

GPU ACCELERATORS AT JSC JÜLICH CHALLENGES HACKATHON

18 March 2021 | Andreas Herten | Forschungszentrum Jülich



Outline

GPUs at JSC
JUWELS
JURECA DC
GPU Architecture
Empirical Motivation
Comparisons
GPU Architecture

Programming GPUs
Libraries
Directives
CUDA C/C++
Performance Analysis
Advanced Topics
Using GPUs on JSC Systems
Compiling
Resource Allocation

Slide 1137





JUWELS Cluster - Jülich's Scalable System

- 2500 nodes with Intel Xeon CPUs (2 × 24 cores)
- 46 + 10 nodes with 4 NVIDIA Tesla V100 cards (32 GB memory)
- 10.4 (CPU) + 1.6 (GPU) PFLOP/s peak performance (Top500: #44)





JUWELS Booster – Scaling Higher!

- \blacksquare 936 nodes with AMD EPYC Rome CPUs (2 \times 24 cores)
- Each with 4 NVIDIA A100 Ampere GPUs (each: FP64TC: 19.5 TFLOP/s)
- ullet InfiniBand DragonFly+ HDR-200 network; 4 imes 200 Gbit/s per node



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Top500 List Nov 2020:

- #1 Europe
- #7 World
- #3* Green500

JUWELS Booster – Scaling Higher!

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JURECA DC – Multi-Purpose

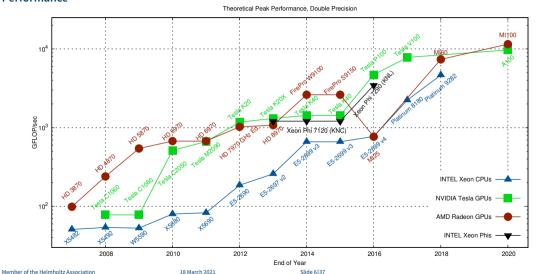
- 768 nodes with AMD EPYC Rome CPUs (2 \times 64 cores)
- 192 nodes with 4 NVIDIA A100 Ampere GPUs
- InfiniBand DragonFly+ HDR-100 network
- Also: JURECA Booster: 1640 nodes with Intel Xeon Phi Knights Landing



GPU Architecture

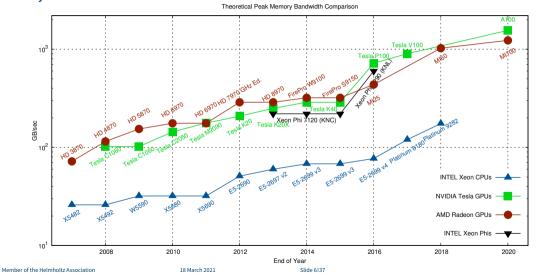
Status Quo Across Architectures

Performance



Status Quo Across Architectures

Memory Bandwidth



A matter of specialties





A matter of specialties



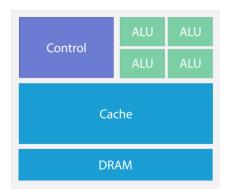
Transporting one



Transporting many

raphics: lee [3] and Shearings Holida

Chip

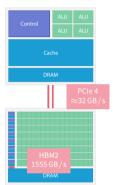






GPU optimized to hide latency

- Memory
 - GPU has small (40 GB), but high-speed memory 1555 GB/s
 - Stage data to GPU memory: via PCIe 4 bus (32 GB/s)

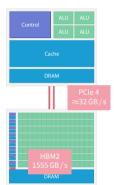


Device



GPU optimized to hide latency

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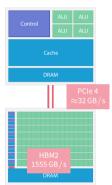


Device



GPU optimized to hide latency

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 - Stage automatically (*Unified Memory*), or manually



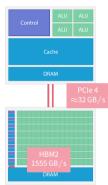
Device



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V100

32 GB RAM, 900 GB/s

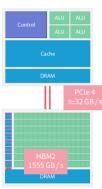


A100

Slide 0137

40 GB RAM, 1555 GB/s





Device



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 - GPU has small (40 GB), but high-speed memory 1555 GB/s
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- Two engines: Overlap compute and copy



