GRAPHICS PROCESSING UNIT ARCHITECTURES

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CSCI B649 PARALLEL ARCHITECTURES AND PROGRAMMING

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GRAPHICS PROCESSING UNIT

• Graphics Processing Unit (GPU)

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INTRO

- A specialized circuit designed to rapidly manipulate and alter memory
- Accelerate the building of images in a frame buffer intended for output to a display
- GPU -> General Purpose Graphics Processing Unit (GPGPU)
 - A general purpose graphics processing unit as a modified form of stream processor
 - Transforms the computational power of a modern graphics accelerator's shader pipeline into general-purpose computing power

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CONCLUSION

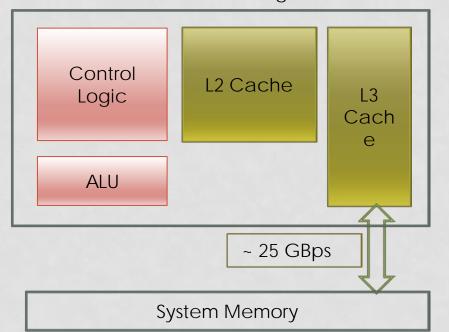
CONVENTIONAL CPU ARCHITECTURE

- Space devoted to control logic instead of ALU
- CPUs are optimized to minimize the latency of a single thread
 - Can efficiently handle control flow intensive workloads
- Multi level caches used to hide latency
- Limited number of registers due to smaller number of active threads

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• Control logic to reorder execution, provide ILP and minimize pipeline stalls

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Conventional CPU Block Diagram

A present day multicore CPU could have more than one ALU (typically < 32) and some of the cache hierarchy is usually shared across cores

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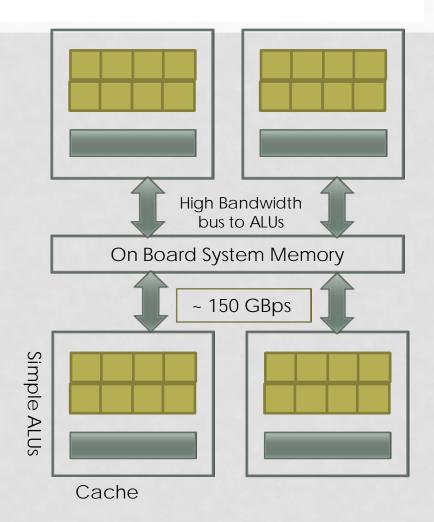
MODERN GPU ARCHITECTURE

- Generic many core GPU
 - Less space devoted to control logic and caches
 - Large register files to support multiple thread contexts
- Low latency hardware managed thread switching
- Large number of ALU per "core" with small user managed cache per core
- Memory bus optimized for bandwidth

INTRO

 ~150 GBps bandwidth allows us to service a large number of ALUs simultaneously

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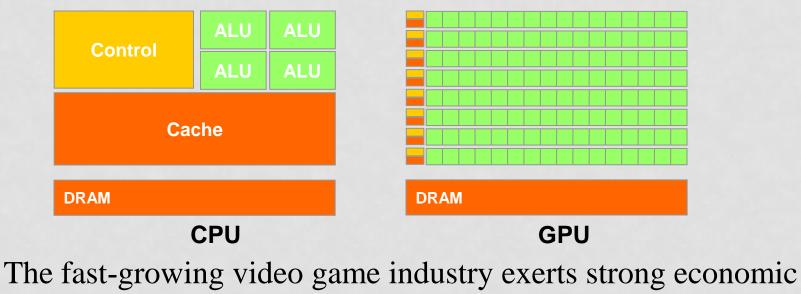
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WHAT IS BEHIND SUCH AN EVOLUTION?

