Parallel Accelerators

Přemysl Šůcha, Jan Dvořák and Libor Bukata

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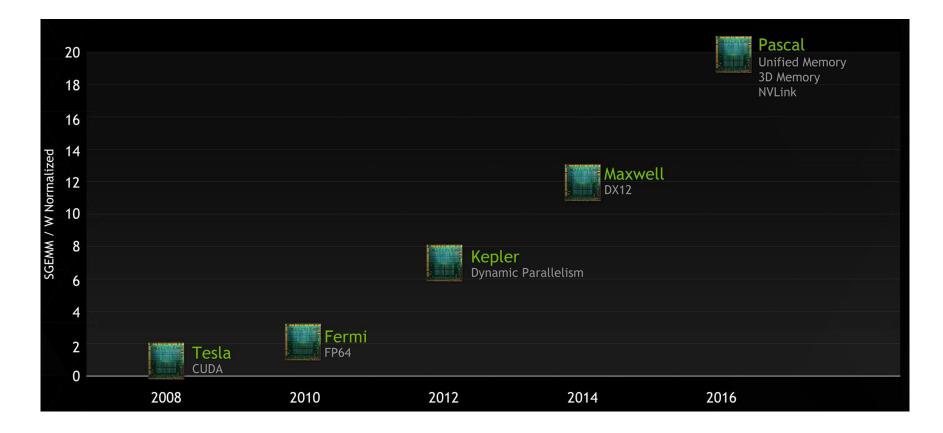
Topic Overview

- Graphical Processing Units (GPU) and CUDA
- Matrix transpose on CUDA
- Intel Xeon Phi
- Matrix equations on Xeon Phi

Graphical Processing Units



Graphical Processing Units



SGEMM - Single precision floating General Matrix Multiply

Graphical Processing Units

- GPU is especially well-suited to address problems that can be expressed as **data-parallel computations**.
- The same program is executed on many data elements in parallel - with high **arithmetic intensity** - the ratio of arithmetic operations to memory operations.
- Applications that process large data sets can use a data-parallel programming model to speed up the computations (3D rendering, image processing, video encoding, ...)
- Many algorithms outside the field of image rendering and processing are accelerated by data-parallel processing too (general signal processing, physics simulation, finance, computational biology).

Streaming Multiprocessor (SM)

SM															
_							Instructi	on Cache	į						
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	Warp Scheduler Dispatch Unit Dispatch Unit						Warp Scheduler Dispetch Unit Dispetch Unit								
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	Register File (32,768 x 32-bit)						Register File (32,768 x 32-bit)								
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Core	Core	DP Unit	Core	Core	DP Unit	LD/ST	SFU	Core	Core	DP Unil	Core	Core	DP Unit	LD/ST	SFU
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Core	Core	DP Unit	Core	Core	DP Unit	LD/ST	SFU	Core	Core	Unit	Core	Core	DP Unit	LD/ST	SFU
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Core	Core	Unit	Core	Core	Unit	LD/ST	SFU	Core	Core	Unit	Core	Core	Unit	LD/ST	SFU
	Texture / L1 Cache														
	Tex				Тех			Tex							
	64KB Shared Memory														

Streaming Multiprocessor (SM) - Pascal

- Pascal SM incorporates 64 single-precision CUDA Cores (Core), 32 double precision CUDA Cores (DP Unit), 16 SFUs (accelerate transcendental functions) and 16 LD/ST (Load / Store) units.
- SM is partitioned into two processing block.
- Each warp scheduler (one per processing block) is capable of dispatching two warp instructions per clock (1 warp = 32 threads).
- SMM has a dedicated **register file** and **shared memory**.