SIMT

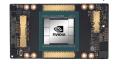
V100

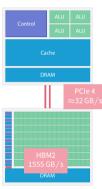
32 GB RAM, 900 GB/s



A100

40 GB RAM, 1555 GB/s





Device



$\mathbf{SIMT} = \mathbf{SIMD} \oplus \mathbf{SMT}$

- CPU:
 - Single Instruction, Multiple Data (SIMD)

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Scalar



 $\mathbf{SIMT} = \mathbf{SIMD} \oplus \mathbf{SMT}$

- CPU:
 - Single Instruction, Multiple Data (SIMD)

Vector



$\mathbf{SIMT} = \mathbf{SIMD} \oplus \mathbf{SMT}$

- CPU:
 - Single Instruction, Multiple Data (SIMD)
 - Simultaneous Multithreading (SMT)

Vector







$\mathbf{SIMT} = \mathbf{SIMD} \oplus \mathbf{SMT}$

- CPU:
 - Single Instruction, Multiple Data (SIMD)
 - Simultaneous Multithreading (SMT)

Vector



SMT



$SIMT = SIMD \oplus SMT$

- CPU:
 - Single Instruction, Multiple Data (SIMD)
 - Simultaneous Multithreading (SMT)
- GPU: Single Instruction, Multiple Threads (SIMT)

Vector



SMT



Slide 10137

$SIMT = SIMD \oplus SMT$

- CPU:
 - Single Instruction, Multiple Data (SIMD)
 - Simultaneous Multithreading (SMT)
- GPU: Single Instruction, Multiple Threads (SIMT)

Vector



SMT





$SIMT = SIMD \oplus SMT$

- CPU:
 - Single Instruction, Multiple Data (SIMD)
 - Simultaneous Multithreading (SMT)
- GPU: Single Instruction, Multiple Threads (SIMT)
 - CPU core ≈ GPU multiprocessor (SM)
 - Working unit: set of threads (32, a warp)
 - Fast switching of threads (large register file)
 - Branching if —

Vector



SMT



SIMT



$SIMT = SIMD \oplus SMT$



Vector



SMT







$SIMT = SIMD \oplus SMT$



Vector



SMT







Multiprocessor

SIMT

 $SIMT = SIMD \oplus SMT$

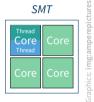


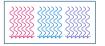


Vector



SMT







Let's summarize this!



Optimized for low latency

- + Large main memory
- + Fast clock rate
- + Large caches
- + Branch prediction
- + Powerful ALU
- Relatively low memory bandwidth
- Cache misses costly
- Low performance per watt



Optimized for high throughput

- + High bandwidth main memory
- + Latency tolerant (parallelism)
- + More compute resources
- + High performance per watt
- Limited memory capacity
- Low per-thread performance
- Extension card



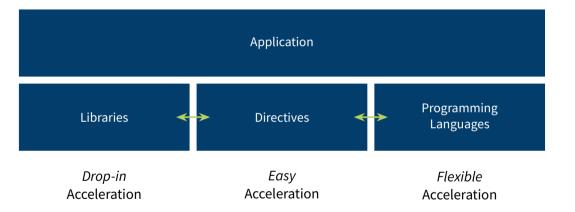
Programming GPUs

Preface: CPU

A simple CPU program!

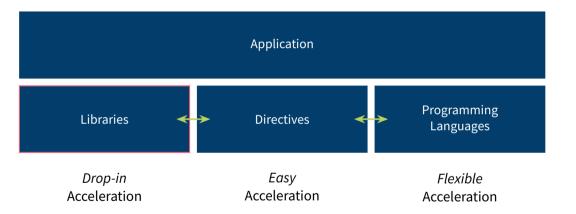
```
SAXPY: \vec{y} = a\vec{x} + \vec{y}, with single precision
Part of LAPACK BLAS Level 1
void saxpy(int n, float a, float * x, float * y) {
  for (int i = 0; i < n; i++)
    y[i] = a * x[i] + v[i];
float a = 42;
int n = 10;
float x[n], y[n];
// fill x, v
saxpy(n, a, x, y);
```

Summary of Acceleration Possibilities





Summary of Acceleration Possibilities





Libraries

Programming GPUs is easy: Just don't!



