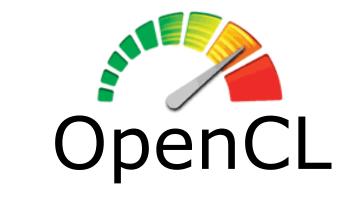
Talk in Parallel Computing



A brief introduction for NVIDIA CUDA programmers

Jan Beneke 16.01.2012

- Motivation
- Introducing OpenCL
- Design Goals
- Hardware/ Execution Model
- Software Stack
- Known Example
- Benchmark
- Summary

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Motivation Advantages of GPGPU

- Runs parallelized code on GPUs
- Speeds up existing applications
- Uses ordinary and "cheap" hardware accessible to almost every user
- Even supported by modern handheld devices
- Easy to code with dedicated language extensions and frameworks
- But: Difficult to create efficient code

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Motivation NVIDIA CUDA

Pro

- Easy to learn language extension for using GPGPU
- Intuitive framework
- Many extensions provided by NVIDIA (CUFFT, CUBLAS, ...)
- Integrated graphics language support (OpenGL, Direct 3D)
- Cross-platform (Microsoft Windows, Linux, Apple Mac OS X)
- Free of charge

Contra

- Only NVIDIA hardware suported
- Only GPUs supported
- Proprietary product/ Closed source

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Introducing OpenCL Open Computing Language



- Language extension for accessing heterogeneous computational devices
- Supportes parallel execution on single or multiple devices of many vendors
 - GPUs (NVIDIA, AMD/ ATI, ...)
 - CPUs (Intel, AMD, ...)
 - Accelerator cards/ Server blades (IBM, ...)
- Desktop, server farm and hand-held profiles
- Integrates OpenGL and Microsoft Direct 3D graphics APIs

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Introducing OpenCL An open standard

- Vendor neutral
- Specifications under review by Khronos OpenCL working group
 - Krohnos also reviews standards like OpenGL, WebGL, Open AL, ...
- Standard based on proposal by Apple and developed with industry leaders



Source: http://www.khronos.org

First released and natively integrated in Mac OS X since Snow Leopard (2009)

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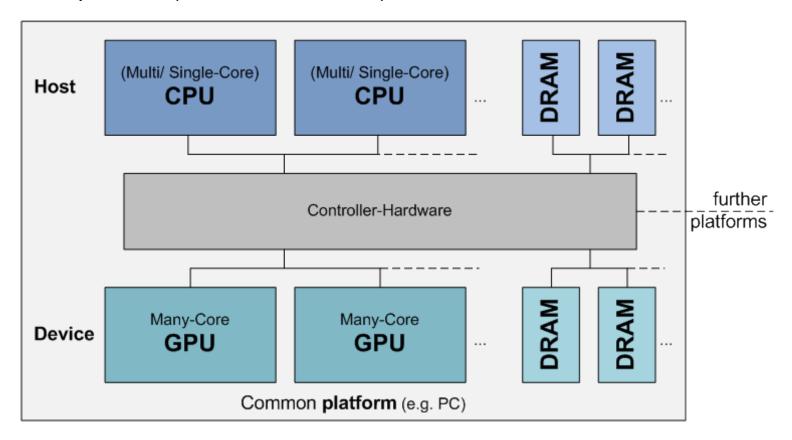
Design Goals Heterogeneous World

- A modern platform includes:
 - One or more CPUs
 - One or more GPUs
 - (DSPs)
 - (Accelerator cards)
 - → Use them as peers.
 - → One portable program uses <u>all</u> available resources.

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Design Goals Heterogeneous World

A common platform (PC or server blade):



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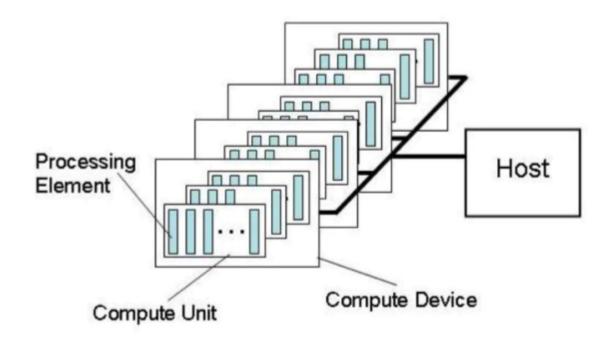
Design Goals Becoming a useable standard

