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# Exploiting Hardware-Accelerated Ray Tracing for Monte Carlo Particle Transport with OpenMC

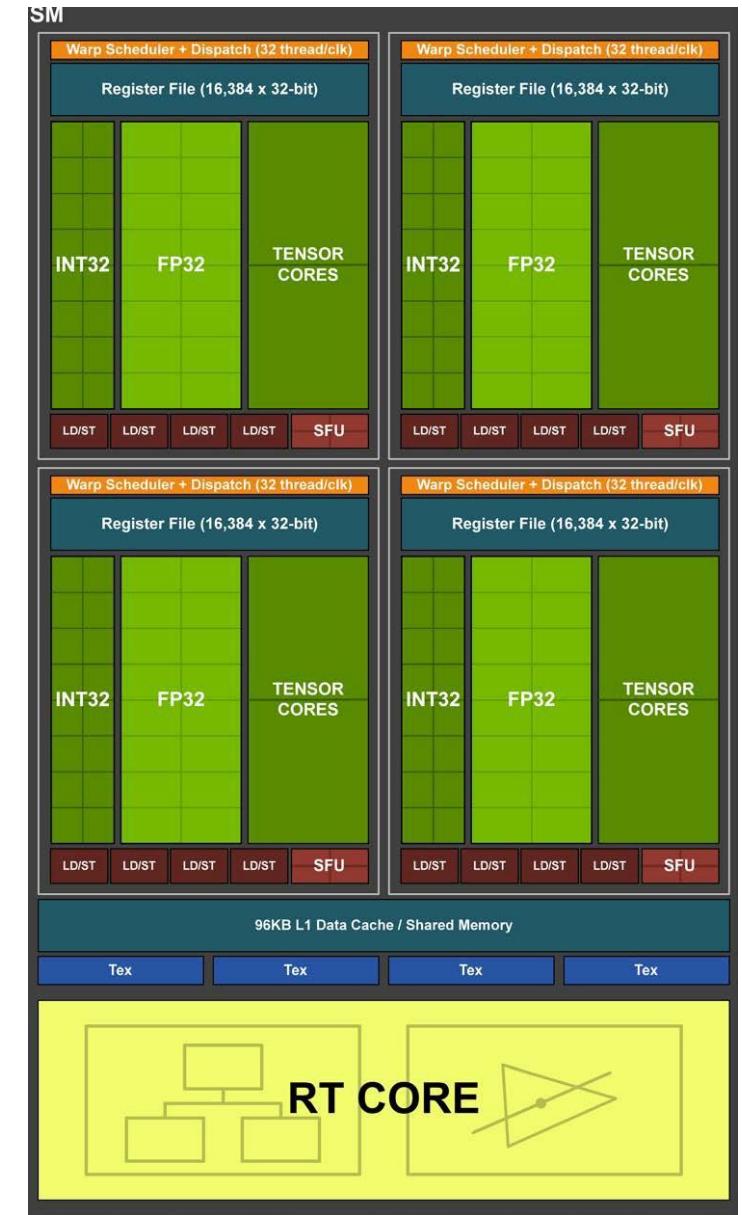
# Increased heterogeneity is an important response to the slowing of Moore's Law

- Expect to see more “application-oriented” optimisations
  - Matrix multiply units in SIMD instruction sets (AVX, SVE)
  - Floating point formats optimized for deep learning (BFLOAT16)
- Important recent example: TensorCores