- The GPU is specialized for compute-intensive, highly data parallel computation (exactly what graphics rendering is about)
 - So, more transistors can be devoted to data processing rather than data caching and flow control



pressure that forces constant innovation

INTRO

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ATI





Vendor	This Quarter Market share	last Quarter Market share	Unit Growth Qtr-Qtr	This quarter last year Market share	Growth Yr-Y
AMD	24.8%	24.2%	13.3%	21.5%	15.4%
Intel	54.4%	52.5%	14.2%	49.6%	9.7%
Nvidia	20.0%	22.5%	-1.7%	28.0%	-28.4%
Matrox	0.05%	0.1%	-11.8%	0.1%	-16.6%
SiS	0.0%	0.0%	0.0%	0.2%	- 100.0%
VIA/S3	0.7%	0.8%	1.2%	0.7%	5.3%
Total	100.0%	100.0%	10.3%	100.0%	

Table 1: Market shares Courtesy of marketing and research firm Jon Peddie Research (JPR) *

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May 4 2011, Integrated GPUs, http://jonpeddie.com/press-releases/details/surprising-q1 -results-graphics-shipments-up-10.3-while-

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Market Share	This Quarter Market Share	Last Quarter Market Share	Market Share Change Qtr-Qtr	This quarter Last Year Market Share	Change Yr-Yr
AMD	40.6%	40.5%	0.1%	41.4%	-0.8%
Nvidia	59.0%	59. 1%	-0.1%	57.9%	1.1%
Others	0.4%	0.4%	0.0%	0.7%	-0.3%

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 Table 1: Market shares over time

 Courtesy of marketing and research firm Jon Peddie Research (JPR)

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May 17 2011, Discrete GPUs, http://jonpeddie.com/publications/add-in-board-report/

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Añ	ter are	Last Quarter Market Share	Market Share Change Qtr-Qtr	This quart Year Marke	©.
	<u> </u>	40.5%	0.1%	41.4	nvidia.
		59.1%	-0.1%	57.9 ⁴	
		0.4%	0.0%	0.7%	

Table 1: Market shares over time

AMD

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Courtesy of marketing and research firm Jon Peddie Research (JPR)

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May 17 2011, Discrete GPUs, http://jonpeddie.com/publications/add-in-board-report/

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• OpenCL

- Standards specification
- Compute Unified Device Architecture (CUDA) Framework
 - A general purpose parallel computing architecture
 - A new parallel programming model and instruction set architecture
 - Leverages the parallel compute engine in NVIDIA GPUs
 - Software environment that allows developers to use C as a high-level programming language

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CUDA Capable devices

• <u>http://developer.nvidia.com/cuda-gpus</u>

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- Tesla
- Quadro
- NVS
- GeForce

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Fermi Architecture (Compute capabilities 2.x)	GeForce 500 Series GeForce 400 Series	Quadro Fermi Series	Tesla 20 Series
Tesla Architecture (Compute capabilities 1.x)	GeForce 200 Series GeForce 9 Series GeForce 8 Series	Quadro FX Series Quadro Plex Series Quadro NVS Series	Tesla 10 Series
	e Entertainment	Professional Graphics	High Performance Computing

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TERMINOLOGY REVIEW

- Device = GPU = set of multiprocessor
- Warp = A scheduling unit of up to 32 threads

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- Multiprocessor = set of processors & shared memory
- Kernel = GPU program

INTRO

- Grid = array of thread blocks that execute a kernel
- Thread block = group of SIMD threads that execute a kernel and can communicate via shared memory

Memory	Location	Cached	Access	Who		
Local	Off-chip	No	Read/write	One thread		
Shared	On-chip	N/A - resident	Read/write	All threads in a block		
Global	Off-chip	No	Read/write	All threads + host		
Constant	Off-chip	Yes	Read	All threads + host		
Texture	Off-chip	Yes	Read	All threads + host		

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• Compute Capability

- Specifications and features depend on compute capability
- "backward compatible but not forward compatible"
- Main differences between 1.x and 2.x
 - Increased the number of threads per blocks
 - Packing up to 462 cores in a single chip
 - Fermi can run multiple kernels simultaneously
 - Shared memory has been increased from 16 KB to 48KB
 - The number of streaming processors in one SM have been increased to 32.
 - ECC support

INTRO

 support for C++, virtual functions, function pointers, dynamic object allocation, and C++ exception handling operations

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• CUDA Programming Guide 4.0 pg. 170-171

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• Compute Capability

- 1.x (1.0, 1.1, 1.2, 1.3)
 - A multiprocessors has 8 CUDA cores
 - 1 double-precision FP unit for double-precision FP arithmetic operations,
 - 2 special function units for single-precision FP transcendental functions (these units can also handle single-precision floating-point multiplications),
 - 1 warp scheduler.