- Efficent parallel model
 - Based on ISO C99
 - Abstract the underlaying hardware
 - Additional work-items and workgroups, vector types, synchronization, address space qualifiers
 - Built-in functions for image manipulation, work-item manipulation, math routines, ...
- Specify accuracy of floating-point computations (IEEE 754)
- Support of most future hardware

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Hardware Model Abstract Platform Model

- One host and multiple computing devices
 - Each device contains computing units
 - Each unit is devided into processing elements



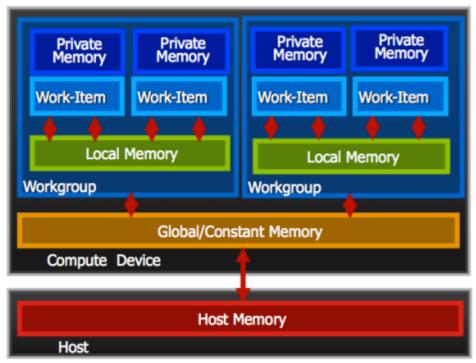
Source: http://www.khronos.org

Hardware Model Abstract Memory Model



Hardware is abstracted by a consitend shared memory model (like CUDA):

- Private Memory
 - Per work-item
- Local Memory
 - Shared in a workgroup
- Global/ Constant Memory
 - All workgroups on one device
- Host Memory
 - On host level (CPU)
- Independent of based hardware.
- Explicit memory model:
 - Mostly move data from host → global → local → ... and back



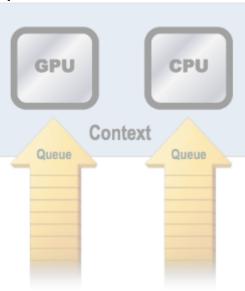
Source: http://www.khronos.org

Execution Model The OpenCL application



- The application runs on a host, that submits work to the compute devices
 - Work-item
 - Basic unit on OpenCL device
 - Kernel
 - The code for a work item an extended C function
 - Command Queue
 - Queues Kernels/ synchronizing event handling
 - Program
 - Collection of kernels and other functions analogous to a dynamic library
 - Context
 - The environment to execute the work-items in, inculdes devices and their memories and command queues

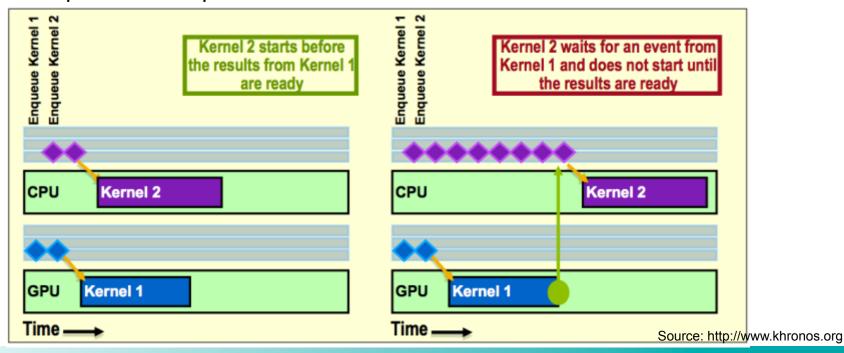
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Execution Model Synchronization

- Applications queue compute kernel execution instances
- Events can be used to synchronize kernel executions between queues
- Example: 2 queues on seperate devices



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Software Stack

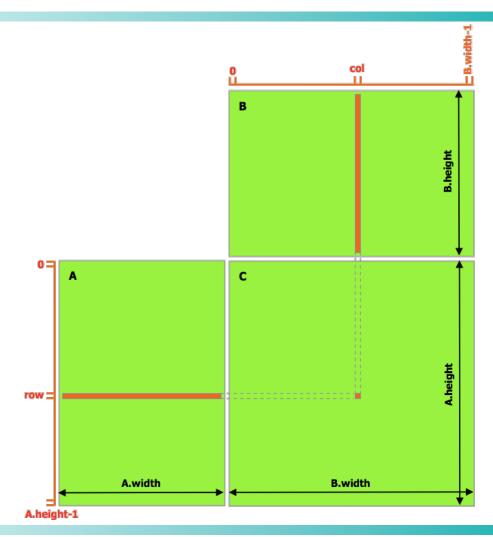


- Platform layer
 - Query and select computing devices
 - Initialize devices
 - Create work contexts and command queues
- Runtime
 - Ressource management
 - Execute kernels
- Compiler
 - Compile and build compute program executable (everything but the host code) at runtime

