Uses multithreading to hide memory latency
 - Has many functional units, as opposed to a few deeply pipelined units like a vector processor

Why GPU?

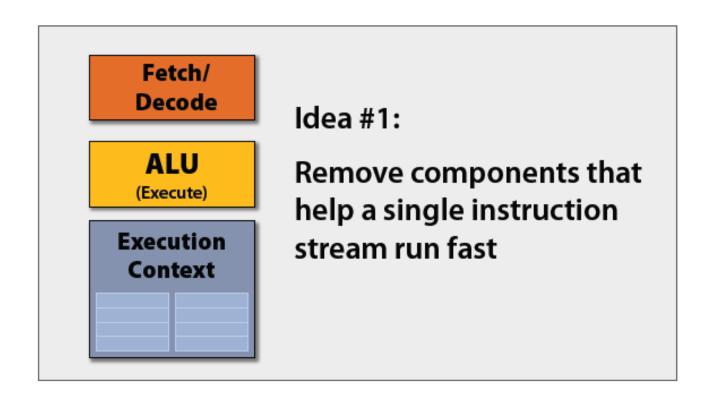
- Design target for CPUs:
 - Make a single thread very fast
 - Take control away from programmer
- GPU Computing takes a different approach:
 - Throughput matters single threads do not
 - Give explicit control to programmer