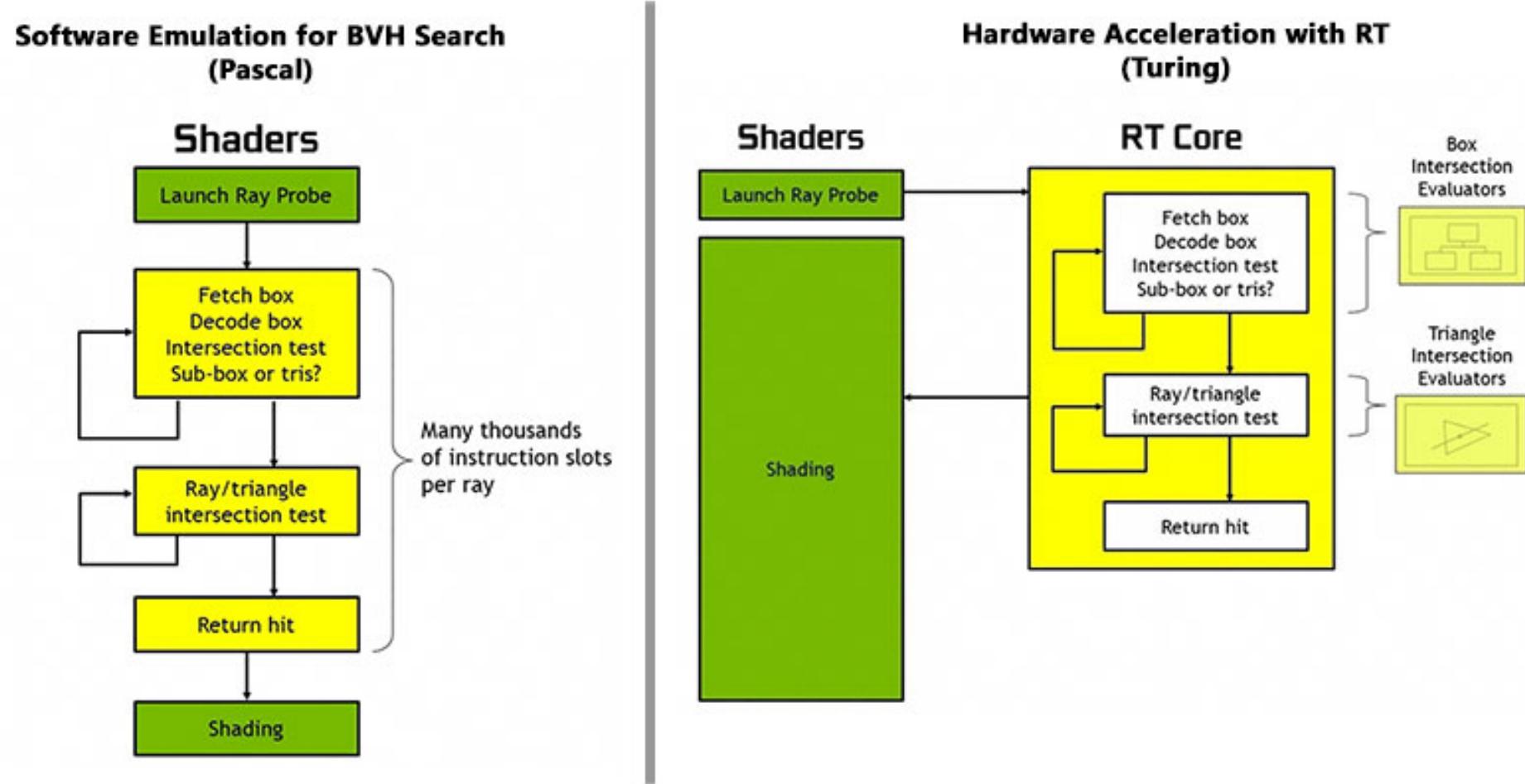
$$D = \begin{pmatrix} A_{0,0} & A_{0,1} & A_{0,2} & A_{0,3} \\ A_{1,0} & A_{1,1} & A_{1,2} & A_{1,3} \\ A_{2,0} & A_{2,1} & A_{2,2} & A_{2,3} \\ A_{3,0} & A_{3,1} & A_{3,2} & A_{3,3} \end{pmatrix}_{\text{FP16 or FP32}} \begin{pmatrix} B_{0,0} & B_{0,1} & B_{0,2} & B_{0,3} \\ B_{1,0} & B_{1,1} & B_{1,2} & B_{1,3} \\ B_{2,0} & B_{2,1} & B_{2,2} & B_{2,3} \\ B_{3,0} & B_{3,1} & B_{3,2} & B_{3,3} \end{pmatrix}_{\text{FP16}} + \begin{pmatrix} C_{0,0} & C_{0,1} & C_{0,2} & C_{0,3} \\ C_{1,0} & C_{1,1} & C_{1,2} & C_{1,3} \\ C_{2,0} & C_{2,1} & C_{2,2} & C_{2,3} \\ C_{3,0} & C_{3,1} & C_{3,2} & C_{3,3} \end{pmatrix}_{\text{FP16 or FP32}}$$

