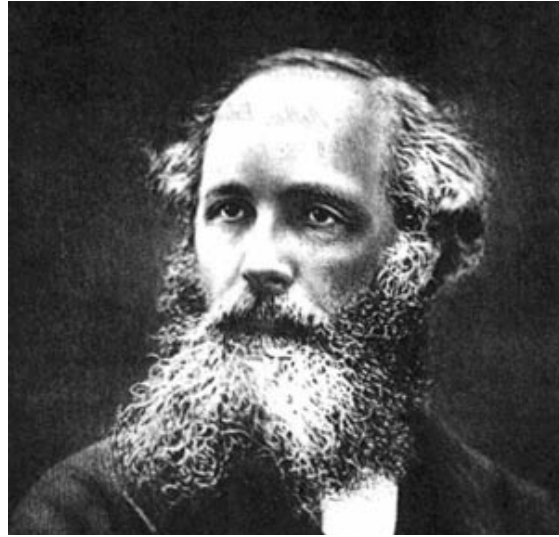




Johannes Kepler



James Maxwell



Blaise Pascal



Alessandro Volta

Who would win a 100 Meter Sprint?

Is Data Placement Optimization Still Relevant On Newer GPUs?

**Md Abdullah Shahneous Bari¹, Larisa Stoltzfus²,
Pei-Hung Lin³, Chunhua Liao³, Murali Emani³, Barbara Chapman^{1,4}**

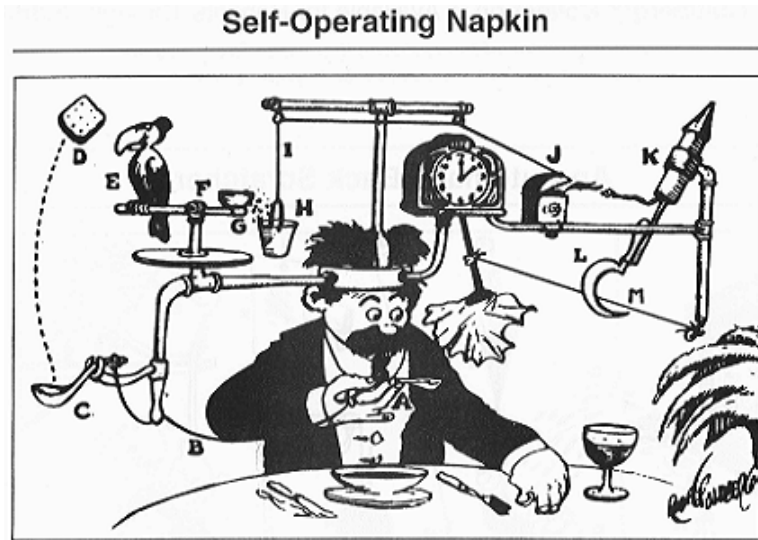
1. Stony Brook University 2. University of Edinburgh
3. Lawrence Livermore National Laboratory 4. Brookhaven National Laboratory



What is a GPU? Should I Really Care?

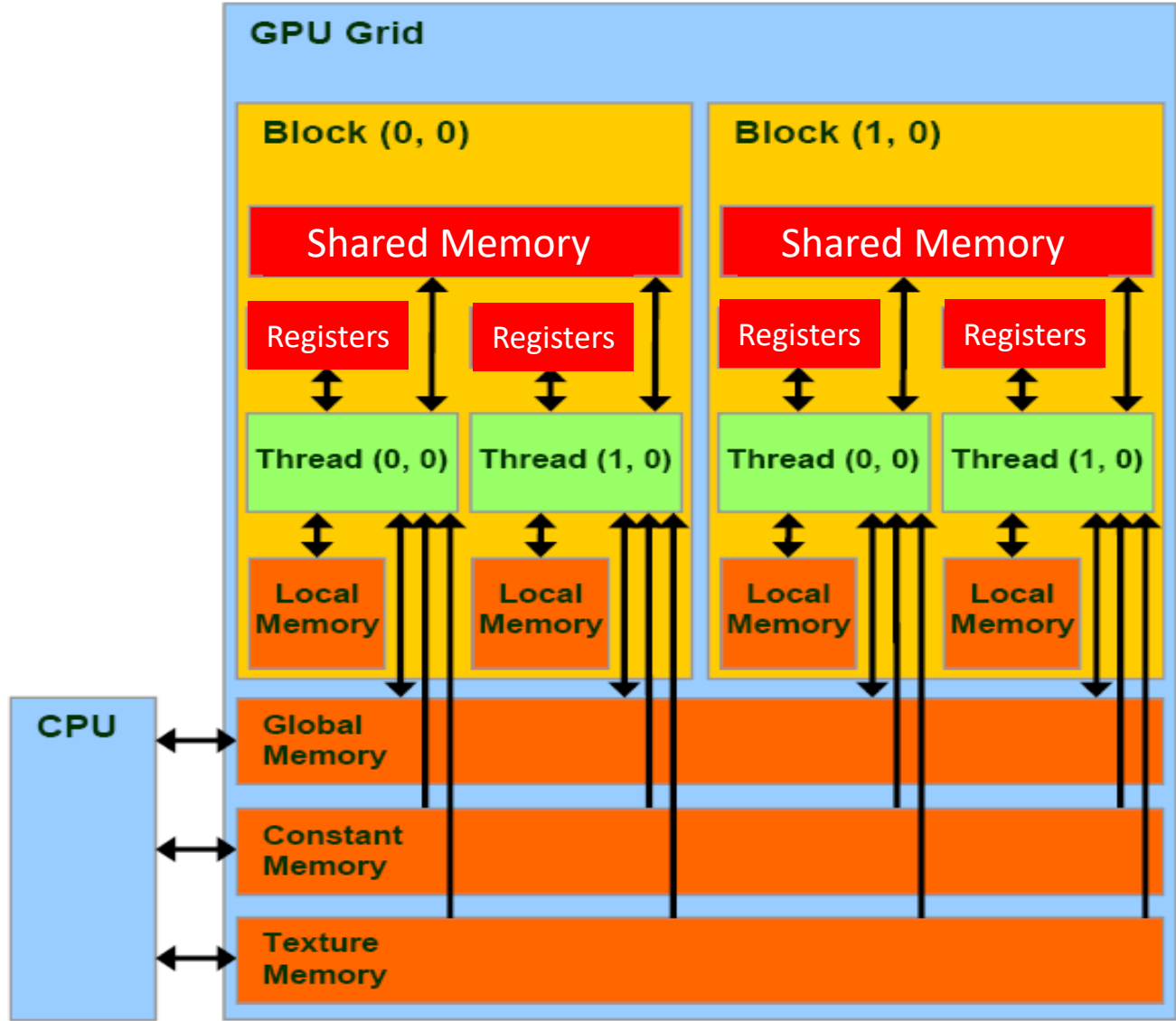
- Graphics Processing Unit
- Massively parallel
- Thousands of cores
- Now-a-days used for general purpose computing
- 5 of the top 10 supercomputers uses NVIDIA GPU
- Aand..... Deep Learning

Easy to Use???