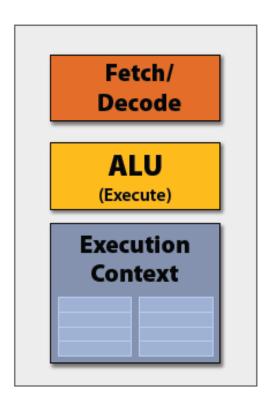
"CPU-style" Cores

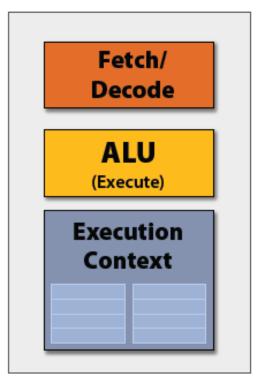


Slimming down

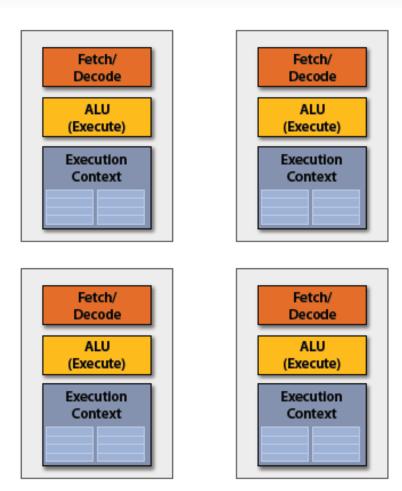


More Space: Double the Number of Cores

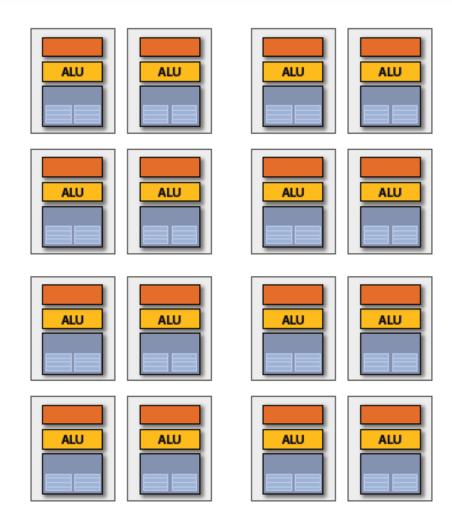




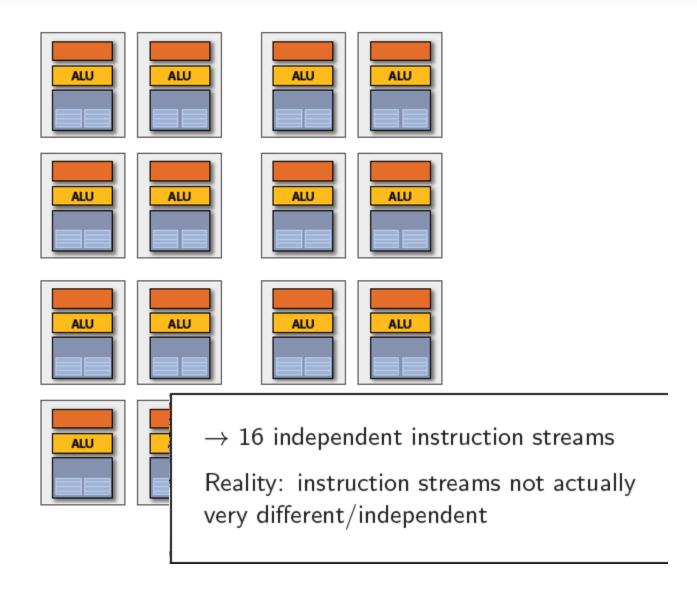
. . . again



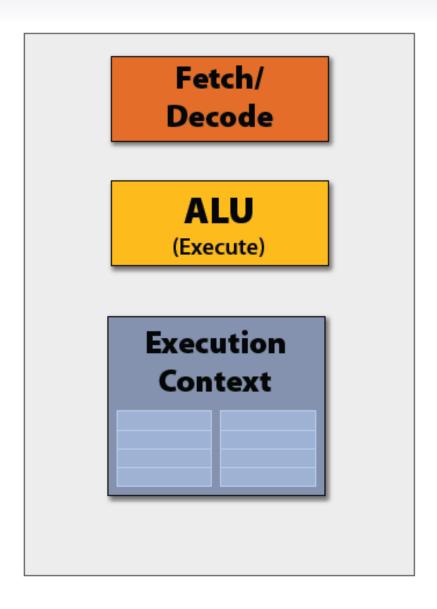
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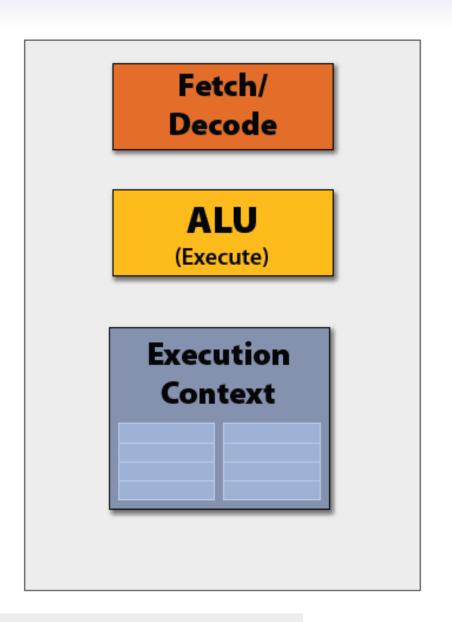
...and again



Saving Yet More Space



Saving Yet More Space

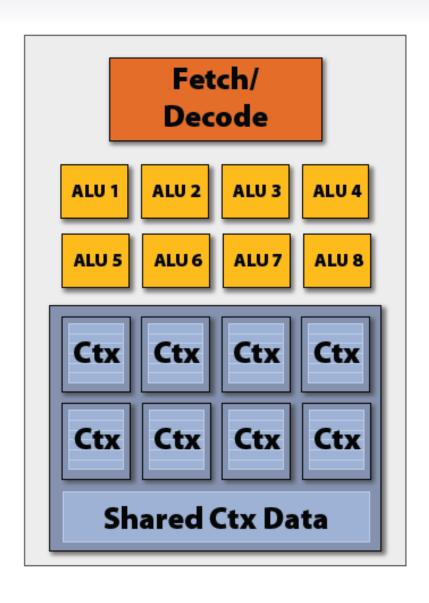


Idea #2

Amortize cost/complexity of managing an instruction stream across many ALUs

 \rightarrow SIMD

Saving Yet More Space

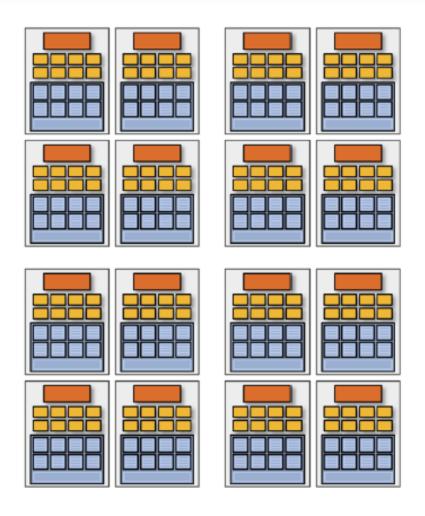


Idea #2

Amortize cost/complexity of managing an instruction stream across many ALUs

 \rightarrow SIMD

Gratuitous Amounts of Parallelism!