# Ray Tracing cores

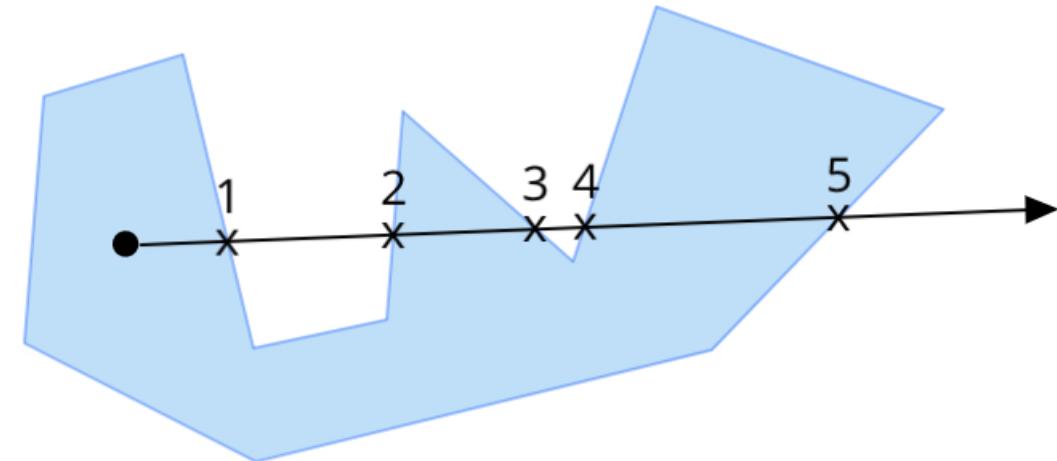
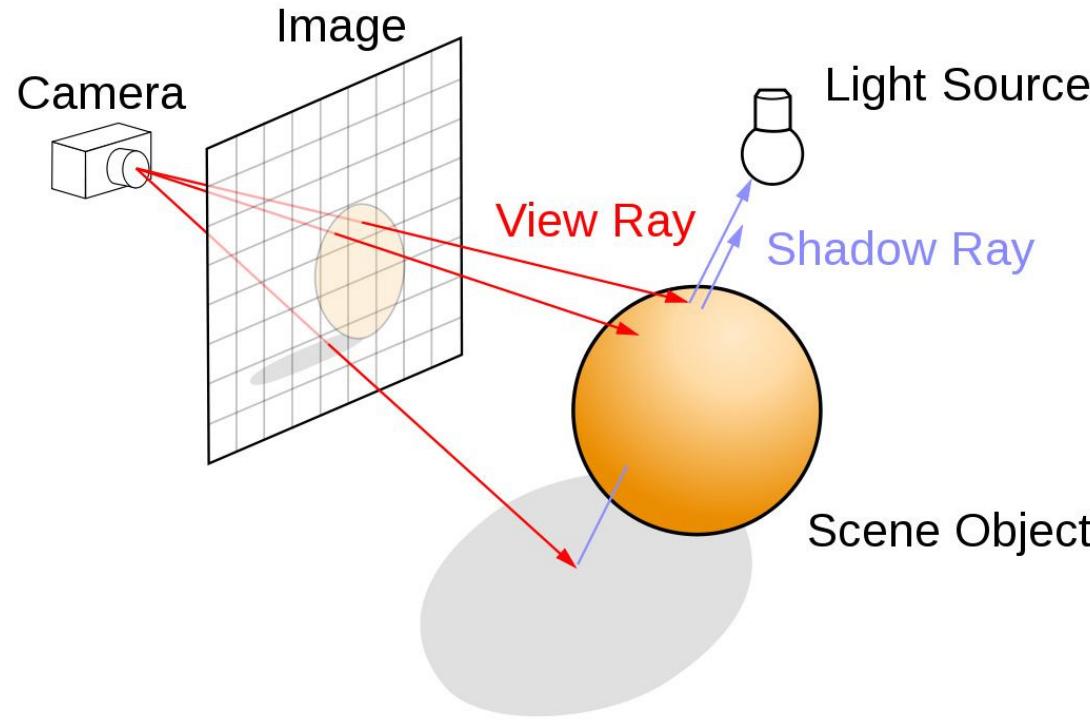
- NVIDIA's latest architectural innovation (Turing-class GPUs)
- Designed to accelerate the ray tracing algorithms used in graphical rendering in games, rather than for HPC
- Potential speedups of up to 10X vs CUDA code on same GPU
- Accelerates ray / surface intersection calculations
  - 10 GigaRays/s on RT cores vs 1-2 GigaRays/s in CUDA on the same GPU