Libraries

Programming GPUs is easy: Just don't!

Use applications & libraries



Libraries

Programming GPUs is easy: Just don't!

Use applications & libraries



Wizard: Breazell [5]

Use applications & libraries























Numba

th⊝ano

Wizard: Breazell [5]

Use applications & libraries

















Thrus



Numba







cuBLAS

Parallel algebra



- GPU-parallel BLAS (all 152 routines)
- Single, double, complex data types
- Constant competition with Intel's MKL
- Multi-GPU support
- → https://developer.nvidia.com/cublas http://docs.nvidia.com/cuda/cublas



cuBLAS

Code example

```
int a = 42: int n = 10:
float x[n]. v[n]:
// fill x, v
cublasHandle t handle:
cublasCreate(&handle):
float * d x. * d v:
cudaMallocManaged(\delta d x. n * sizeof(x[0])):
cudaMallocManaged(&d y, n * sizeof(y[0]));
cublasSetVector(n, sizeof(x[0]), x, 1, d x, 1):
cublasSetVector(n, sizeof(y[0]), y, 1, d y, 1);
cublasSaxpy(n, a, d x, 1, d y, 1);
cublasGetVector(n. sizeof(v[0]), d v. 1. v. 1):
cudaFree(d x); cudaFree(d y);
cublasDestrov(handle):
```

cuBLAS

Code example

```
int a = 42: int n = 10:
float x[n]. v[n]:
// fill x, v
cublasHandle t handle:
cublasCreate(&handle):
float * d x. * d v:
                                                                               Allocate GPU memory
cudaMallocManaged(&d x. n * sizeof(x[0])):●
cudaMallocManaged(&d y, n * sizeof(y[0]));
                                                                                   Copy data to GPU
cublasSetVector(n. sizeof(x[0]), x, 1, d x, 1):
cublasSetVector(n, sizeof(y[0]), y, 1, d y, 1);
                                                                                    Call BLAS routine
cublasSaxpy(n, a, d x, 1, d y, 1); \bullet
                                                                                  Copy result to host
cublasGetVector(n. sizeof(v[0]). d v. 1. v. 1):
                                                                                            Finalize
cudaFree(d x); cudaFree(d y);
```



cublasDestrov(handle):

Directives

Programming GPUs

GPU Programming with Directives

Keepin' you portable

Annotate serial source code by directives

```
#pragma acc loop
for (int i = 0; i < 1; i++) {};</pre>
```



GPU Programming with Directives

Keepin' you portable

Annotate serial source code by directives

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#pragma acc loop
for (int i = 0; i < 1; i++) {};</pre>
```

- OpenACC: Especially for GPUs; OpenMP: Has GPU support
- Compiler interprets directives, creates according instructions



GPU Programming with Directives

Keepin' you portable

Annotate serial source code by directives

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#pragma acc loop
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- OpenACC: Especially for GPUs; OpenMP: Has GPU support
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Pro

- Portability
 - Other compiler? No problem! To it, it's a serial program
 - Different target architectures from same code
- Easy to program

Con

- Compiler support only raising
- Not all the raw power available
- Harder to debug
- Easy to program wrong



OpenACC

Code example

```
void saxpy_acc(int n, float a, float * x, float * y) {
    #pragma acc kernels
    for (int i = 0; i < n; i++)
        y[i] = a * x[i] + y[i];
}

