University of Manchester

Valeriy Titarenko

School of Materials, University of Manchester, UK

Valeriy.Titarenko@manchester.ac.uk

Filtered backprojection, CPU and GPU

Input data: 4k $\times$ 6k; output data: 4k $\times$ 4k

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GPU: Tesla C870

	2k $\times$ 2k	4k $\times$ 6k
1k $\times$ 1k	0.70	2.97
2k $\times$ 2k	1.48	5.09
3k $\times$ 3k	2.75	8.57
4k $\times$ 4k	4.50	13.69
5k $\times$ 5k	6.87	19.94

GPU (graphics processing unit)

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CUDA (Compute Unified Device Architecture)

Tesla project

1 Multiprocessor = 8 Processors

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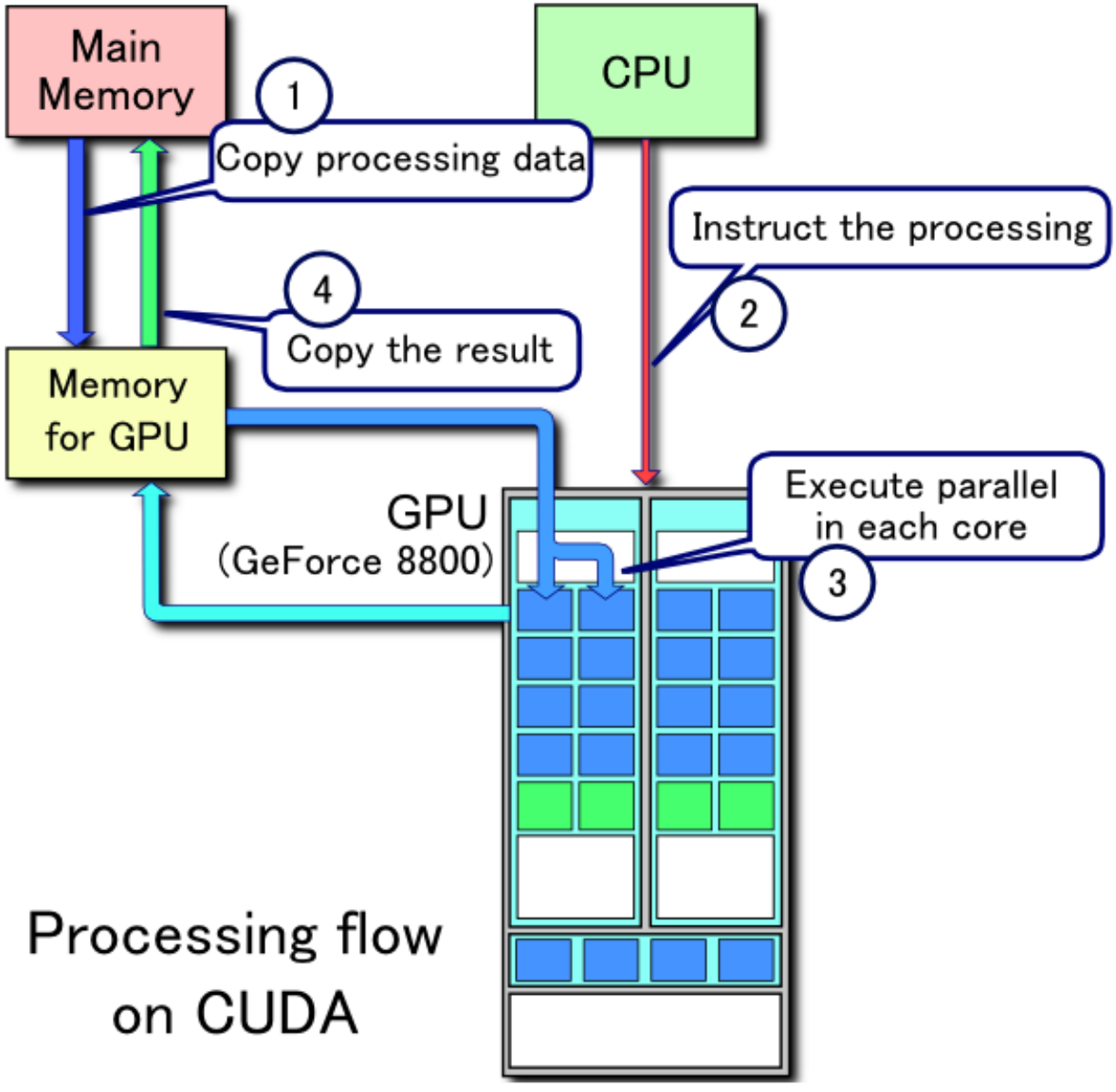
Personal supercomputer