# Which parts of ray tracing do the RT cores accelerate?



# Observation: Monte Carlo particle transport has similarities to RT



Both require large numbers of linear geometric queries  
to be executed over complex 3D geometric models

# Monte Carlo particle transport

- MC particle transport has applications in fission and fusion reactor design, radiography, and accelerator design
- Requires large numbers of particles  $\geq O(10^6)$ , therefore computationally expensive
- Many codes and mini-apps developed for MC particle transport:
  - OpenMC, MCNP, Quicksilver, Branson, neutral [1]
- We've focused on **OpenMC** for this work

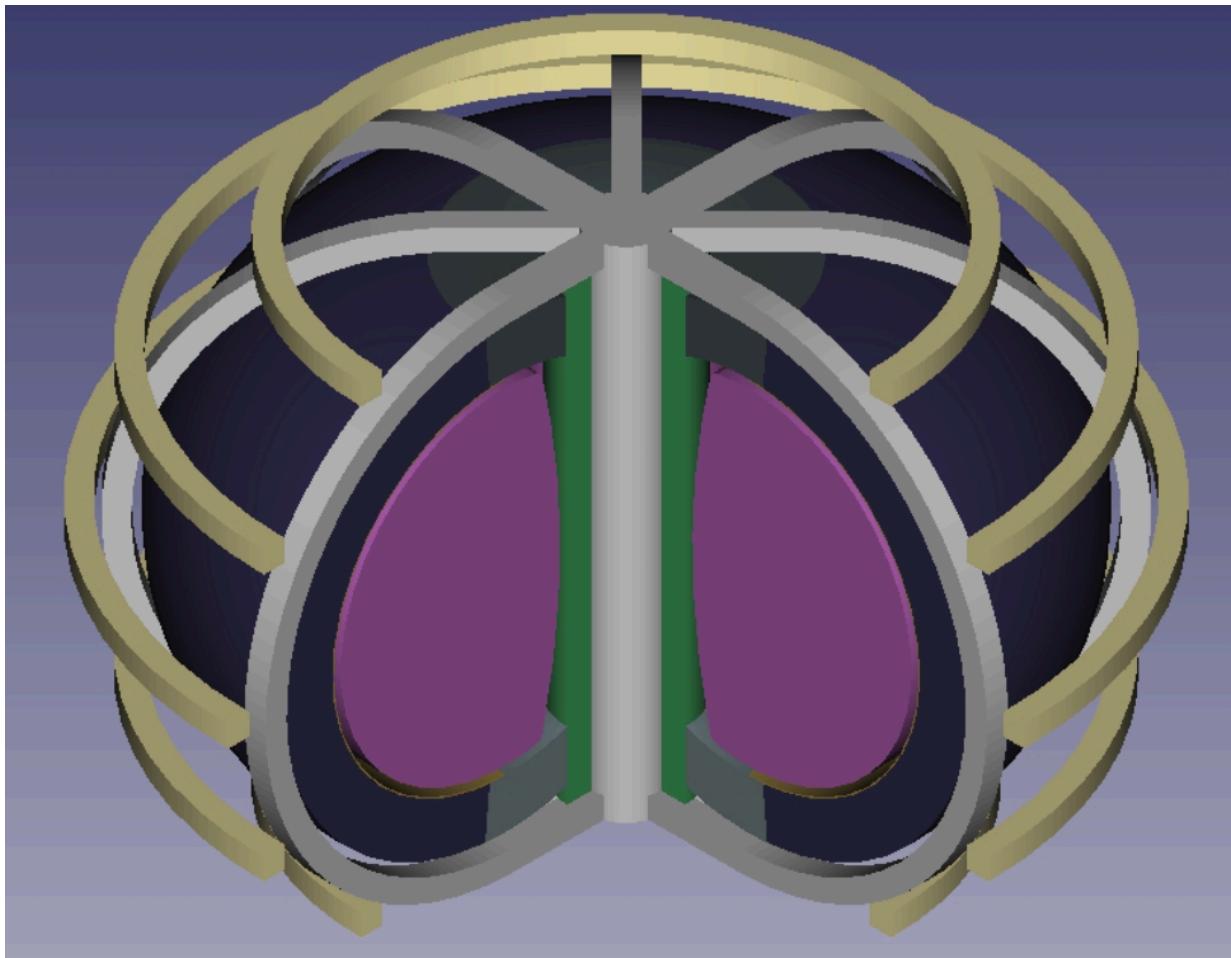