- Programmability: OpenMP, OpenACC, **CUDA**, OpenCL
- Getting performance improvement
 - Easy if the algorithm is compliant
- Getting GOOD performance
 - Not so easy
- But, but, why??
 - **Complex memory hierarchy**

NVIDIA GPU
Memory
Hierarchy



Global/Device Memory

- Largest off-chip memory
- Serves as the main memory
- Long latency
- Limited bandwidth

Constant Memory

- Predefined part of global memory
- Cached and globally visible to all threads
- Read-only
- Can be as fast as cache
- Limited size
- Good for read-only data that needs to be repetitively broadcast to all GPU threads

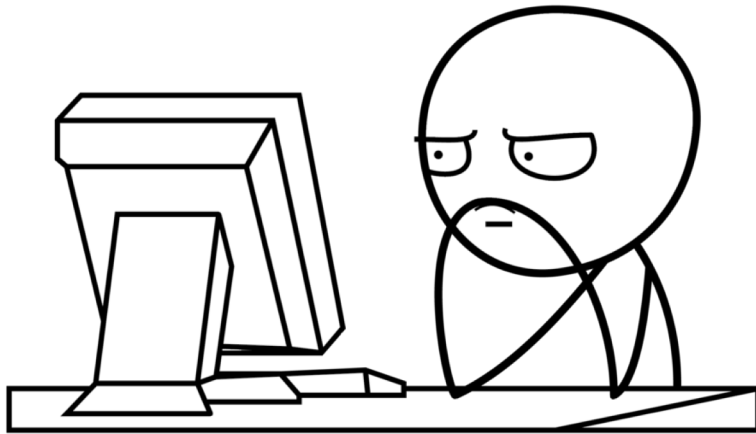
Shared Memory

- Software managed on-chip data cache
- Per Streaming Multiprocessor (SM)
- Limited size
- Low latency and high bandwidth

Texture Memory

- Off-chip, cached and read-only
- Actual memory bound to device memory
- Can occupy the whole device memory bound to the texture unit
- Texture cache specially optimized for 2-D, 3-D spatial locality

How to Get GOOD Performance?



- Duh, use the memory hierarchy well
 - Optimize your code
 - Place your data in appropriate memories
- Not an easy job
- Not to mention, change in hierarchy could undo everything
- NVIDIA tend to change the hierarchy almost in every generation

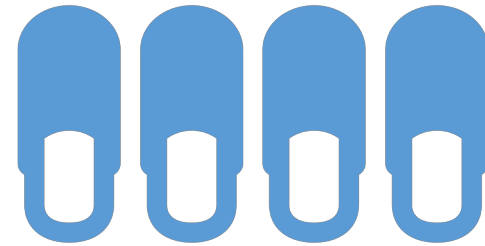


Well, What Are the Engineers at NVIDIA doing???
Is It Any Better Now???

Let's Figure It Out



How has the impact of data placement changed over time?



Four kinds of memory

Global memory (GPU DRAM)

Constant memory

Shared memory

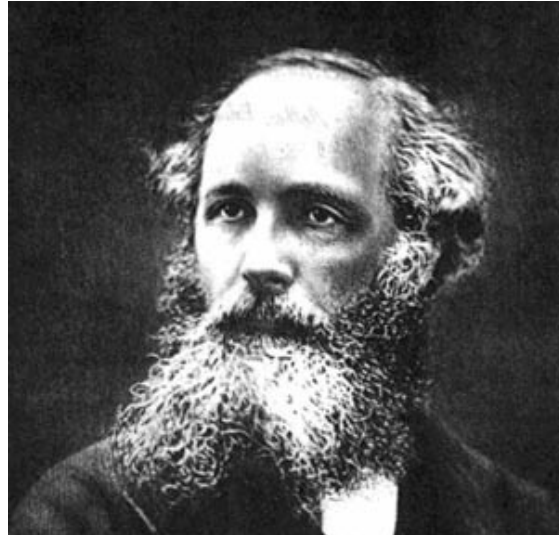
Texture memory



Designing The Experiments



Johannes Kepler



James Maxwell

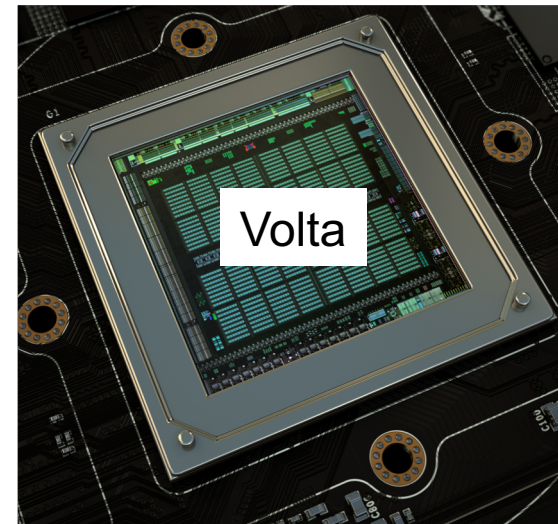
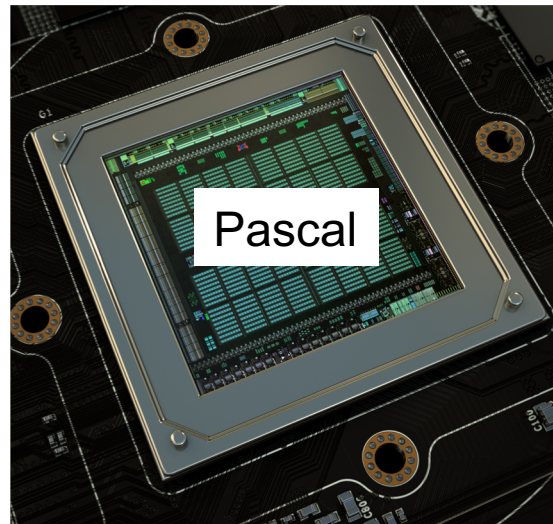
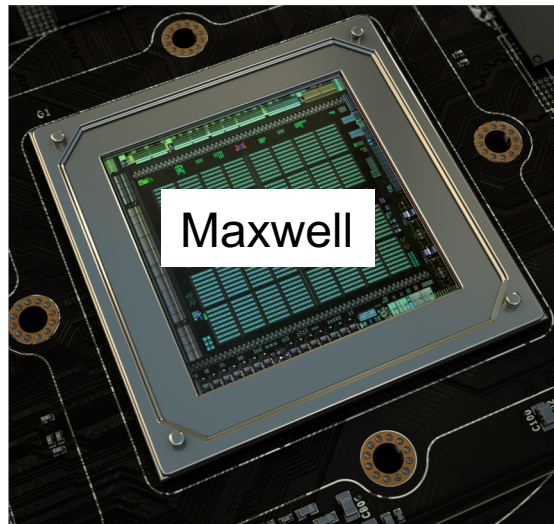
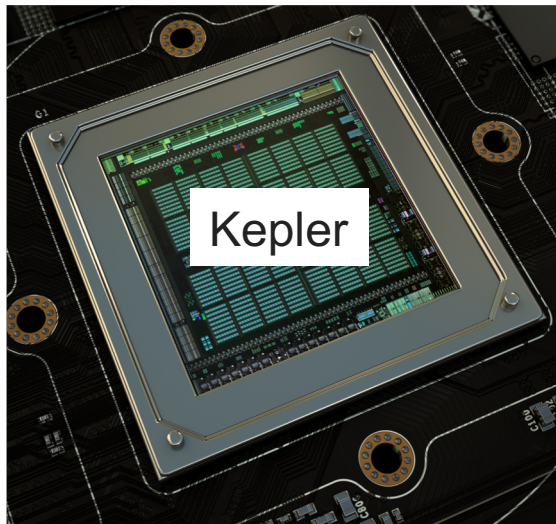