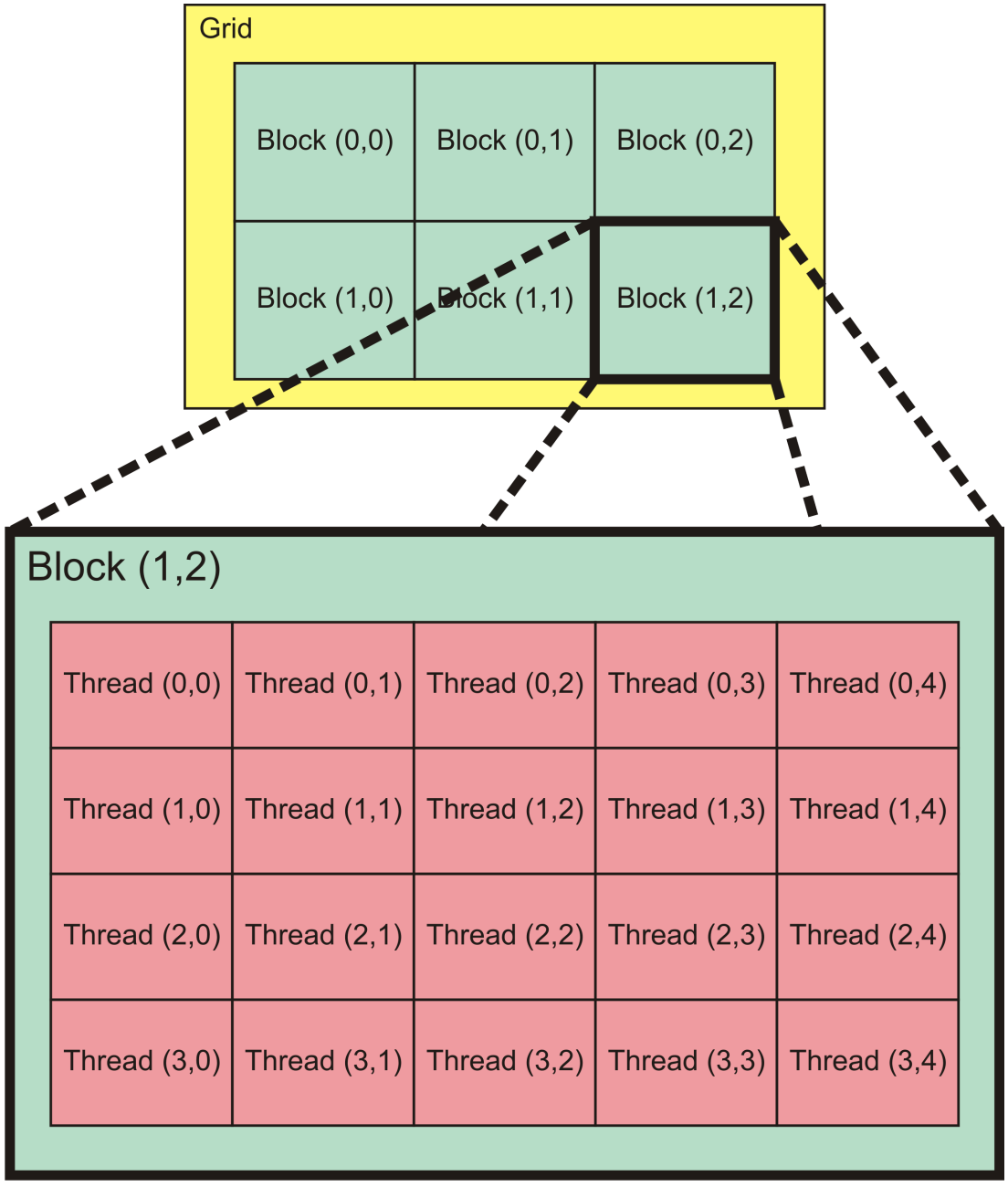
- 4 GeForce GTX 295 (240 multiprocessors), 4 $\times$ \$560
- motherboard Asus P6T7 WS SuperComputer, \$400
- memory 12 GB, \$600
- Intel Xeon processor, \$250
- case (for dual power supplies), \$280
- power supplies, 2 $\times$ \$230
- . . . Total $\approx$ \$4600