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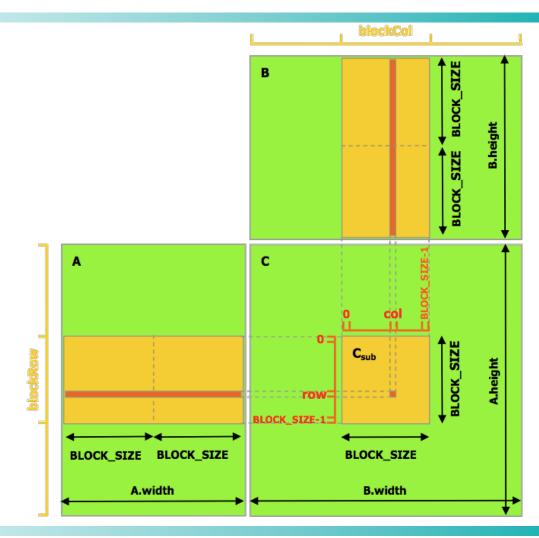
Known Example Matrix Multiplication



Source: http://www.nvidia.com

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Known Example Matrix Multiplication



Source: http://www.nvidia.com



Known Example Matrix Multiplication

- Code is devided into
 - Host API Code (matrixMul.cpp)
 - OpenCL Kernel Code (matrixMul.cl)
- Kernel code will be familiar to CUDA programmers

Source: http://www.khronos.org

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Known Example Compute Kernel Code

```
// Workgroup (thread block) dimensions
#define BLOCK_SIZE 16

// Fast element access
#define AS( i, j) As[ j + i * BLOCK_SIZE]
#define BS( i, j) Bs[ j + i * BLOCK_SIZE]
```

Known Example Compute Kernel Code

```
kernel void matrixMul(
     __global float* C, __global float* A, __global float* B,
     __local float* As, __local float* Bs, int widthA, int widthB)
 int bx = get_group_id( 0);  // Workgroup (block) index
                        // Workgroup (block) index
 int by = get group id(1);
                        // Work-item (thread) index
 int tx = get local id( 0);
                        // Work-item (thread) index
 int ty = get_local_id( 1);
 int aBegin = widthA * BLOCK SIZE * by; // Sub-matrix A start
 int aEnd = aBegin + widthA - 1; // Sub-matrix A end
                      // Sub-matrix A stepsize
 int aStep = BLOCK SIZE;
 float Csub = 0.0f:
                                 // Result of block
 // . . .
```