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• Compute Capability

- 2.x (2.0, 2.1 Fermi family)
 - For devices of compute capability 2.0:
 - 32 CUDA cores for integer and FP arithmetic operations,
 - 4 special function units for single-precision FP
 - For devices of compute capability 2.1:
 - 48 CUDA cores for integer and FP arithmetic operations,
 - 8 special function units for single-precision FP
 - 2 warp schedulers.
 - GPUDirect

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- GPUDirect
 - NVIDIA GPUDirect[™] Accelerated Communication With Network And Storage Devices

Without GPUDirect

Same data copied three times:

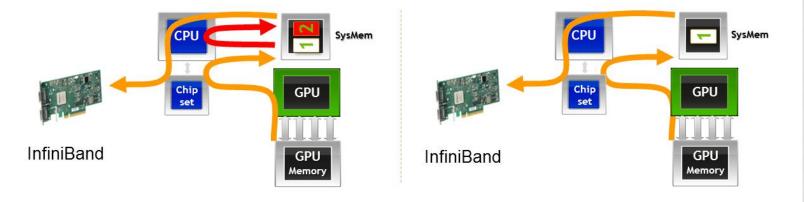
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- 1. GPU writes to pinned sysmem1
- 2. CPU copies from sysmem1 to sysmem2
- 3. InfiniBand driver copies from sysmem2

With GPUDirect

CONCLUSION

Data only copied twice Sharing pinned system memory makes sysmem-to-sysmem copy unnecessary

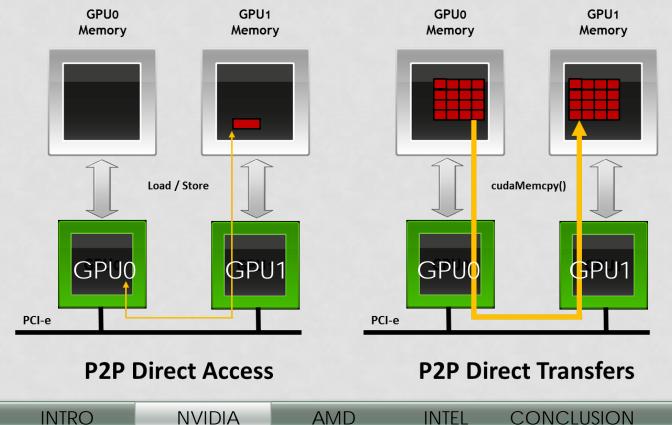


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- GPUDirect
 - NVIDIA GPUDirect Peer-To-Peer (P2P) Communication Between GPUs On The Same PCIe Bus.



G80 HARDWARE IMPLEMENTATION: A SET OF SIMD MULTIPROCESSORS

- The device is a set of 16 multiprocessors
- Each multiprocessor is a set of 32-bit processors with a Single Instruction Multiple Data architecture – shared instruction unit
- At each clock cycle, a multiprocessor executes the same instruction on a group of threads called a warp

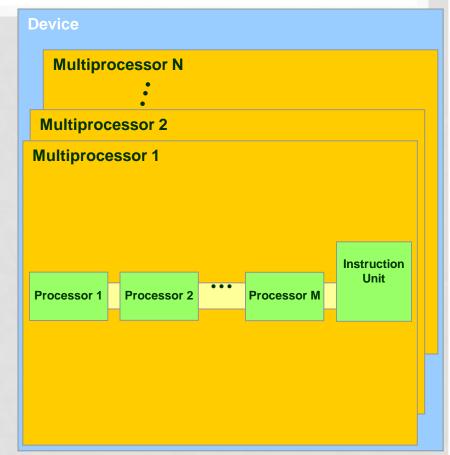
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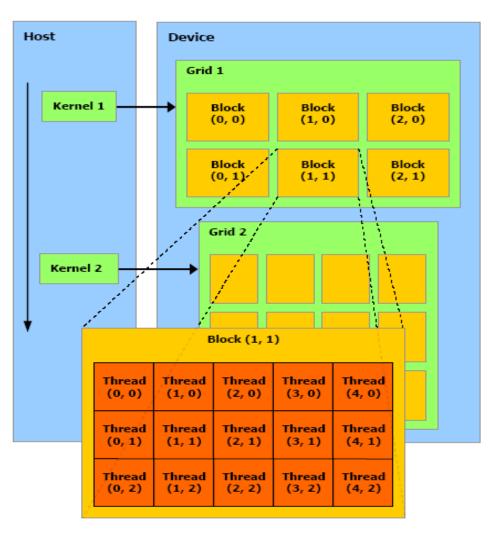
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• The number of threads in a warp is the warp size