float a = 42;
int n = 10;
float x[n], y[n];
// fill x, y

saxpy_acc(n, a, x, y);</pre>
```

OpenACC

Code example

```
void saxpv acc(int n, float a, float * x, float * v) {
  #pragma omp target map(to:x[0:n]) map(tofrom:y[0:n]) loop
  for (int i = 0; i < n; i++)
   v[i] = a * x[i] + v[i]:
float a = 42;
int n = 10:
float x[n], y[n];
// fill x. v
saxpv acc(n, a, x, v);
```

CUDA C/C++

Programming GPUs

Finally...



Slide 22137

Finally...

OpenCL Open Computing Language by Khronos Group (Apple, IBM, NVIDIA, ...) 2009

- Platform: Programming language (OpenCL C/C++), API, and compiler
- Targets CPUs, GPUs, FPGAs, and other many-core machines
- Fully open source



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CUDA NVIDIA's GPU platform 2007

- Platform: Drivers, programming language (CUDA C/C++), API, compiler, tools, ...
- Only NVIDIA GPUs
- Compilation with nvcc (free, but not open)
 clang has CUDA support, but CUDA needed for last step
- Also: CUDA Fortran; and more in NVIDIA HPC SDK



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- Choose what flavor you like, what colleagues/collaboration is using
- Hardest: Come up with parallelized algorithm



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HIP AMD's new unified programming model for AMD (via ROCm) and NVIDIA GPUs 2016+

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In software: Threads, Blocks



In software: Threads, Blocks

- Methods to exploit parallelism:
 - Thread

3

In software: Threads, Blocks

- Methods to exploit parallelism:
 - Threads





In software: Threads, Blocks

Methods to exploit parallelism:

 $\blacksquare \quad \underline{\mathsf{Threads}} \to \underline{\mathsf{Block}}$





In software: Threads, Blocks

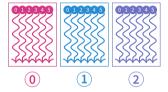
- $\bullet \quad \underbrace{\mathsf{Threads}}_{} \to \underbrace{\mathsf{Block}}_{}$
- Block





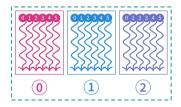
In software: Threads, Blocks

- $\blacksquare \quad \text{Threads} \rightarrow \quad \text{Block}$
- Blocks



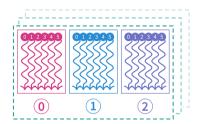
In software: Threads, Blocks

- Threads → Block
- lacks ightarrow Grid



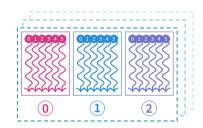
In software: Threads, Blocks

- Methods to exploit parallelism:
 - $\blacksquare \quad \text{Threads} \rightarrow \quad \text{Block}$
 - lacks ightarrow Grid
 - Threads & blocks in 3D



In software: Threads, Blocks

- Methods to exploit parallelism:
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- Parallel function: kernel
 - __global__ kernel(int a, float * b) { }
 - Access own ID by global variables threadIdx.x, blockIdx.y,...
- Execution entity: threads
 - Lightweight → fast switchting!
 - $lue{}$ 1000s threads execute simultaneously ightarrow order non-deterministic!

CUDA SAXPY

With runtime-managed data transfers

```
global void saxpy cuda(int n, float a, float * x, float * y) {
 int i = blockIdx.x * blockDim.x + threadIdx.x;
 if (i < n)
   v[i] = a * x[i] + v[i]:
int a = 42;
int n = 10;
float x[n], y[n];
// fill x, y
cudaMallocManaged(&x. n * sizeof(float));
cudaMallocManaged(&y, n * sizeof(float));
saxpy cuda<<<2, 5>>>(n, a, x, y);
```

CUDA SAXPY

```
With runtime-managed data transfers
```

```
Specify kernel
global ← void saxpy cuda(int n, float a, float * x, float * y) {
  int i = blockIdx.x * blockDim.x + threadIdx.x:
                                                                                  ID variables
  if (i < n)•
    v[i] = a * x[i] + v[i]:
                                                                               Guard against
                                                                              too many threads
int a = 42;
int n = 10;
float x[n], y[n];
                                                                          Allocate GPU-capable
// fill x, y
cudaMallocManaged(&x. n * sizeof(float)):
                                                                              Call kernel
cudaMallocManaged(&y, n * sizeof(float));
                                                                        2 blocks, each 5 threads
saxpv cuda<<<2, 5>>>(n, a, x, v);
                                                                                   Wait for
```

kernel to finish

cudaDeviceSvnchronize():

Programming GPUs

Performance Analysis

GPU Tools

The helpful helpers helping helpless (and others)

NVIDIA

Nsight Systems GPU program profiler with timeline Nsight Compute GPU kernel profiler

OpenCL: CodeXL (Open Source, GPUOpen/AMD) – debugging, profiling.



Nsight Systems

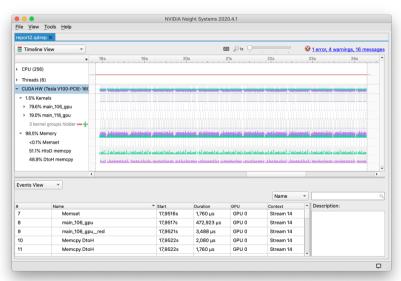
CLI