- A Monte Carlo particle transport code focused on neutron criticality simulations, recently developed in the Computational Reactor Physics Group at MIT [1]
- Modern C++
- Being evaluated by the UK Atomic Energy Authority (UKAEA) as a tool for simulating the ITER nuclear fusion reactor [2]
- CPUs only, using OpenMP for on-node parallelism and MPI for inter-node parallelism

[1] P. K. Romano and B. Forget, “The OpenMC Monte Carlo particle transport code,” *Annals of Nuclear Energy*, vol. 51, pp. 274–281, 2013.

[2] A. Turner, “Investigations into alternative radiation transport codes for ITER neutronics analysis,” in *Transactions of the American Nuclear Society*, 2017.

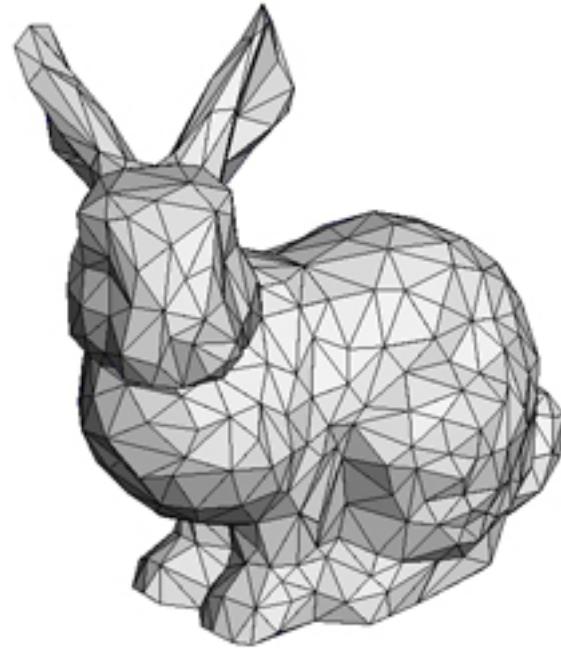
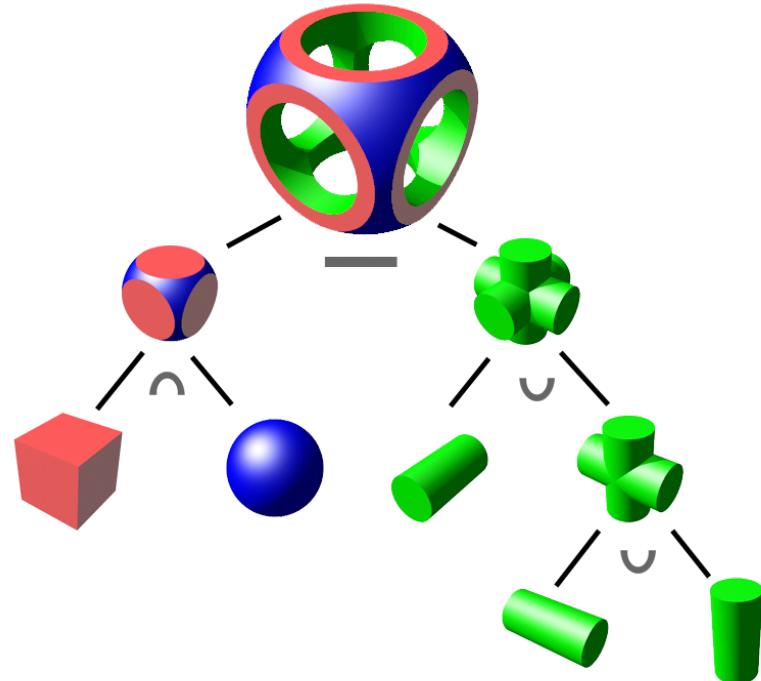
# Motivation – fusion reactor design