matrixMul.cl

Known Example Compute Kernel Code

```
// Loop over A and B to compute the block sub-matrix
for( int a = aBegin, b = bBegin; a <= aEnd; a += aStep, b += bStep)</pre>
    // Each thread loads one element of the mat. from device to shared memory
    AS(ty, tx) = A[a + widthA * ty + tx];
    BS( tv, tx) = B[ b + widthB * tv + tx];
    // Synchronize to make sure the matrices are loaded
    barrier( CLK LOCAL MEM FENCE);
    // Multiply the two matrices
    #pragma unroll
    for( int k = 0; k < BLOCK SIZE; ++k)
        Csub += AS( ty, k) * BS( k, tx);
    // Synchronize to make sure that the preceding
    barrier( CLK LOCAL MEM FENCE);
// Write the block sub-matrix to device memory
C[ get_global_id( 1) * get_global_size( 0) + get_global_id( 0)] = Csub;
                                                                       matrixMul.cl
```

```
#include <CL/opencl.h>
// Matrix dimensions
int WA = 32 * BLOCK SIZE;  // Input matrix A width
int HA = 128 * BLOCK SIZE;  // Input matrix A height
int WB = 32 * BLOCK_SIZE;  // Input matrix B width
int HB = WA;
                                // Input matrix B height
int WC = WB;
                                // Resulting matrix C width
int HC = HA;
                                // Resulting matrix C height
int main( int argc, char** argv)
                                      // Device context handle
   cl context GPUContext;
    cl kernel matMulKernel;
                                      // Kernel handle
    cl command queue commandQueue;
                                    // Command queue handle
    cl program program;
                                        // Program handle
   cl device id deviceID = NULL;
                                        // IDs of all OpenCL devices
   // . . .
```

```
// Allocate memory for matrizes on host level
float* hA = (float*) malloc( WA * HA * sizeof(float));
float* hB = (float*) malloc( WB * HB * sizeof(float));
float* hC = (float*) malloc( WC * HC * sizeof(float));
// Fill arrays with random numbers
for(int i = 0; i < WA * HA; i++)
    hA[i] = rand() / (float) RAND MAX;
for(int i = 0; i < WB * HB; i++)
    hB[i] = rand() / (float) RAND MAX;
// Get OpenCL devices
clGetDeviceIDs( NULL, CL_DEVICE_TYPE_GPU, 1, &deviceID, NULL);
// Create OpenCL context
GPUContext = clCreateContext( 0, 1, &deviceID, NULL, NULL);
// Create command queue
commandQueue = clCreateCommandQueue( GPUContext, deviceID,
                     CL QUEUE PROFILING ENABLE, NULL);
// . . .
```

```
// Load OpenCL kernel source code from source file
size t programLength;
char* source = loadProgSource( SOURCE PATH, "", &programLength);
// Create OpenCL program
program = clCreateProgramWithSource( GPUContext, 1, (const char**) &source,
                      &programLength, NULL);
// Build OpenCL program
clBuildProgram( program, 0, NULL, "", NULL, NULL);
// Create OpenCL kernel
matMulKernel = clCreateKernel( program, "matrixMul", NULL);
// . . .
```

```
// Create OpenCL buffer pointing to the host memory
cl mem hABuffer = clCreateBuffer( GPUContext, CL MEM READ ONLY |
                      CL MEM USE HOST PTR, WA*HA*sizeof(float), hA, NULL);
// Create buffers (memory on device) for matrizes
cl_mem dA = clCreateBuffer( GPUContext, CL_MEM_READ_ONLY,
                     WA*HA*sizeof(float), NULL, NULL);
// Copy from host to device, now
clEnqueueCopyBuffer( commandQueue, hABuffer, dA, 0, 0, WA*HA*sizeof(float),
                     O. NULL. NULL);
// Create buffer on device, it will be initialized from the host at first use
cl mem dB = clCreateBuffer( GPUContext, CL MEM READ ONLY)
                     CL MEM COPY HOST PTR, WB*HB*sizeof(float), hB, NULL);
// Create buffer for resulting matrix
cl mem dC = clCreateBuffer( GPUContext, CL MEM WRITE ONLY,
                     WC*HA*sizeof(float), NULL,NULL);
// . . .
```

```
// Set the argument values for the kernel
clSetKernelArg( matMulKernel, 0, sizeof(cl_mem), (void *) &dC);
clSetKernelArg( matMulKernel, 1, sizeof(cl_mem), (void *) &dA);
clSetKernelArg( matMulKernel, 2, sizeof(cl_mem), (void *) &dB);
clSetKernelArg( matMulKernel, 3, BLOCK SIZE*BLOCK SIZE*sizeof(float), 0);
clSetKernelArg( matMulKernel, 4, BLOCK SIZE*BLOCK SIZE*sizeof(float), 0);
clSetKernelArg( matMulKernel, 5, sizeof(cl_int), (void *) &WA);
clSetKernelArg( matMulKernel, 6, sizeof(cl int), (void *) &WB);
// Execute Multiplication in parallel
size_t localWorkSize[] = { BLOCK_SIZE, BLOCK_SIZE};
size t globalWorkSize[] = { WC, HA};
// Launch kernel
// Multiplication - non-blocking execution: launch and push to device
    clEngueueNDRangeKernel( commandQueue, matMulKernel, 2, 0, globalWorkSize,
             localWorkSize, 0, NULL, NULL);
// Sync to host
clFinish( commandQueue);
// . . .
                                                                     matrixMul.cpp
```