GPU Architecture - Pascal



GPU Architecture

Tesla Products	Tesla K40	Tesla M40	Tesla P100	
GPU	GK110 (Kepler)	GM200 (Maxwell)	GP100 (Pascal)	
SMs	15	24	56	
FP32 CUDA Cores / SM	192	128	64	
FP32 CUDA Cores / GPU	2880	3072	3584	
Base Clock	745 MHz	948 MHz	1328 MHz	
Peak FP32 GFLOPs	5040	6840	10600	
Memory Interface	384-bit GDDR5	384-bit GDDR5	4096-bit HBM2	
Memory Size	Up to 12 GB	Up to 24 GB	16 GB	
L2 Cache Size	1536 KB	3072 KB	4096 KB	
Register File Size / SM	256 KB	256 KB	256 KB	
TDP	235 Watts	250 Watts	300 Watts	
Transistors	7.1 billion	8 billion	15.3 billion	

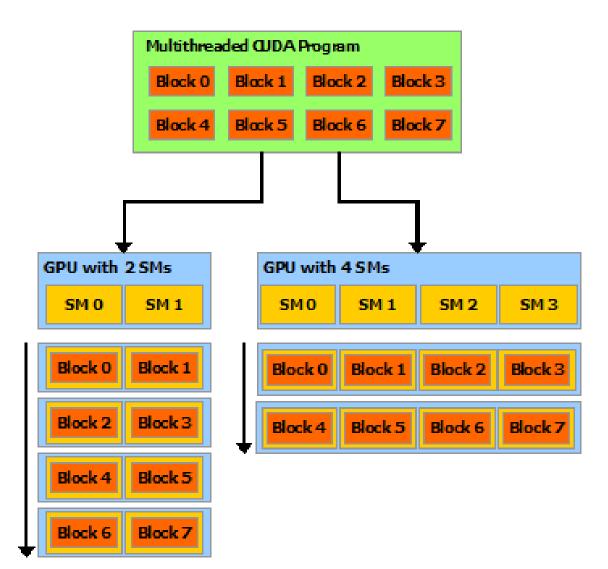
Single-Instruction, Multiple-Thread

- SIMT is an execution model where single instruction, multiple data (SIMD) is combined with multithreading.
- The SM creates, manages, schedules, and executes threads in groups of 32 parallel threads called **warps**.
- A warp start together at the same program address, but they have their own instruction address counter and register state and are therefore free to branch and execute independently.

CUDA

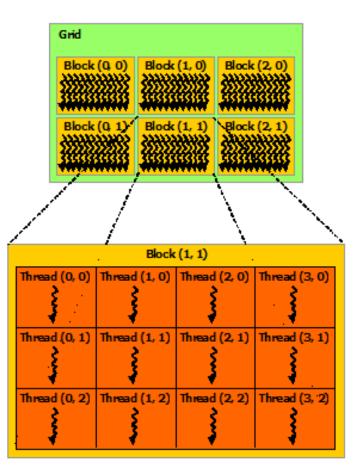
- The NVIDIA GPU architecture is built around a **scalable** array of multithreaded Streaming Multiprocessors (SMs).
- CUDA (Compute Unified Device Architecture) provides a way how a CUDA program can be executed on any number of SMs.
- A multithreaded program is partitioned into **blocks** of threads that execute independently from each other, so that a GPU with more multiprocessors will automatically execute the program in less time than a GPU with fewer multiprocessors.

CUDA



Grid/Block/Thread

- threads can be identified using a 1-D, 2-D, or 3-D thread index, forming a 1-D, 2-D, or 3-D block of threads, called a thread block.
- Blocks are organized into a 1-D, 2-D, or 3-D grid of thread blocks.



2-D grid with 2-D thread blocks

Kernel

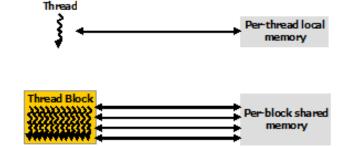
 CUDA C extends C by allowing the programmer to define C functions, called kernels.