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The host issues a succession of kernel invocations to the device. Each kernel is executed as a batch of threads organized as a grid of thread blocks

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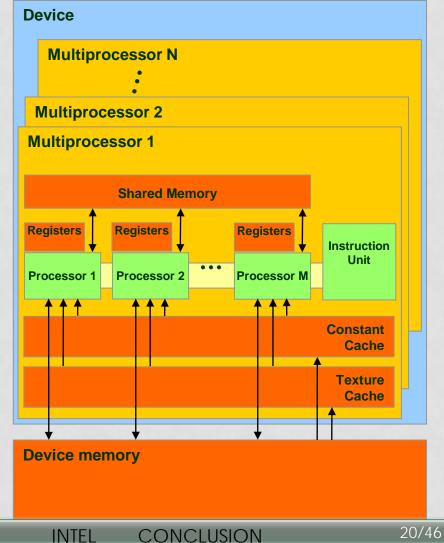
HARDWARE IMPLEMENTATION: MEMORY ARCHITECTURE

- The local, global, constant, and texture spaces are regions of device memory
- Each multiprocessor has:
 - A set of 32-bit registers per processor
 - On-chip shared memory
 - Where the shared memory space resides
 - A read-only constant cache
 - To speed up access to the constant memory space
 - A read-only texture cache

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• To speed up access to the texture memory space

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HARDWARE IMPLEMENTATION: EXECUTION MODEL (REVIEW)

- Each thread block of a grid is split into warps, each gets executed by one multiprocessor (SM)
 - The device processes only one grid at a time
- Each thread block is executed by one multiprocessor
 - So that the shared memory space resides in the on-chip shared memory
- A multiprocessor can execute multiple blocks concurrently
 - Shared memory and registers are partitioned among the threads of all concurrent blocks
 - So, decreasing shared memory usage (per block) and register usage (per thread) increases number of blocks that can run concurrently

ACCESS TIMES

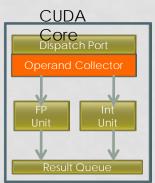
- Register dedicated HW single cycle
- Shared Memory dedicated HW single cycle
- Local Memory DRAM, no cache *slow*
- Global Memory DRAM, no cache *slow*
- Constant Memory DRAM, cached, 1...10s...100s of cycles, depending on cache locality
- Texture Memory DRAM, cached, 1...10s...100s of cycles, depending on cache locality

NVIDIA GPU – FERMI ARCHITECTURE

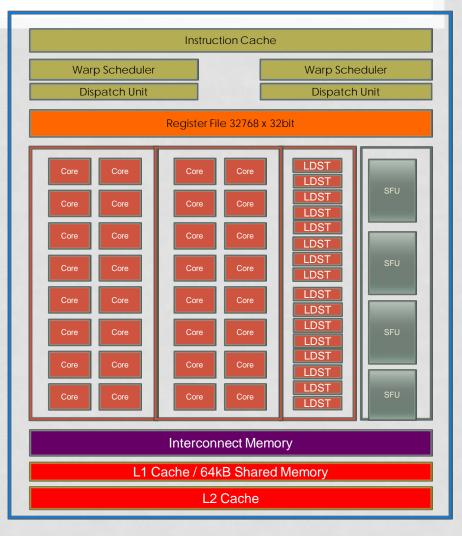
- SM executes threads in groups of 32 called warps.
 - Two warp issue units per SM
- Concurrent kernel execution
 - Execute multiple kernels simultaneously to improve efficiency
- CUDA core consists of a single ALU and floating point unit FPU