Main Memory
CPU
1
Copy processing data
Instruct the processing
2
4
Copy the result
Memory for GPU
GPU
(GeForce 8800)
Execute parallel in each core
3
Processing flow on CUDA

Matrix addition, 1

```
// Kernel definition

__global__ void MatAdd(float A[N][N], float B[N][N], float C[N][N])
{
    int i = threadIdx.x;
    int j = threadIdx.y;
    C[i][j] = A[i][j] + B[i][j];
}

int main()
{
    // Kernel invocation
    dim3 dimBlock(N, N);
    MatAdd<<<1, dimBlock>>>(A, B, C);
}
```

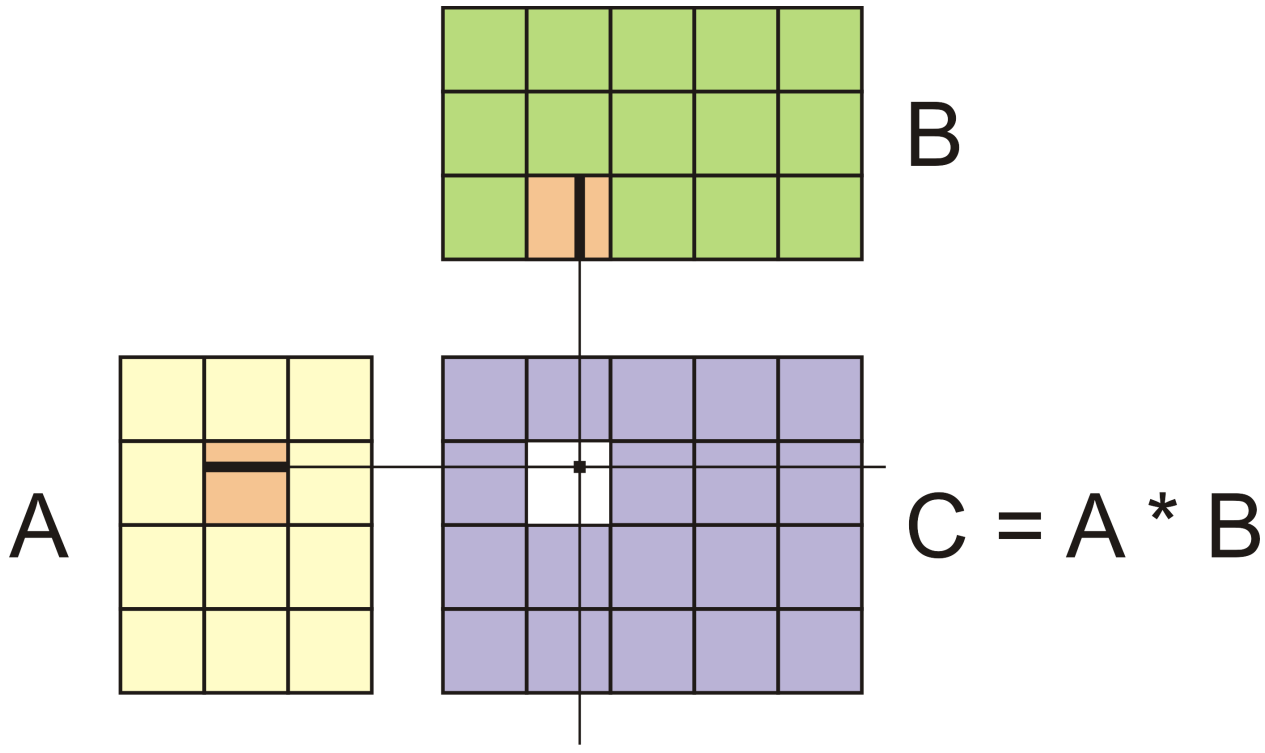
Threads, blocks, grid