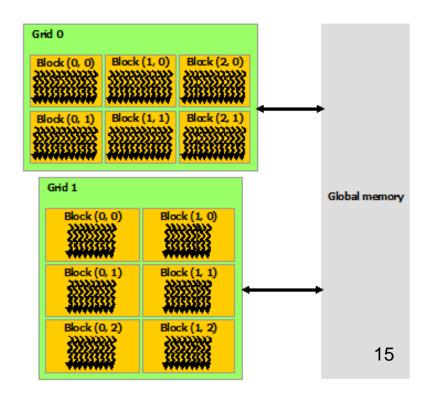
```
// Kernel definition
__global__ void VecAdd(float* A, float* B, float* C)
{
    int i = threadIdx.x;
    C[i] = A[i] + B[i];
}
int main()
{...
    // Kernel invocation with N threads inside 1 thread block
    VecAdd<<<1, N>>>(A, B, C);
}
```

 threadIdx is a 3-component vector, so that threads can be identified using a 1-D, 2-D, or 3-D thread index.

Memory Hierarchy

- Each thread has private set of registers and local memory.
- Each thread block has shared memory visible to all threads of the block.
- All threads have access to the same **global memory**.
- There are also two additional read-only memory spaces accessible by all threads (constant and texture memory).





GPU Programming - Demo

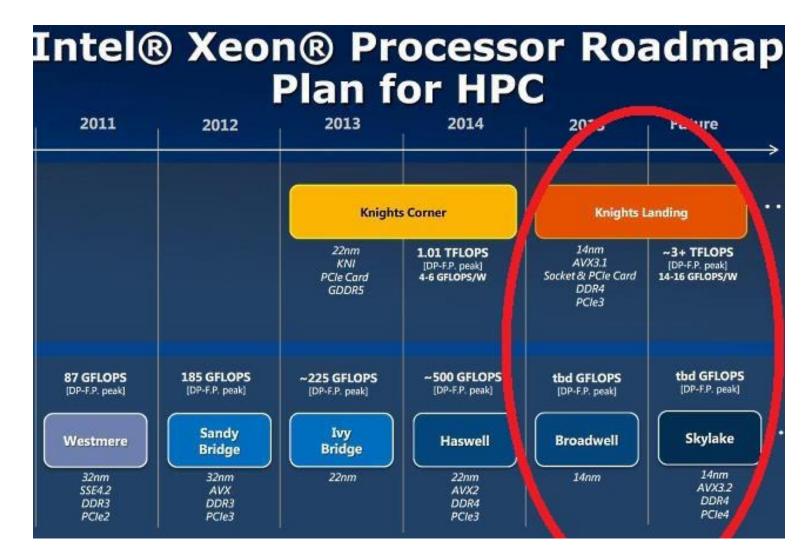
• Matrix transposition.

[1] Greg Ruetsch, *Optimizing Matrix Transpose in CUDA*, NVIDIA, 2009, http://www.cs.colostate.edu/~cs675/MatrixTranspose.pdf.

Intel Xeon Phi



Intel Xeon Phi

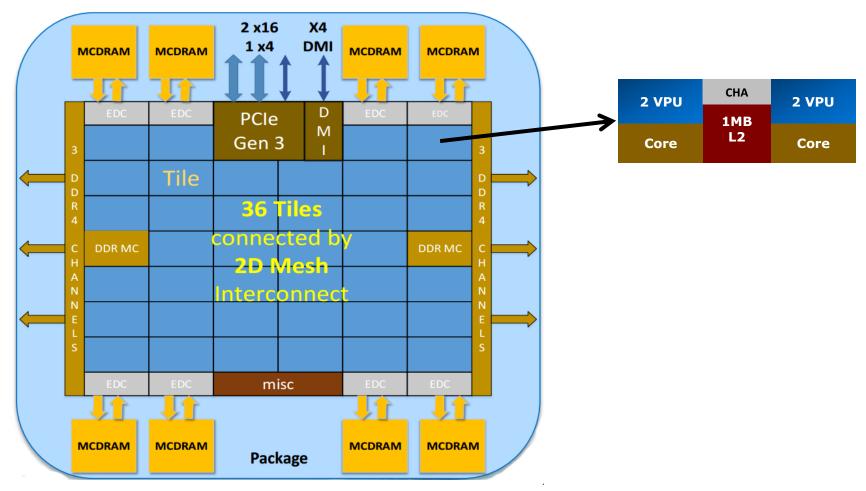


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Intel Xeon Phi

- Intel Xeon Phi coprocessors are designed to extend the reach of applications that have demonstrated the ability to fully utilize the scaling capabilities of Intel Xeon processor-based systems.
- Code compiled for Xeon processors can be run on an Xeon Phi (Knights Landing).
- For successful parallelization it requires a program with **lots of threads** and operations with **vectors**.