Source: NVIDIA's Next Generation CUDA Compute Architecture Whitepaper

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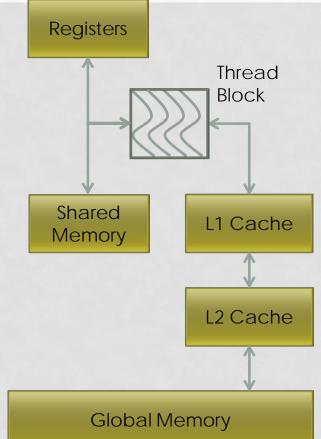
NVIDIA MEMORY HIERARCHY

- L1 cache per SM configurable to support shared memory and caching of global memory
 - 48 KB Shared / 16 KB of L1 cache
 - 16 KB Shared / 48 KB of L1 cache
- Data shared between work items of a group using shared memory
- Each SM has a 32K register bank
- L2 cache (768KB) that services all operations (load, store and texture)

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• Unified path to global for loads and stores

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WARPS

time

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- The SM schedules threads in groups of 32 parallel threads called warps.
- Each SM has
 - 2 warp scheduler
 - 2 instruction dispatch unit

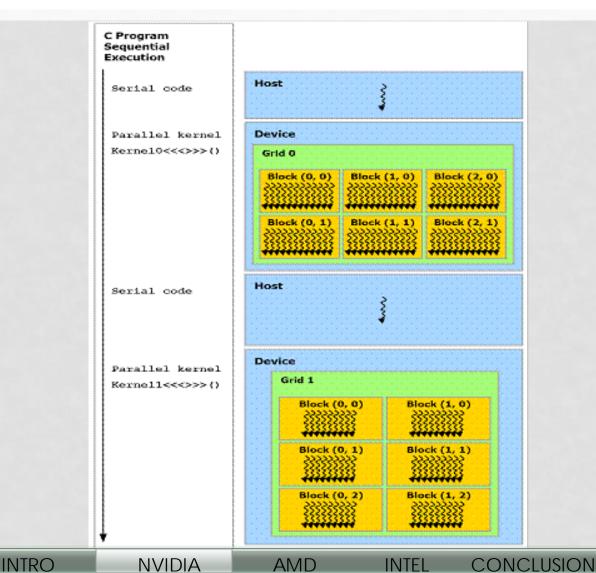
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- Allowing two warps to be issued and executed concurrently
- dual warp scheduler
 - selects two warps, and issues one instruction from each warp to a group of sixteen cores, sixteen load/store units, or four SFUs

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Warp Scheduler	Warp Scheduler
Inst Dispatch Unit	Inst Dispatch Unit
Warp 8 inst 11	Warp 9 inst 11
Warp 2 inst 42	Warp 3 inst 33
Warp 14 inst 95	Warp 15 inst 95
Warp 8 inst 12	Warp 9 inst 12
Warp 14 inst 96	Warp 3 inst 34
Warp 2 inst 46	Warp 15 inst 96

HETEROGENEOUS COMPUTING



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Performance Guidelines

- Performance optimization revolves around three basic strategies:
 - Maximize parallel execution to achieve maximum utilization;
 - Optimize memory usage to achieve maximum memory throughput;
 - Optimize instruction usage to achieve maximum instruction throughput

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- Performance Guidelines
 - Utilization
 - Application
 - Device
 - Multiprocessor Level
 - Maximize Memory Throughput
 - Device Memory Accesses
 - Global
 - Local
 - Shared
 - Texture
 - Constant

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CONCLUSION

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- Performance Guidelines
 - Utilization
 - Application
 - Device
 - Application should maximize parallel execution between the multiprocessors of a device.
 - 1.x, only one kernel can execute on a device at one time, so the kernel should be launched with at least as many thread blocks as there are multiprocessors in the device.
 - 2.x, multiple kernels can execute concurrently on a device, so maximum utilization can also be achieved by using streams to enable enough kernels to execute concurrently
 - Multiprocessor Level

- Performance Guidelines
 - Utilization
 - Application
 - Device
 - Multiprocessor Level
 - Utilization is therefore directly linked to the number of resident warps.
 - At every instruction issue time, a warp scheduler selects a warp that is ready to execute its next instruction, if any.
 - Hide the latency by having more instructions (more warps)