Blaise Pascal



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Who would win a 100 Meter Sprint?



GPUs?

Applications

Microbenchmarks

CUDA Kernels

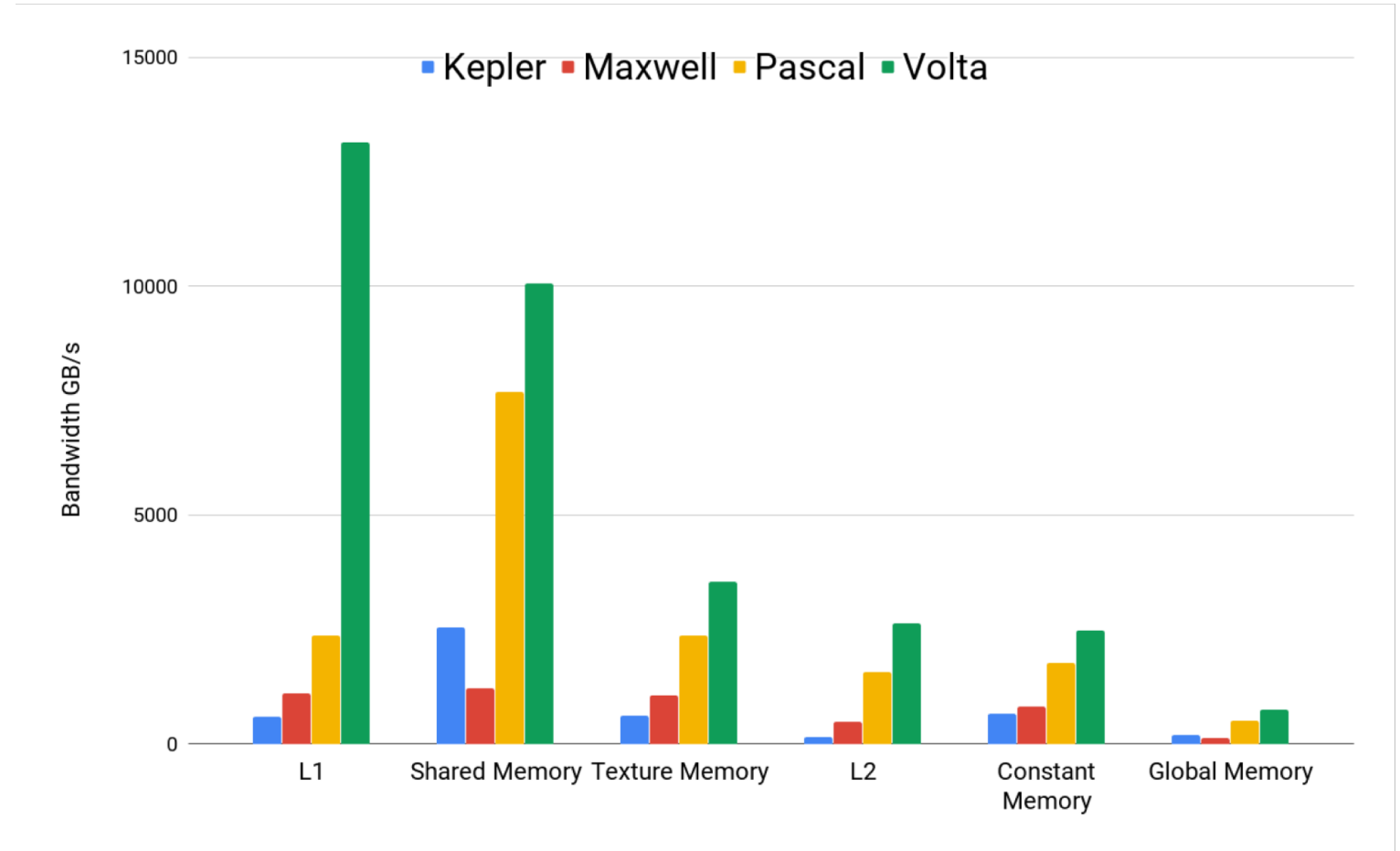
- Using only one type of special memory
- Using mixed types of special memories