Knights Landing Architecture



Knights Landing Architecture

- The chip is constituted by 36 tiles interconnected by 2D mesh.
- The tile has two Cores (Atom Silvermont architecture), two vector processing units (VPU) and 1M L2 cache.
- A tile can execute concurrently 4 threads.
- The tiles are interconnected a cache-coherent 2D mesh; which provides a higher bandwidth and lower latency compare to the 1D ring interconnect on Knights corner.
- The mesh enforces **XY routing** rule.

Knights Landing Architecture

- Xeon Phi has 2 types of memory: (i) **MCDRAM** (Multichannel DRAM) and (ii) **DDR**.
- MCDRAM is a **high-bandwidth memory** integrated on the package. There are 8 of them 2 GB each.
- MCDRAM can be configured at boot time into one of three modes:
 - Cache mode MCDRAM is a cache for DDR,
 - Flat mode MCDRAM is a standard memory in the same address space as DDR,
 - Hybrid a combination
- DDR is a **high-capacity memory** which is external to the Knight Landing package.

Vectorization

- Each tile has two VPUs (Vector Processing Unit).
- It is the heard of computation. It processes all floating point computations using SSE, AVX, AVX2, ..., **AVX-512**.
- Thus each tile can **execute two 512-bit vector multiple-add instructions per cycle**, i.e. compute 32 double precision resp. 64 single precision floating point operation in each cycle.

Knights Corner vs. Knights Landing

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Product Name	Intel® Xeon Phi™ Coprocessor 7120X (16GB, 1.238 GHz, 61 <u>core)</u>	Intel® Xeon Phi™ Processor 7290F (16GB, 1.50 GHz, 72 <u>core)</u>
Code Name	Knights Corner	Knights Landing
Lithography	22 nm	14 nm
Recommended Customer Price	N/A	\$6401.00
# of Cores	61	72
Processor Base Frequency	1.24 GHz	1.50 GHz
Cache	30.5 MB L2	36 MB L2
ТДР	300 W	260 W
Max Memory Size (dependent on memory type)	16 GB	384 GB
Max Memory Bandwidth	352 GB/s	490 GB/s

Offloading

 Choose highly-parallel sections of code to run on the coprocessor. Serial code offloaded to the coprocessor will run much slower than on the CPU.

```
int x, y[100];
void f()
{
    x = 55;
    // x sent from CPU, y computed on coprocessor
    ...
#pragma offload target(mic:0) in(x) nocopy(y)
{ y[50] = 66; }
...
#pragma offload target(mic:0) nocopy(x,y)
{ // x and y retain previous values }
}
```

Xeon Phi Programming - Demo

• Simple matrix equation.

[2] James Jeffers and James Reinders, *Intel Xeon Phi Coprocessor High-Performance Programming*, Morgan Kaufmann, 2013.

References

[1] Greg Ruetsch, *Optimizing Matrix Transpose in CUDA*, NVIDIA, 2009.

[2] James Jeffers and James Reinders, *Intel Xeon Phi Coprocessor High-Performance Programming*, Morgan Kaufmann, 2013.

[3] NVIDIA, CUDA Toolkit Documentation v8.0, 2016. (http://docs.nvidia.com/cuda/index.html)

[4] Avinash Sodani, *Knights Landing (KNL): 2nd Generation* Intel® Xeon Phi[™] Processor, Intel, 2016. ()

[5] James Jeffers and James Reinders and Avinash Sodani, Intel Xeon Phi Processor High Performance Programming, 2nd Edition, Morgan Kaufmann, 2016.