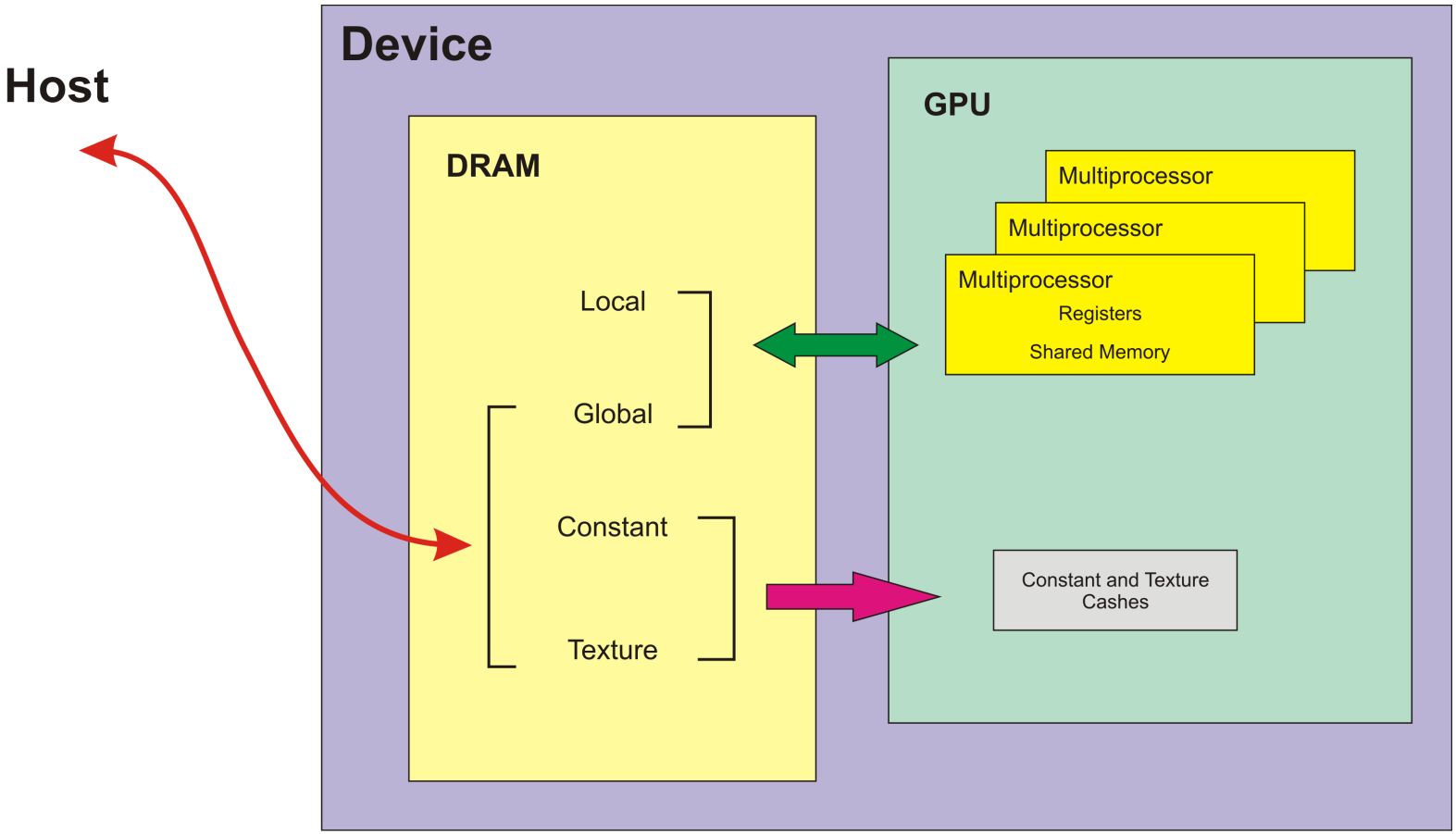
Matrix addition, 2

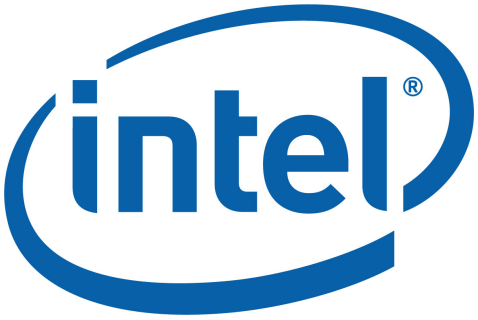
```
// Kernel definition

__global__ void MatAdd(float A[N][N], float B[N][N], float C[N][N])
{
    int i = blockIdx.x * blockDim.x + threadIdx.x;
    int j = blockIdx.y * blockDim.y + threadIdx.y;
    if (i < N && j < N)
        C[i][j] = A[i][j] + B[i][j];
}

int main()
{
    // Kernel invocation
    dim3 dimBlock(16, 16);
    dim3 dimGrid((N + dimBlock.x – 1) / dimBlock.x, (N + dimBlock.y – 1) / dimBlock.y);
    MatAdd<<<dimGrid, dimBlock>>>(A, B, C);
}
```