- Tokamak model from UKAEA
- CAD model
- $O(10^8)$  triangles in mesh
- $O(10)$  GBytes of data

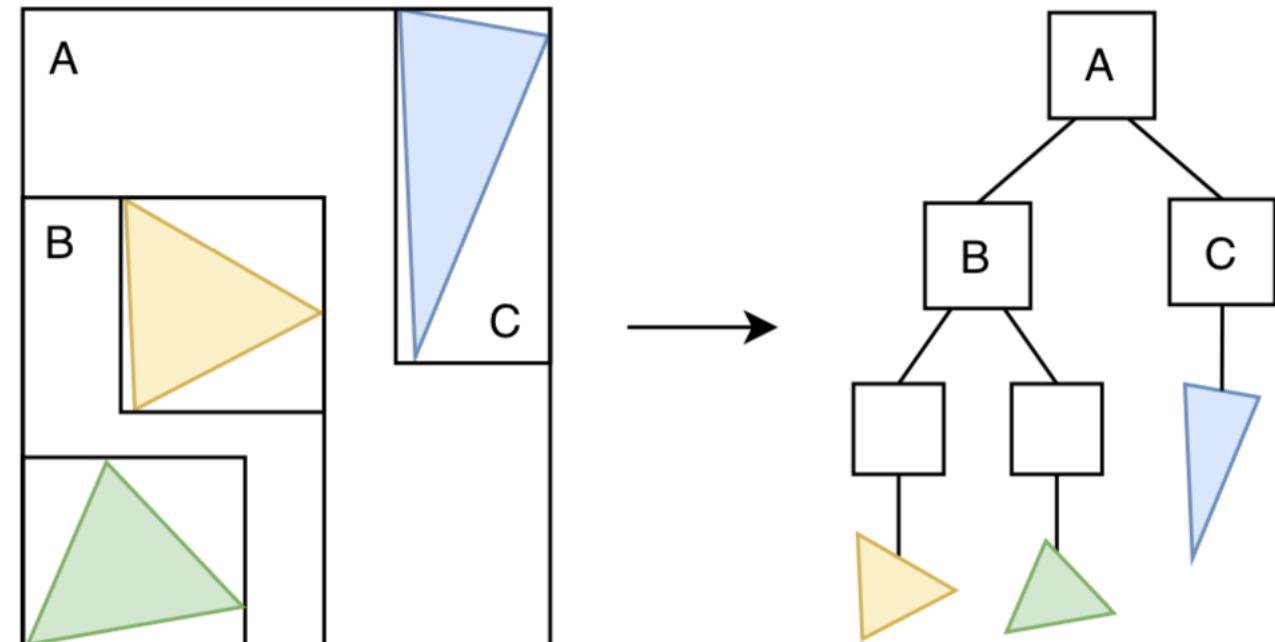
# Generating the input geometries

- Can be Constructive Solid Geometry (CSG) or meshes of many small triangles from CAD tools