Proxy App

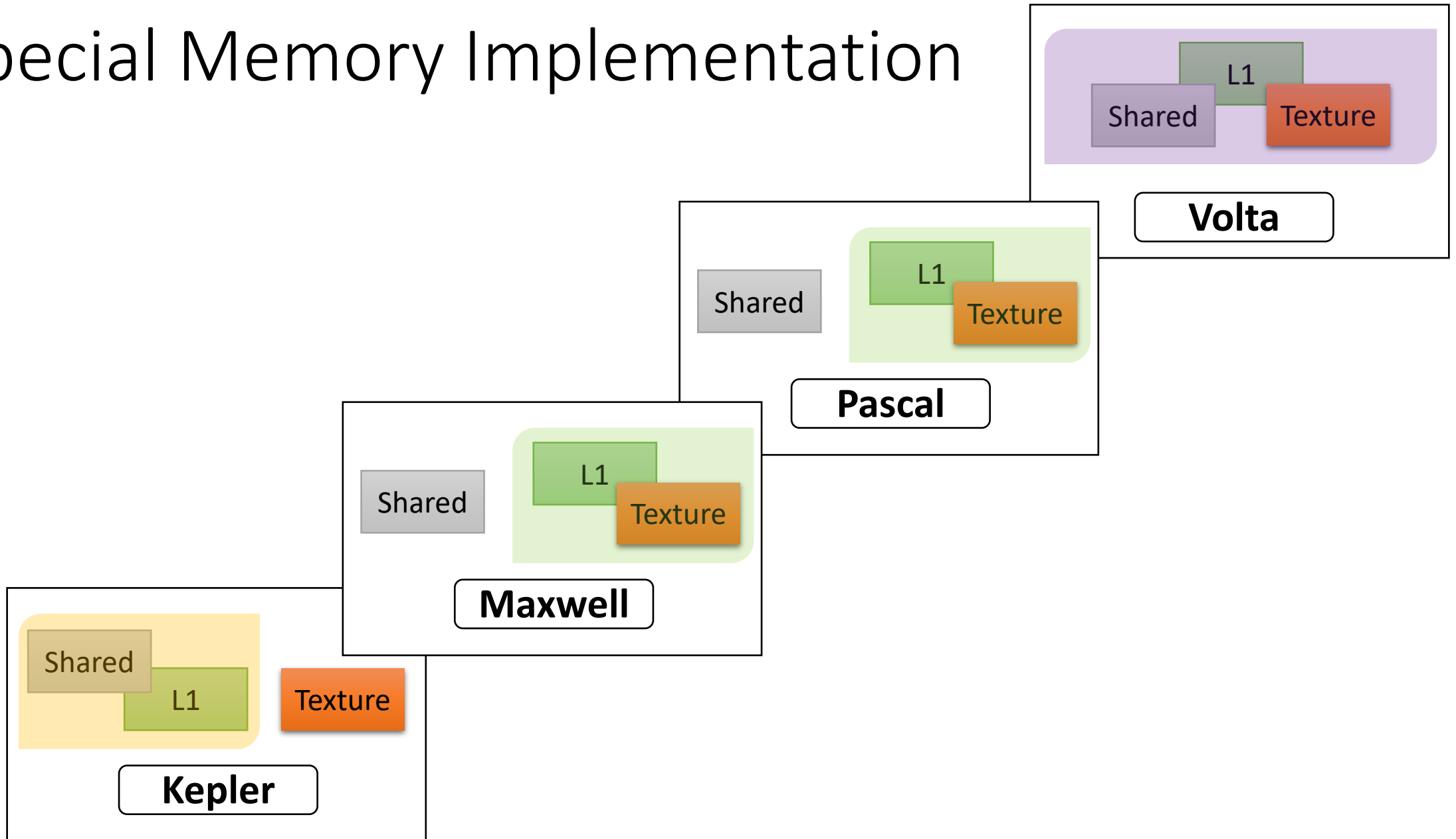
Microbenchmark

- Used several microbenchmarks
 - GPUmembench
 - Pointer chasing benchmark
- Measured metrics
 - Global, Constant, Shared, Texture memory and L1, L2 cache properties
 - Size, latency, bandwidth etc.

Bandwidth Across GPUs



Special Memory Implementation

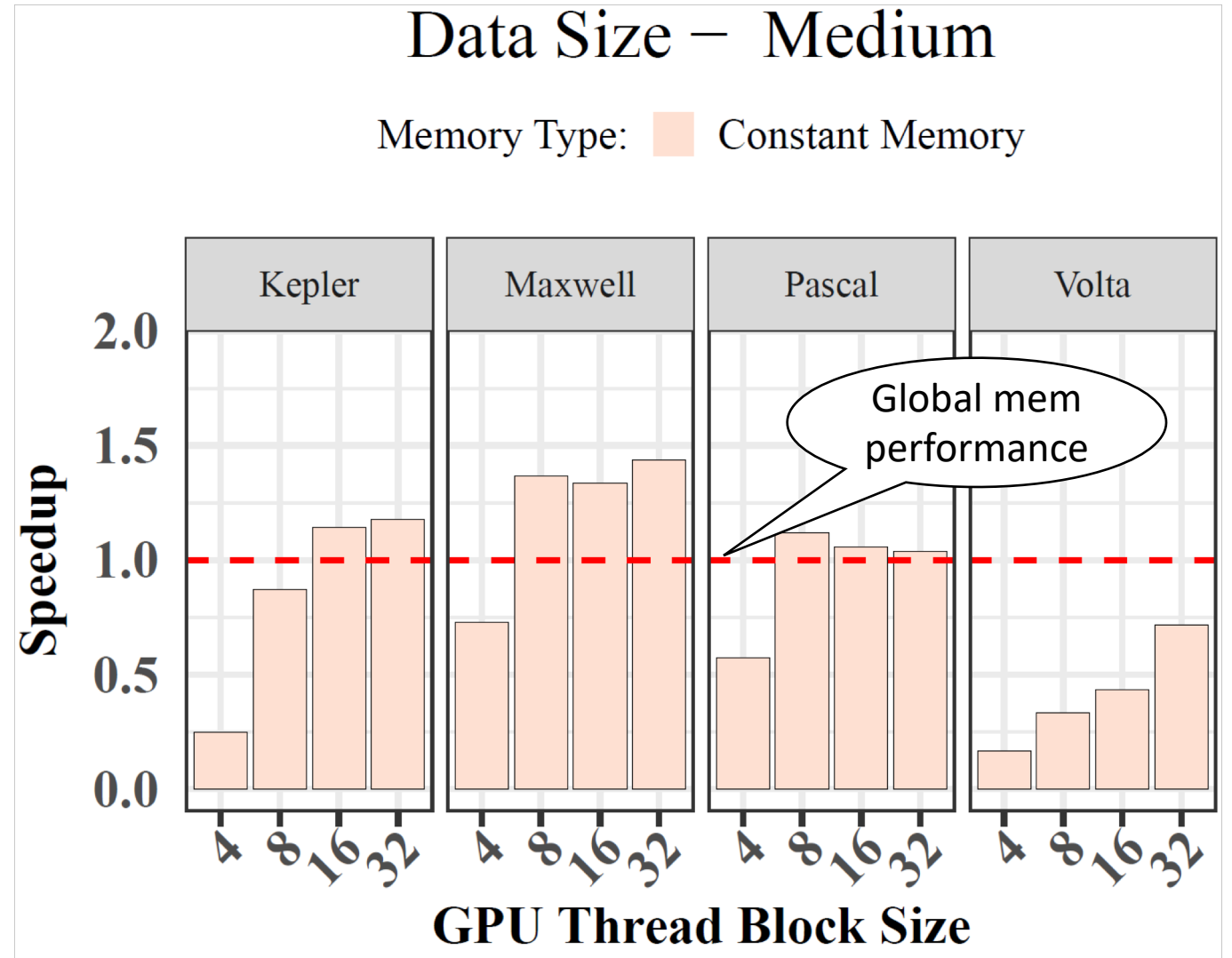


CUDA Kernels – Single Type of Memory

- Used 3 representative kernels that historically showed good performance with special memories
 - Ray Tracing – Constant memory
 - Matrix Matrix Multiplication – Shared memory
 - Heat Transfer Simulation – Texture memory
- Other configurable parameters
 - Different data sizes
 - GPU Thread block size