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```
// Non-blocking copy of result from device to host
clEnqueueReadBuffer( commandQueue, dC, CL_FALSE, 0, WC*HA*sizeof(float),
                      hC, 0, NULL, NULL);
// Release mem objects (device memory)
clReleaseMemObject( hABuffer);
clReleaseMemObject( dA);
clReleaseMemObject( dC);
clReleaseMemObject( dB);
// Clean up OpenCL resources
clReleaseKernel( matMulKernel);
clReleaseCommandOueue( commandOueue);
clReleaseProgram( program);
clReleaseContext( GPUContext);
// Clean up host memory
free( hA); free( hB); free( hC);
// Exit application
return( 0);
```

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Benchmark Test System

Intel Core i3-2100

2 cores (2 Threads each) @ 3.1 GHz - 3MB cache - 8GB DDR3 (1333MHz) max. 52 GFLOPS / s (LINPACK)

MSI N560GTS-Ti Twin Frozr II/OC

NVIDIA GeForce GTX 560 Ti – 384 cores @ 880Mhz – 1024MB GDDR5 (4200MHz) max. 1.253 GFLOPS / s (approximated)

Compiled with Microsoft Visual Studio 2008 Professional x86 under Microsoft Windows 7 Professional SP1 x64 against

- OpenCL 1.2 (NVIDIA GPU Computing SDK 4.0 x86)
- NVIDIA CUDA 4.0 x86
- NVIDIA CUBLAS 2.0 (NVIDIA GPU Computing SDK 4.0 x86)
- OpenMP 2.0 (Microsoft Visual Studio 2008 x86)

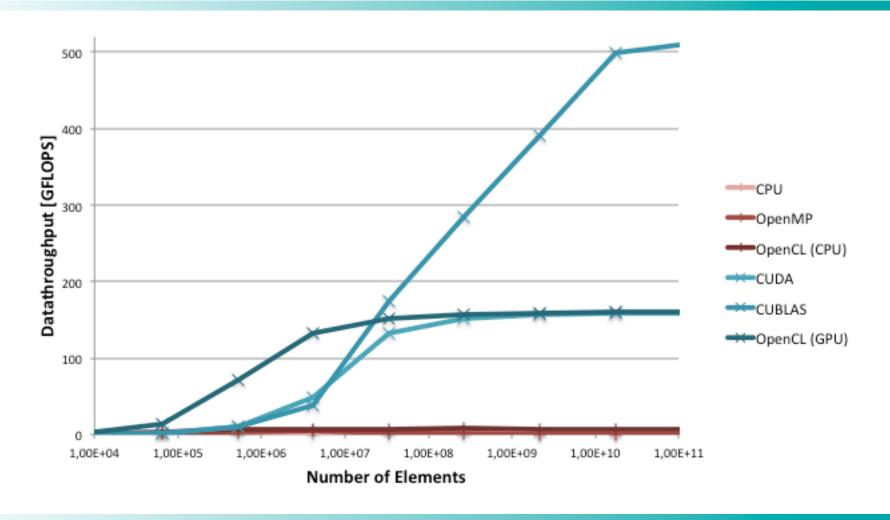


$$A^{2048 \times 8192} \cdot B^{4096 \times 2048} = C^{4096 \times 8196}$$