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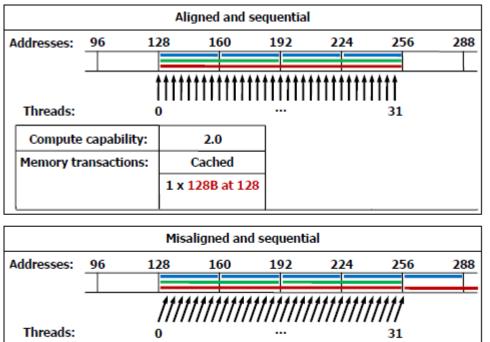
Performance Guidelines

- Maximum memory throughput
 - Minimize data transfers with low bandwidth
 - Host Device
 - Global memory
 - Memory access patterns
 - Global memory
 - Shared memory

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Performance Guidelines

- Memory access patterns
 - Global memory resides in device memory and device memory is accessed via 32-, 64-, or 128-byte memory transactions, called segments
 - maximize coalescing



Compute capability: 2.0 Memory transactions: Cached 1 x 128B at 128 1 x 128B at 256

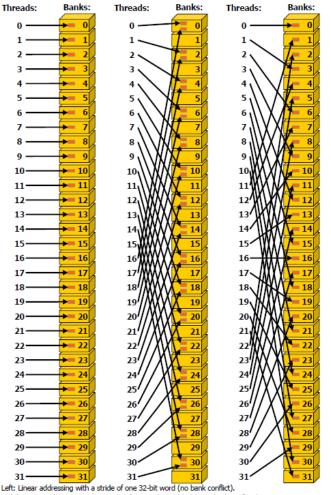
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- Performance Guidelines
 - Memory access patterns
 - Shared memory
 - Bank Conflict free access



Left: Linear addressing with a stride of one 32-bit word (no bank conflict). Middle: Linear addressing with a stride of two 32-bit words (2-way bank conflicts). Right: Linear addressing with a stride of three 32-bit words (no bank conflict).

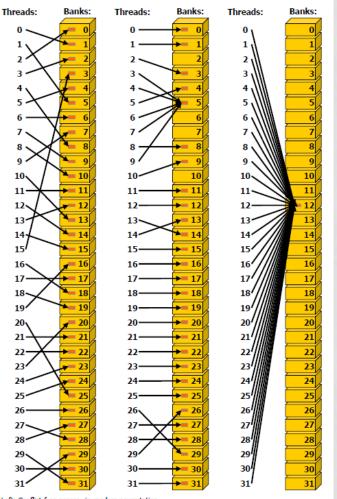
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- Performance Guidelines
 - Memory access patterns
 - Shared memory
 - Bank Conflict free access



Left: Conflict-free access via random permutation.

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Middle: Conflict-free access since threads 3, 4, 6, 7, and 9 access the same word within bank 5. Right: Conflict-free broadcast access (all threads access the same word).

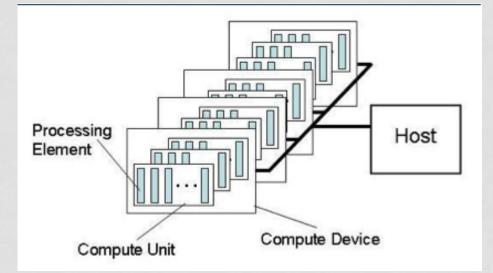
CONCLUSION

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AMD PLATFORM - OPENCL

- Individual work-items execute on a single processing element
- Processing element refers to a single VLIW core
- Multiple work-groups execute on a compute unit
- A compute unit refers to a SIMD Engine