Constant Memory

With newer generations, constant memory data placement increasingly insignificant



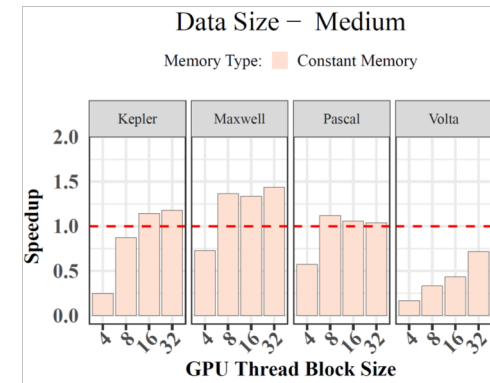
Analysis



Improvement in Global memory bandwidth along with L1, L2 cache

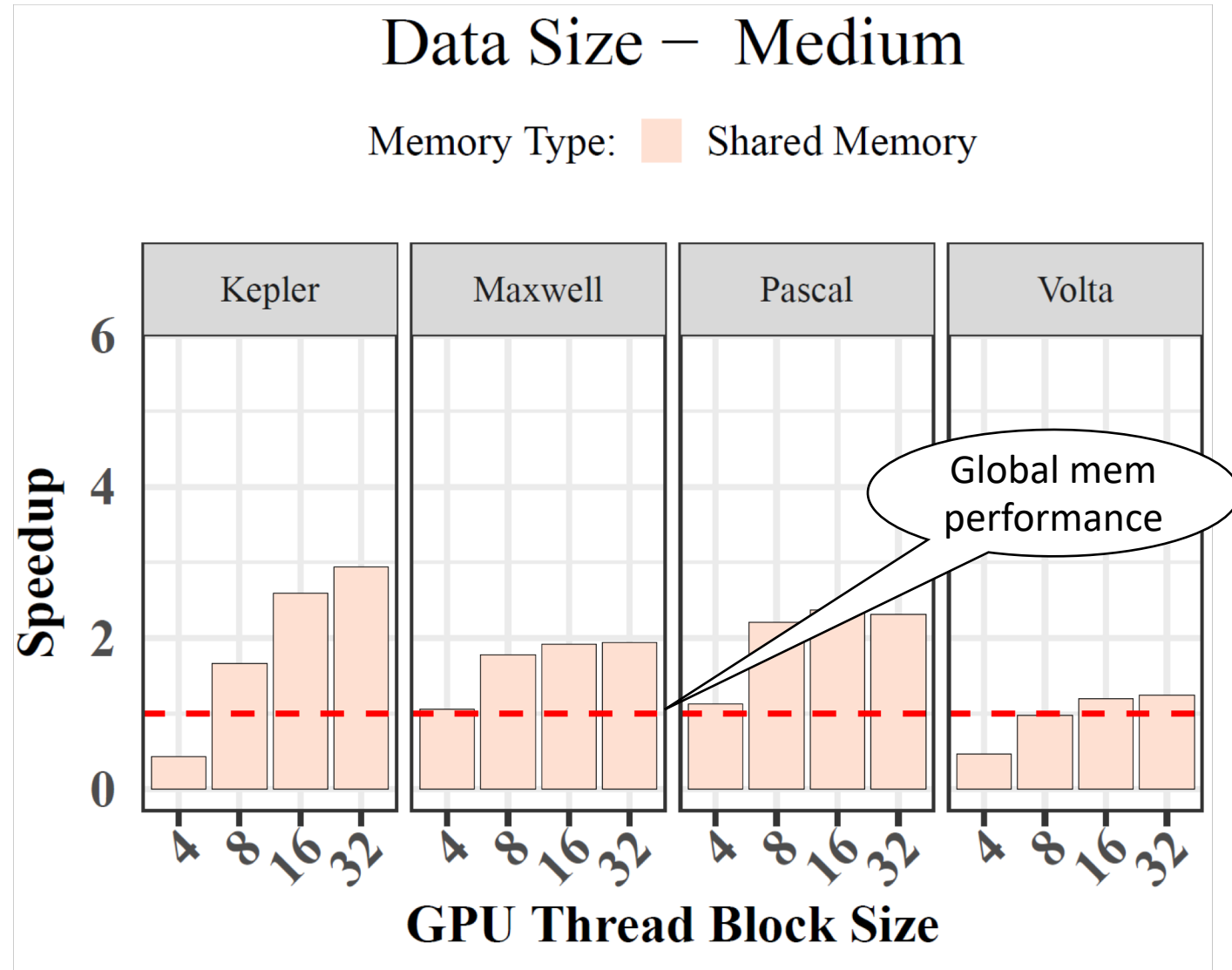


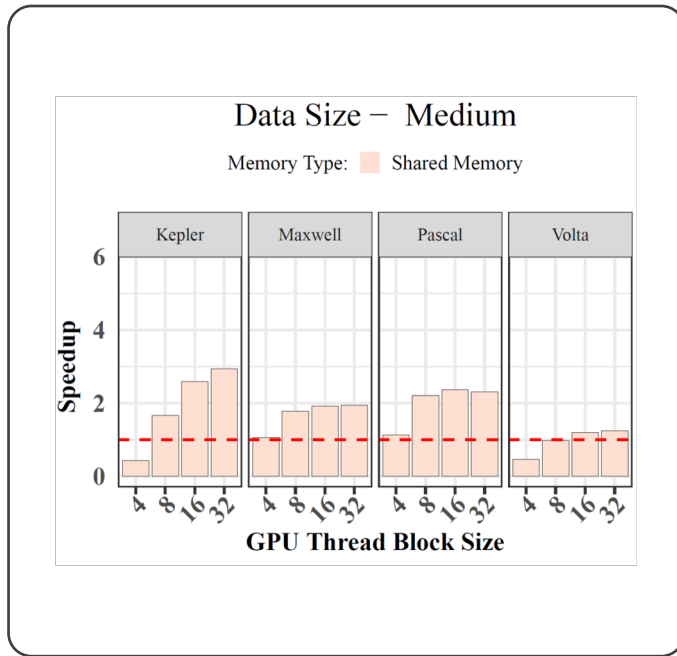
Increasing percentage of stalls due to pipeline busy



Shared Memory

With newer generations, data placement increasingly insignificant





HBM2 and unified memory design results in Volta global memory performance improvement