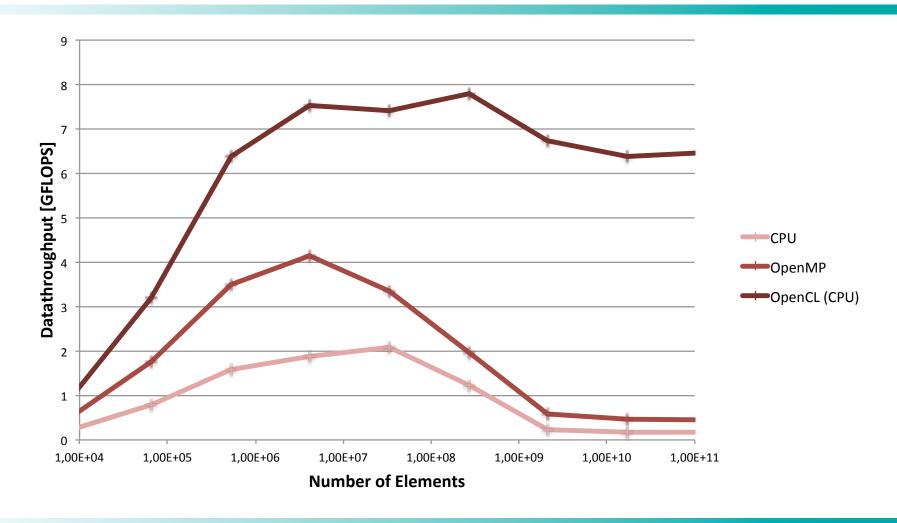
 \Rightarrow 137.4·10⁹ elements to compute

		Execution Time [s]	Throughput [GFLOPS / s]
CPU	Single threaded	818.13	0.16
	OpenMP	460.05	0.30
	OpenCL 1.2	21.76	6.32
GPU	NVIDIA CUDA 4.0	0.87	158.23
	OpenCL 1.2	0.86	159.72
	NVIDIA CUBLAS 2.0	0.24	580.90

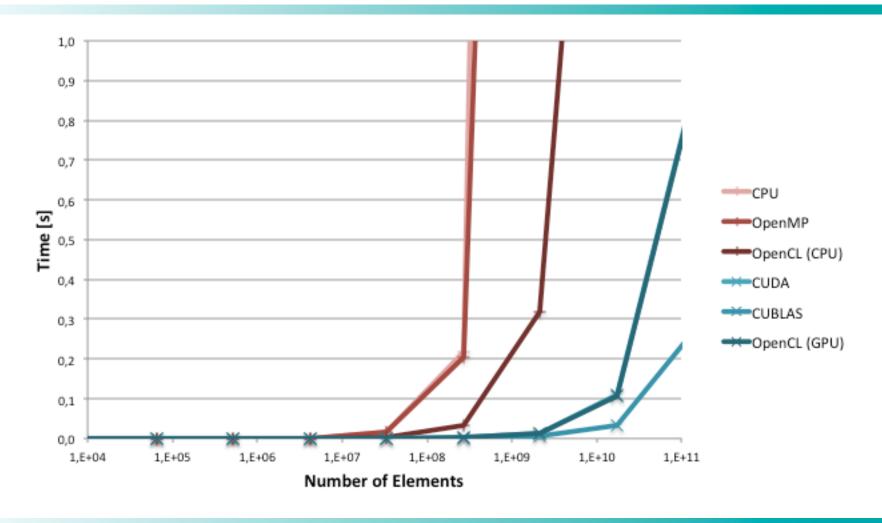
Average after 100 iterations







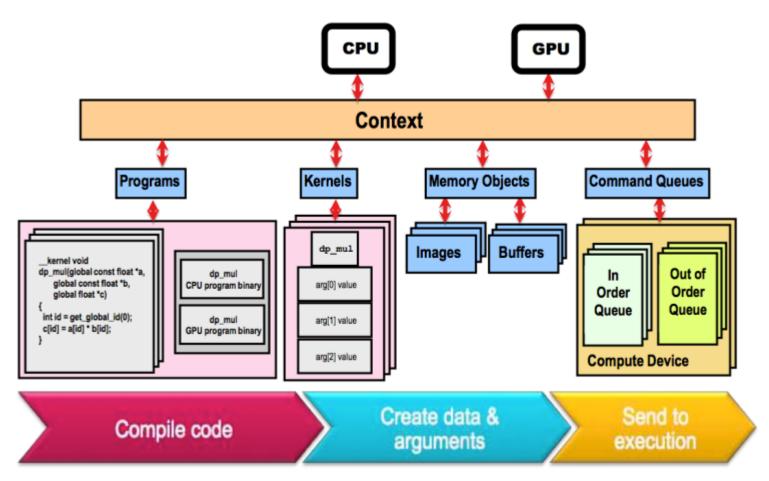




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Summary The OpenCL workflow



Source: http://www.khronos.org





	OpenCL 1.2	NVIDIA CUDA 4.0
Free of charge	✓	✓
Vendor independent/ Multiple SDK-vendors	✓	
Open standard	✓	
Use different types of heterogeneous hardware	✓	
Cross-platform	✓	✓
Interoperability with graphics libraries	✓	✓
Included libraries for common tasts		✓
Peer-to-peer-comunication between devices		✓
Global virtual addressing space		✓

Summary Conclusion

OpenCL:

- More complicated to code and redistribute
- Using different types of devices
- Vendor independend
- Outstanding performance on many CPUs

NVIDIA CUDA:

- More enhanced
- Already widely spread
- Deliveres well-engineered libraries for many tasks

Thank you for your attention.

Source code at http://janbeneke.de/ParComp

References

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