```
$ nsvs profile --stats=true ./poisson2d 10 # (shortened)
CUDA APT Statistics:
 Time(%) Total Time (ns) Num Calls
                                      Average
                                                  Minimum
                                                              Maximum
                                                                                Name
    90.9
              160,407,572
                                   5.346.919.1
                                                 1.780 25.648.117 cuStreamSynchronize
CUDA Kernel Statistics:
 Time(%)
          Total Time (ns) Instances
                                      Average
                                                  Minimum
                                                              Maximum
                                                                              Name
                                10 15,868,661.7 14,525,819 25,652,783 main_106_gpu
   100.0
              158.686.617
     0.0
                                         2.512.0
                                                                  3,680 main 106 gpu_red
                  25.120
                                10
                                                      2.304
```



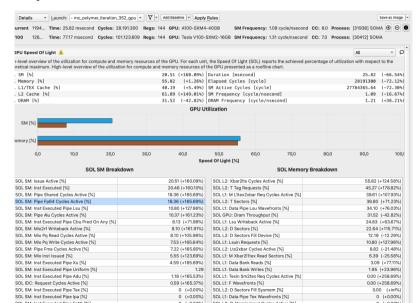
Nsight Systems

GUI



Nsight Compute

GUI



Advanced Topics

So much more interesting things to show!

- Optimize memory transfers to reduce overhead
- Optimize applications for GPU architecture
- Drop-in BLAS acceleration with NVBLAS (\$LD_PRELOAD)
- Tensor Cores for Deep Learning
- Libraries, Abstractions: Kokkos, Alpaka, Futhark, HIP, SYCL, C++AMP, C++ pSTL, ...
- Use multiple GPUs
 - On one node
 - Across many nodes o MPI



- ..
- Some of that: Addressed at dedicated training courses

Using GPUs on JSC Systems

Compiling

CUDA

- Module: module load CUDA/11.0
- Compile: nvcc file.cu Default host compiler: g++; use nvcc_pgc++ for PGI compiler
- Example cuBLAS: g++ file.cpp -I\$CUDA_HOME/include -L\$CUDA_HOME/lib64 -lcublas -lcudart

OpenACC

- Module: module load NVHPC/20.9-GCC-9.3.0
- Compile: nvc++ -acc=gpu file.cpp

MPI CUDA-aware MPIs (with direct Device-Device transfers)

ParaStationMPI module load ParaStationMPI/5.4.7-1 mpi-settings/CUDA

OpenMPI module load OpenMPI/4.1.0rc1 mpi-settings/CUDA

Running

Dedicated GPU partitions

```
JUWELS
```

```
--partition=gpus 46 nodes (Job limits: \leq 1 d) 
--partition=develgpus 10 nodes (Job limits: \leq 2 h, \leq 2 nodes)
```

JUWELS Booster

```
--partition=booster 926 nodes 
--partition=develbooster 10 nodes (Job limits: \leq 1 d, \leq 2 nodes)
```

JURECA DC

```
--partition=dc-gpu 192 nodes --partition=dc-gpu-devel ?? nodes
```

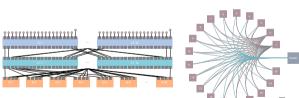
- Needed: Resource configuration with --gres=gpu:4
- → See online documentation



Running

JUWELS Booster Topology

- JUWELS Booster: NPS-4 (in total: 8 NUMA Domains)
- Not all have GPU or HCA affinity!
- Network is structured into two levels: In-Cell and Inter-Cell (DragonFly+ network)



→ Documentation: apps.fz-juelich.de/jsc/hps/juwels/



Example

- 16 tasks in total, running on 4 nodes
- Per node: 4 GPUs