Matrix multiplication



GPU memory

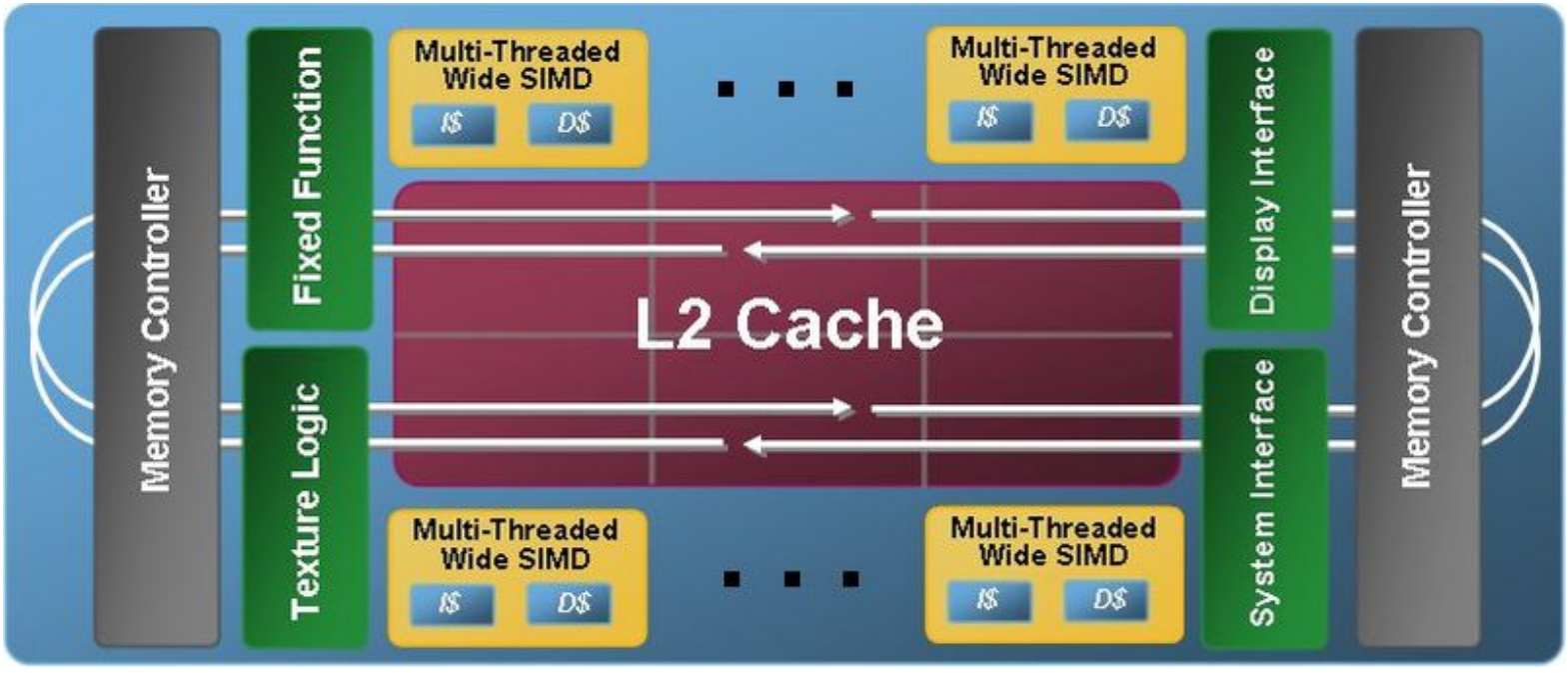

Host
Device
GPU
DRAM
Multiprocessor
Multiprocessor
Multiprocessor
Registers
Shared Memory
Local
Global
Constant
Texture
Constant and Texture
Cashes



- Visual Fortran and C++ Compilers
  OS: Windows, Linux, and Mac OS X
- Threading Building Blocks (TBB)
  C++ template library that abstracts threads to tasks to create reliable, portable, and scalable parallel applications
- Math Kernel Library (MKL)
  a library of highly optimized, extensively threaded math routines for science, engineering, and financial applications that require maximum performance
- Integrated Performance Primitives (IPP)

nVidia vs Intel (future)

- Intel's Larrabee



- nVidia's Fermi