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CONCLUSION

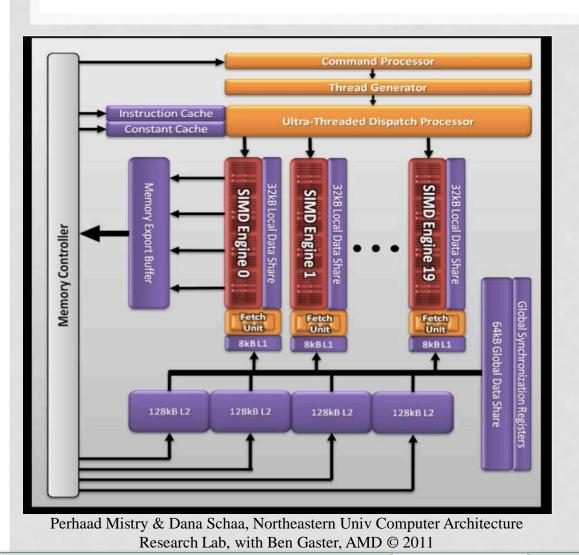
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AMD GPU - CYPRESS



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- 20 SIMD engines
- 16 SIMD units per core
- 5 multiply-adds per functional unit (VLIW processing)
- 2.72 Teraflops Single Precision
- 544 Gigaflops Double Precision

Source: Introductory OpenCL *SAAHPC2010*, Benedict R. Gaster

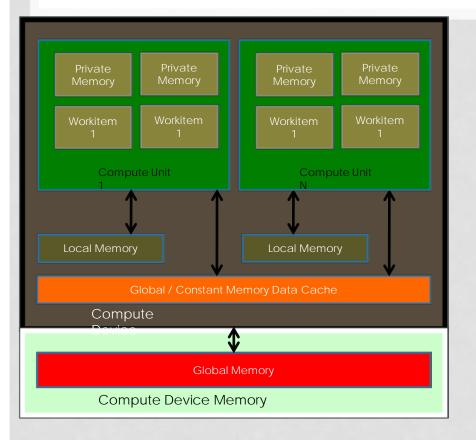
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AMD MEMORY MODEL - OPENCL



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- Subset of hardware memory exposed in OpenCL
- Local Data Share (LDS) exposed as local memory
 - Share data between items of a work group designed to increase performance
 - High Bandwidth access per SIMD Engine
- Private memory utilizes registers per work item

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• Constant Memory

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AMD FUSION

• AMD Fusion

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- is a new approach to processor design and software development, delivering powerful CPU and GPU capabilities:
 - workloads in a single-die processor called an APU(Accelerated Processing unit).
- it is the first heterogeneous multi-core processors, where it integrate multiple x86 cores and graphical processing unit (GPU) cores in a single die.
- The idea behind the Fusion is to enable a mix-and-match strategy at the silicon level, through a broad range of design initiatives.
- Leverages multi-core concept and Direct Connect architecture, enables a homogeneous programming model for all AMD products, and standardizes the coprocessor interface for on-die platform connectivity.

AMD

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WHAT IS AN APU?

• <u>http://www.youtube.com/watch?v=BihrG7DhhBM</u>



• Intel GMA

- Series of Intel integrated graphics processors built into various motherboard chip sets.
- These integrated graphics products allow a computer to be built without a separate graphics card, which can reduce cost, power consumption and noise.
- They rely on the computer's main memory for storage
 - Performance penalty
 - CPU and GPU access memory over the same bus
- In early 2007, 90% of all PCs sold had integrated graphics.
- Sandy Bridge / Ivy Bridge

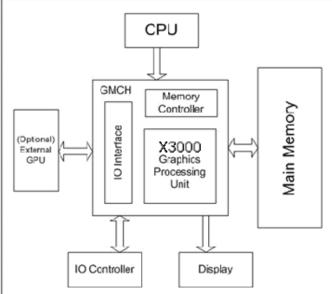
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- The built-in GPU has up to 12 execution units in Sandy Bridge
- Ivy Bridge will have next Generation Intel HD Graphics with DirectX 11, OpenGL 3.1, and OpenCL 1.1 support.