```
#!/bin/bash -x
#SBATCH --nodes=4
#SBATCH --ntasks=16
#SBATCH --ntasks-per-node=4
#SBATCH --output=gpu-out.%j
#SBATCH --error=gpu-err.%j
#SBATCH --time=00:15:00
#SBATCH --partition=gpus
#SBATCH --gres=gpu:4
```

srun ./gpu-prog



Conclusion

- GPUs provide highly-parallel computing power
- We have many devices installed at JSC, ready to be used!



- GPUs provide highly-parallel computing power
- We have many devices installed at JSC, ready to be used!
- Training courses by JSC
 CUDA Course April 2021
 OpenACC Course October 2021
- Generally: see online documentation and sc@fz-juelich.de



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- Further consultation via our lab: NVIDIA Application Lab in Jülich; contact me!



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Appendix

Appendix Glossary References



Glossary I

- AMD Manufacturer of CPUs and GPUs. 49, 50, 51, 52, 53, 54, 84, 86
- Ampere GPU architecture from NVIDIA (announced 2019). 4, 5, 6
 - API A programmatic interface to software by well-defined functions. Short for application programming interface. 49, 50, 51, 52, 53, 54
 - CUDA Computing platform for GPUs from NVIDIA. Provides, among others, CUDA C/C++. 2, 48, 49, 50, 51, 52, 53, 54, 55, 56, 57, 58, 59, 60, 61, 62, 63, 64, 65, 73, 78, 79, 80, 81, 86
 - HIP GPU programming model by AMD to target their own and NVIDIA GPUs with one combined language. Short for Heterogeneous-compute Interface for Portability. 49, 50, 51, 52, 53, 54



Glossary II

JSC Jülich Supercomputing Centre, the supercomputing institute of Forschungszentrum Jülich, Germany. 2, 78, 79, 80, 81, 85

JURECA A multi-purpose supercomputer at JSC. 6

JUWELS Jülich's new supercomputer, the successor of JUQUEEN. 3, 4, 5, 74

MPI The Message Passing Interface, a API definition for multi-node computing, 71, 73

NVIDIA US technology company creating GPUs. 3, 4, 5, 6, 26, 27, 28, 49, 50, 51, 52, 53, 54, 67, 78, 79, 80, 81, 84, 86

OpenACC Directive-based programming, primarily for many-core machines, 43, 44, 45, 46, 47, 73, 78, 79, 80, 81



Glossary III

- OpenCL The *Open Computing Language*. Framework for writing code for heterogeneous architectures (CPU, GPU, DSP, FPGA). The alternative to CUDA. 49, 50, 51, 52, 53, 54, 67
- OpenMP Directive-based programming, primarily for multi-threaded machines. 43, 44, 45
 - ROCm AMD software stack and platform to program AMD GPUs. Short for Radeon Open Compute (*Radeon* is the GPU product line of AMD). 49, 50, 51, 52, 53, 54
 - SAXPY Single-precision $A \times X + Y$. A simple code example of scaling a vector and adding an offset. 31, 64, 65
 - Tesla The GPU product line for general purpose computing computing of NVIDIA. 3



Glossary IV

- CPU Central Processing Unit. 3, 6, 10, 11, 12, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 31, 49, 50, 51, 52, 53, 54, 84, 86
- GPU Graphics Processing Unit. 2, 3, 4, 5, 6, 7, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 30, 34, 35, 36, 37, 38, 39, 42, 43, 44, 45, 48, 49, 50, 51, 52, 53, 54, 65, 66, 67, 71, 72, 74, 75, 76, 78, 79, 80, 81, 84, 85, 86
- SIMD Single Instruction, Multiple Data. 19, 20, 21, 22, 23, 24, 25, 26, 27, 28
- SIMT Single Instruction, Multiple Threads. 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28
 - SM Streaming Multiprocessor. 19, 20, 21, 22, 23, 24, 25, 26, 27, 28
- SMT Simultaneous Multithreading. 19, 20, 21, 22, 23, 24, 25, 26, 27, 28



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