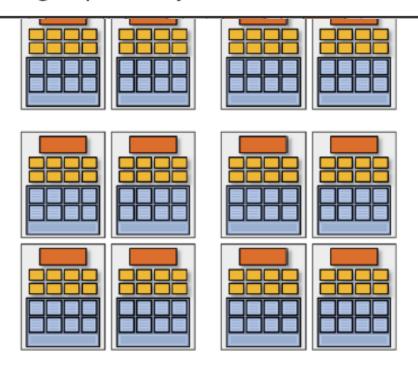


Gratuitous Amounts of Parallelism!

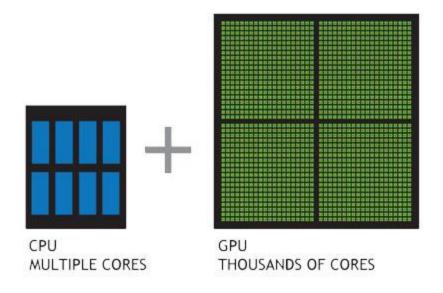
Example:

128 instruction streams in parallel

16 independent groups of 8 synchronized streams



CPU vs. GPU



https://www.youtube.com/watch?v=-P28LKWTzrI

What is CUDA?

- Scalable parallel programming model and a software environment for parallel computing
 - Minimal extensions to C/C++ environment
 - Heterogeneous serial-parallel programming model

What is OpenCL?

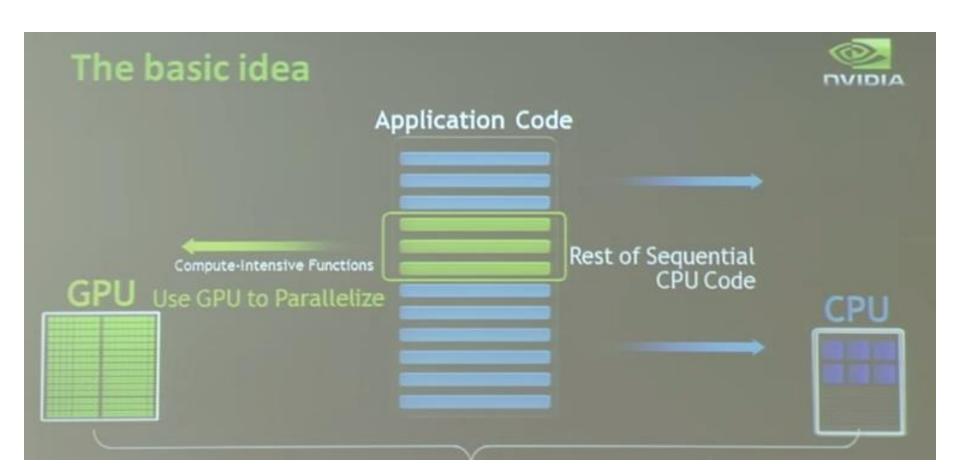
- OpenCL (Open Computing Language) is an open, royalty-free standard for general purpose parallel programming across CPUs, GPUs, and other processors.
 - Device neutral
 - Vendor neutral

GPGPU

GPU Computing: an emerging field seeking to harness GPUs for general-purpose computation.

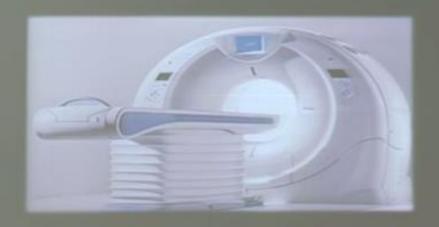
Motivation: Flexible and Precise

- Modern GPUs are deeply programmable
 - Programmable pixel, vector, video engines
 - Solidifying high-level language support
- Modern GPUs support high precision
 - 32 bit floating point throughout the pipeline
 - High enough for many (not all) applications
 - Newest GPUs have 64 bit support



Reducing Radiation from CT Scans





Est. 28,000 people/year get cancer from CT scans

UCSD: advanced CT reconstruction reduces radiation by 35-70x

CPUs: 2 hours (unusable)

CUDA: 2 minutes (clinically practical)

Operating on a Beating Heart