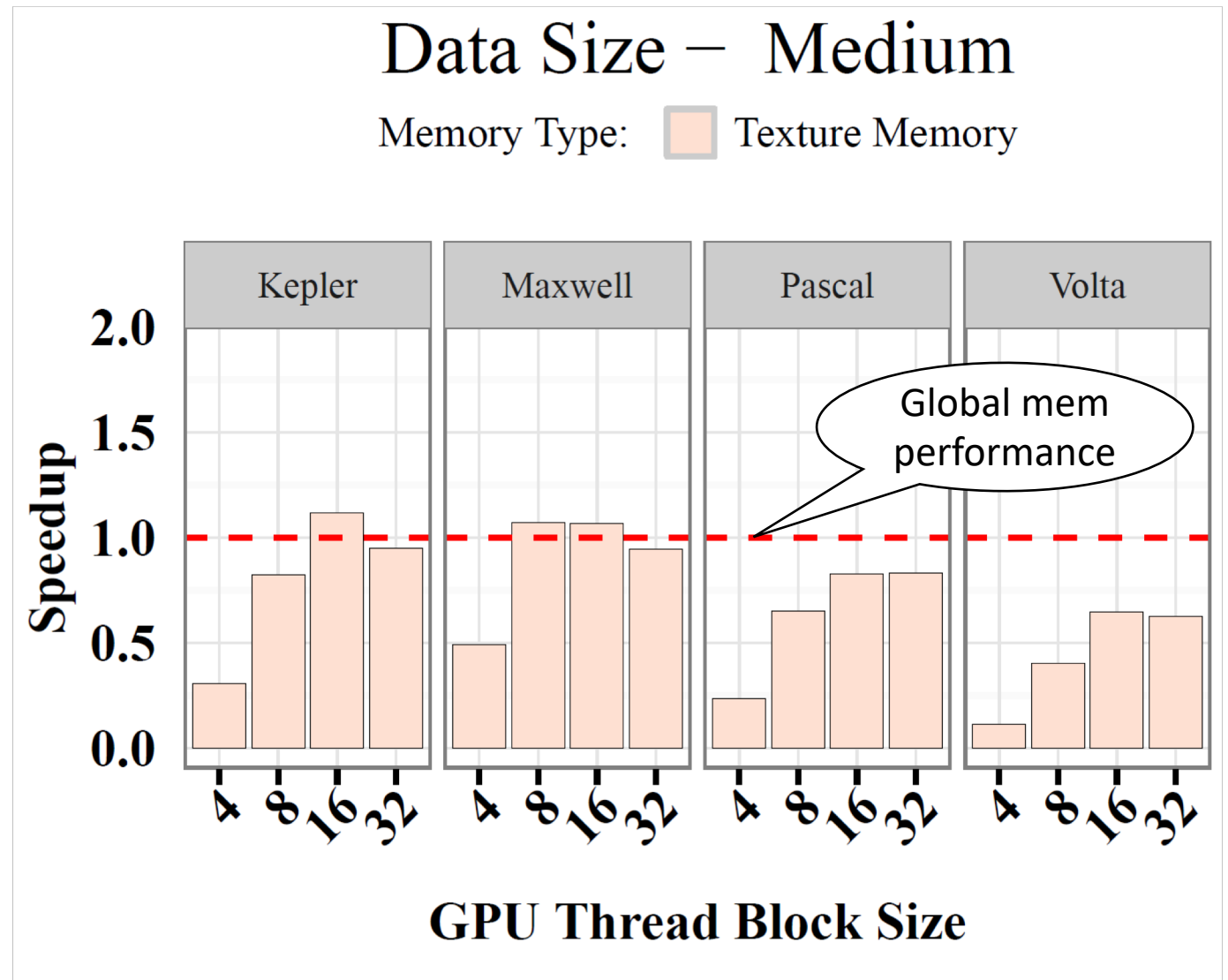


Why not Pascal then?
- Special memory not unified

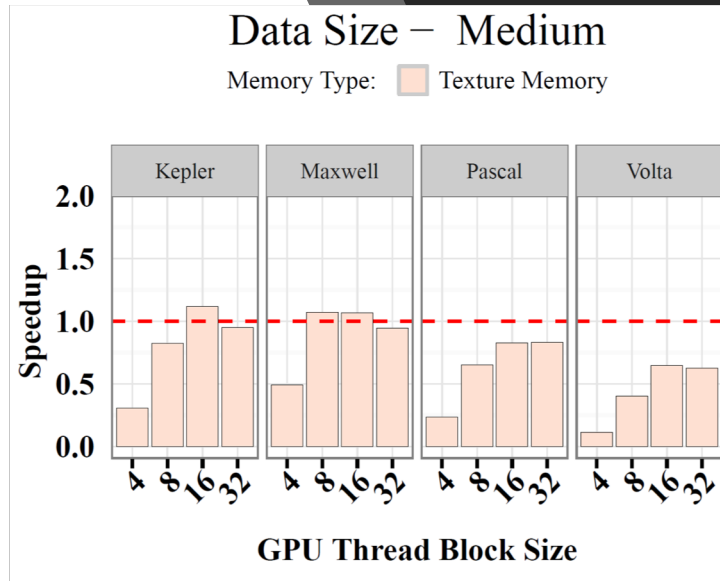
Analysis

Texture Memory

With newer generations, data placement increasingly insignificant



Analysis



Memory design?

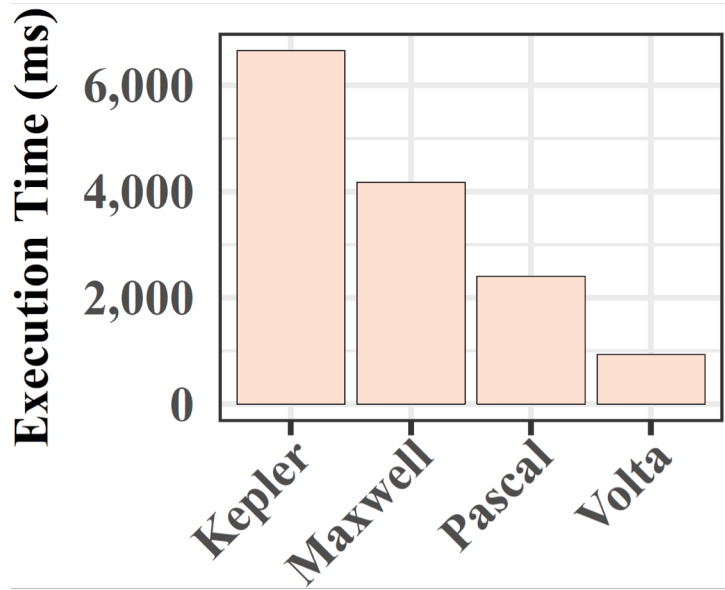
- L1 and Texture cache in the same unit in Maxwell, Pascal and Volta