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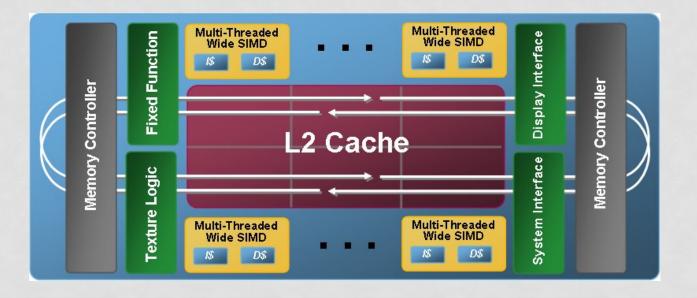
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• Larrabee

INTRO

- GPGPU chip that Intel is an effort to develop separately from integrated
- Intel claims Larrabee has a fully programmable pipeline, in contrast to current generation graphics cards which are only partially programmable.
- Project is terminated the product is moved to a co-processor for high performance computing instead of a discrete one

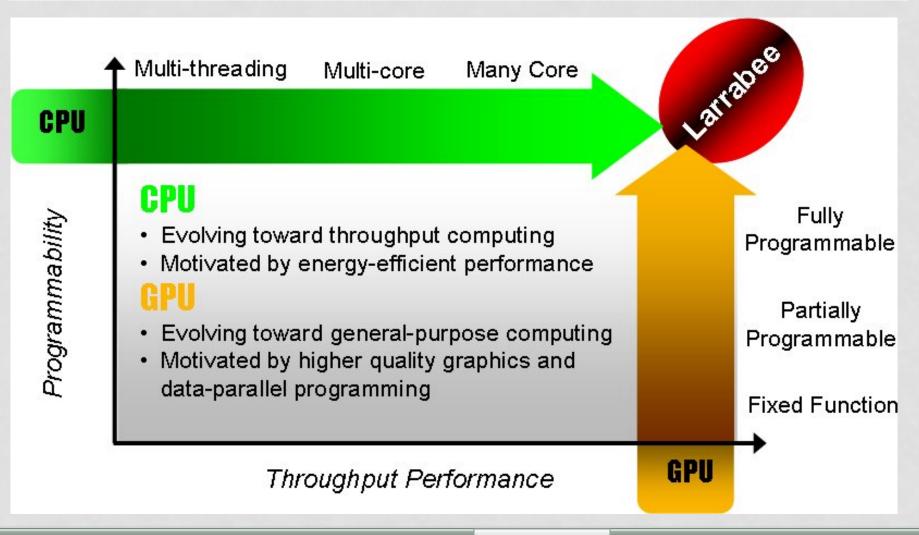


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Larrabee

INTRO

- Differences with current GPUs
 - Larrabee will use the x86 instruction set with Larrabee-specific extensions.
 - Larrabee will feature cache coherency across all its cores.
 - Larrabee will include very little specialized graphics hardware, instead performing tasks like z-buffering, clipping, and blending in software, using a tile-based rendering approach.

	ic	In-Order CPU core	In-Order CPU core		In-Order CPU core	In-C CPU		Interfaces	
	- go	Interprocessor Ring Network						erfa	
	tion L	Coherent L2 cache	Coherent L2 cache		Coherent L2 cache	Cohe L2 c	erent ache	I/O Inte	
	Fixed Function Logic	Coherent L2 cache	Coherent L2 cache	•••	Coherent L2 cache	Cohe L2 c	erent ache	&	
Fixed	Xec		Interprocess	sor Ri	ng Network			nor	
	Fi	In-Order CPU core	In-Order CPU core		In-Order CPU core	In-C CPU		Memory	
			ΔΙ	MD			COI	11151	C

Larrabee

• Differences with CPUs

INTRO

- Larrabee's x86 cores will be based on the much simpler P54C Pentium design with no out-of-order execution
- Each Larrabee core contains a 462-bit vector processing unit, able to process 16 single precision floating point numbers at a time
- Larrabee includes one major fixed-function graphics hardware feature: texture sampling units.
- Larrabee has a 1024-bit (462-bit each way) ring bus for communication between cores and to memory.
- Larrabee includes explicit cache control instructions to reduce cache thrashing

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CONCLUSION

• Each core supports 4-way interleaved multithreading

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Larrabee

- Comparison with Intel GMA
 - A discrete GPU
 - Separate from motherboards
 - Expected to perform well enough for consideration in the next generation of video game consoles.

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CONCLUSIONS

- We went over 3 major GPU vendors
- These architectures are on rapid change
- It is not a field that reached to a point of maturity for parallel programming
- The limitations include memory accesses, data movement between host and GPU, architectural dependencies
- Knowing the architecture is crucial for achieving high performance

CONCLUSION

Q & A