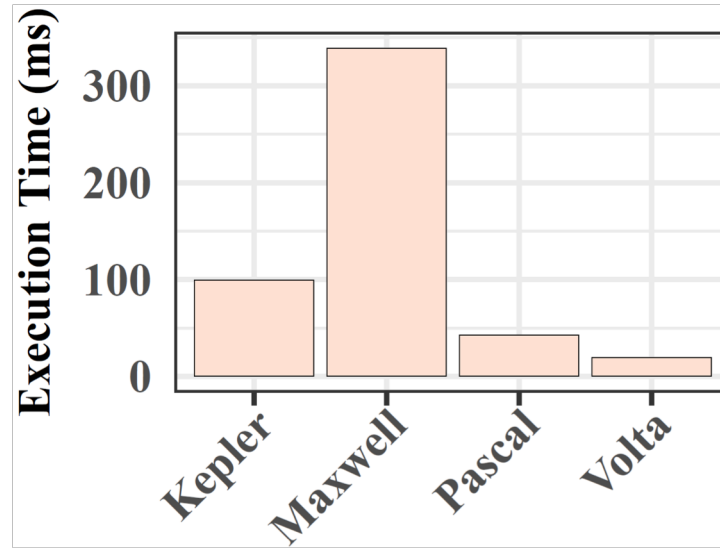
Bandwidth?

- Global memory, L1 out weighs Texture

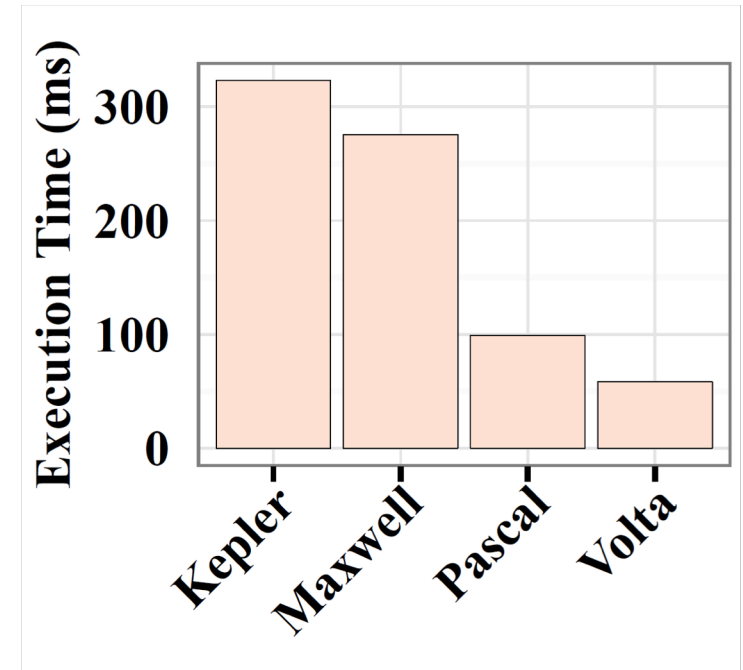
Constant memory



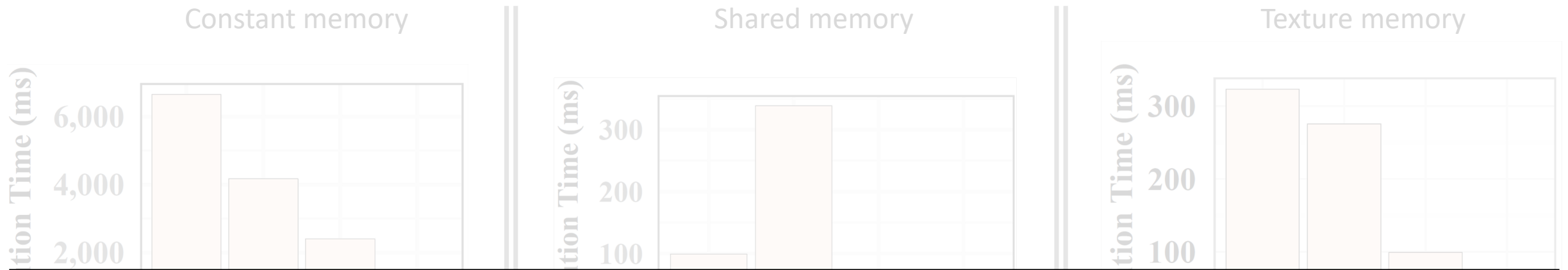
Shared memory



Texture memory



Special Memory Units' Performance Improvement Across GPUs



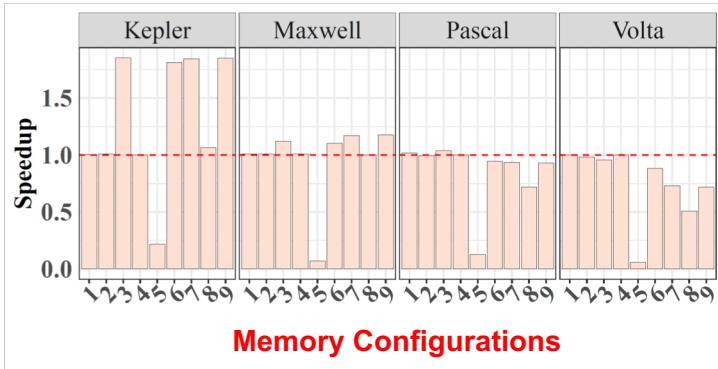
All types of memories on newer GPUs have improved performances

Special Memory Units' Performance Improvement Across GPUs

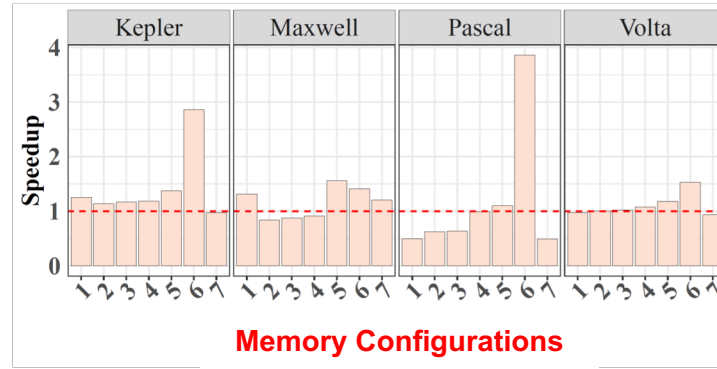
CUDA Kernels – Mixed Types of Memory

- Used 3 representative kernels
 - Sparse Matrix - Vector Multiplication (SPMV)
 - Matrix - Matrix Multiplication
 - Computational Fluid Dynamics
- Each data placement configuration uses multiple types of special memories

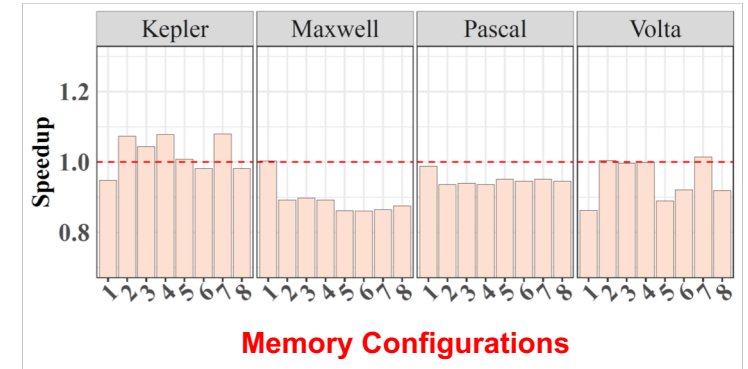
SPMV



MM



CFD



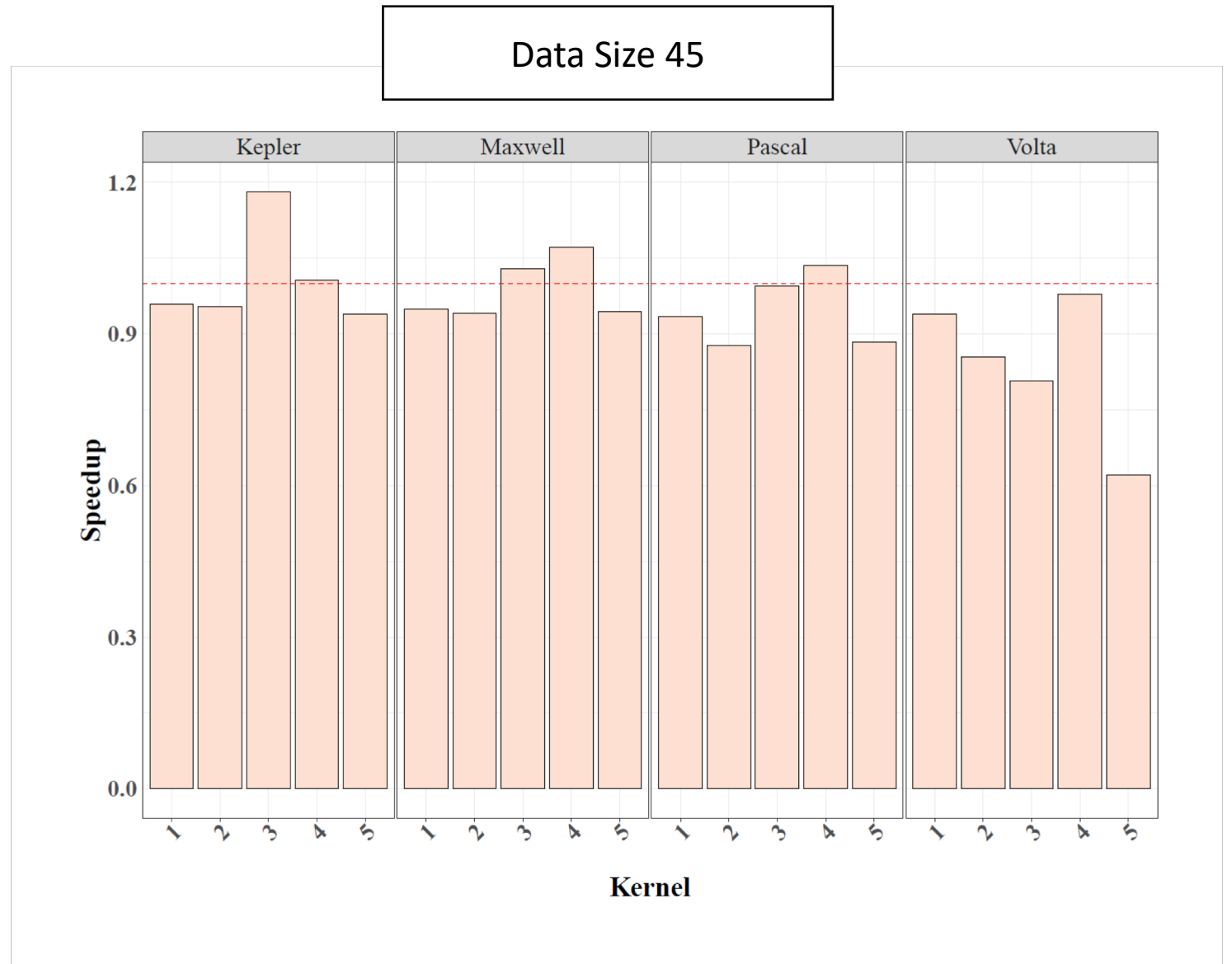
Speedup

Proxy App - Lulesh

- Used mixed memory implementation
- Using special memory in real application is cumbersome
- Require a lot of modification
- Used 3 different data sizes (e.g., 4, 45, 90)
 - For larger data sizes (e.g., 45, 90) unable to use constant, shared memory

Speedup

Memory properties of special memories significantly limit their usage in real application



Key Takeaways



- All types of memories on newer GPUs have improved performances
- Global memory bandwidth, unified cache design helps narrow the performance gap between global and special memories
- Memory properties of special memories significantly limit their usage in real application

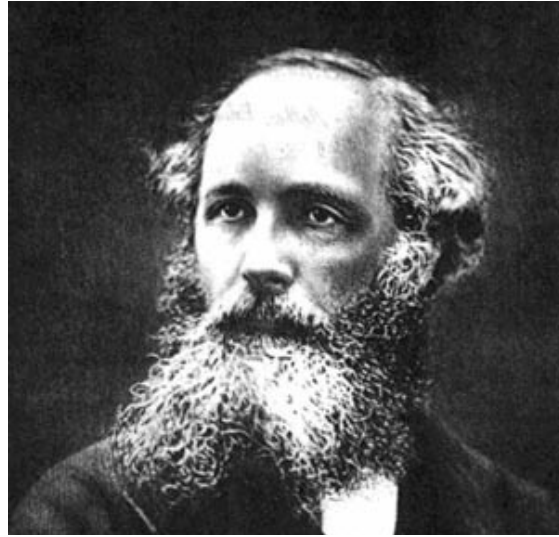


Future Work

- Investigate the data placement optimization on energy consumption
- Automated code transformation to exploit special memories



Johannes Kepler



James Maxwell



Blaise Pascal



Alessandro Volta

So, Who Won?



STONY BROOK
THE UNIVERSITY OF NEW YORK



Thank you 😊

- All types of memories on newer GPUs have improved performances
- Global memory bandwidth, unified cache design helps narrow the performance gap between global and special memories
- Memory properties of special memories significantly limit their usage in real application