# Acceleration structures

- For large models, finding the intersection points is expensive
- Acceleration structures use a hierarchy of progressively smaller bounding boxes around model sub-regions
- These boxes are then tested for intersection in a binary tree style search, massively reducing the number of surfaces that need to be tested
- E.g. Bounding Volume Hierarchy (BVH) trees, Octrees and Kd-trees



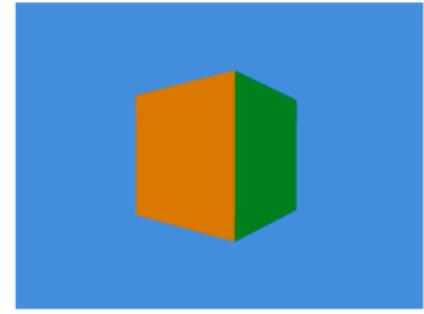
# How the RT cores work

- Each SM on the GPU has access to its own RT core to which it can issue “ray probe” requests
- Each RT core has triangle intersection and BVH traversal units
- Can cache triangle vertex and BVH tree data
- The two units in the RT core execute the ray probe asynchronously, writing the result back to an SM register once complete. The SM can perform other work in parallel.

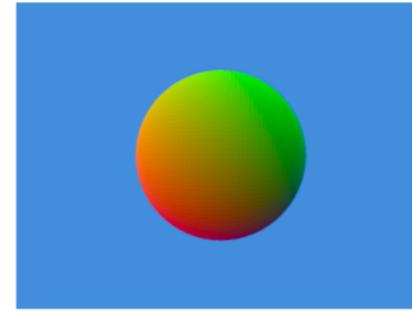
# Implementation

- It's not yet possible to program the RT cores directly via CUDA et al
- Have to use a library: **NVIDIA's OptiX™** ray tracing library, or Vulkan, Microsoft DXR, ...
- In OptiX, the user provides a set of CUDA-like kernel programs as PTX strings, each of which performs a specific function in the ray tracing pipeline
  - E.g. generating rays, handling intersections or handling rays which miss the geometry entirely
- These programs are then compiled on-the-fly by OptiX and woven into a single “mega kernel”
- OptiX then handles scheduling of kernel launches internally, automatically balancing load across the GPU

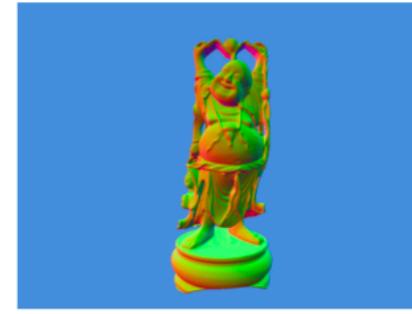
# Benchmarking RT cores for raytracing



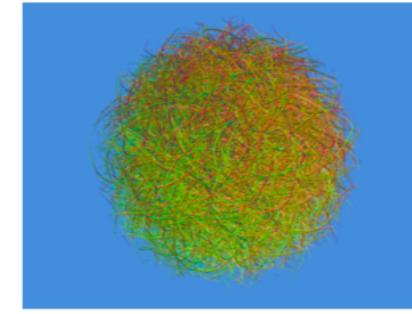
(a) Cube, 12



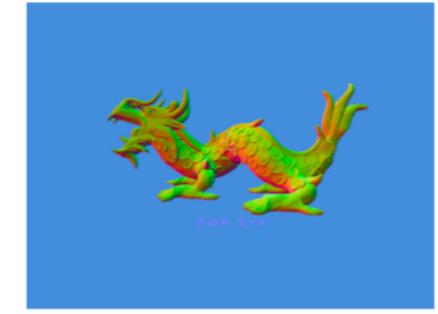
(b) Sphere, 89k



(c) Happy Buddha, 1.1m



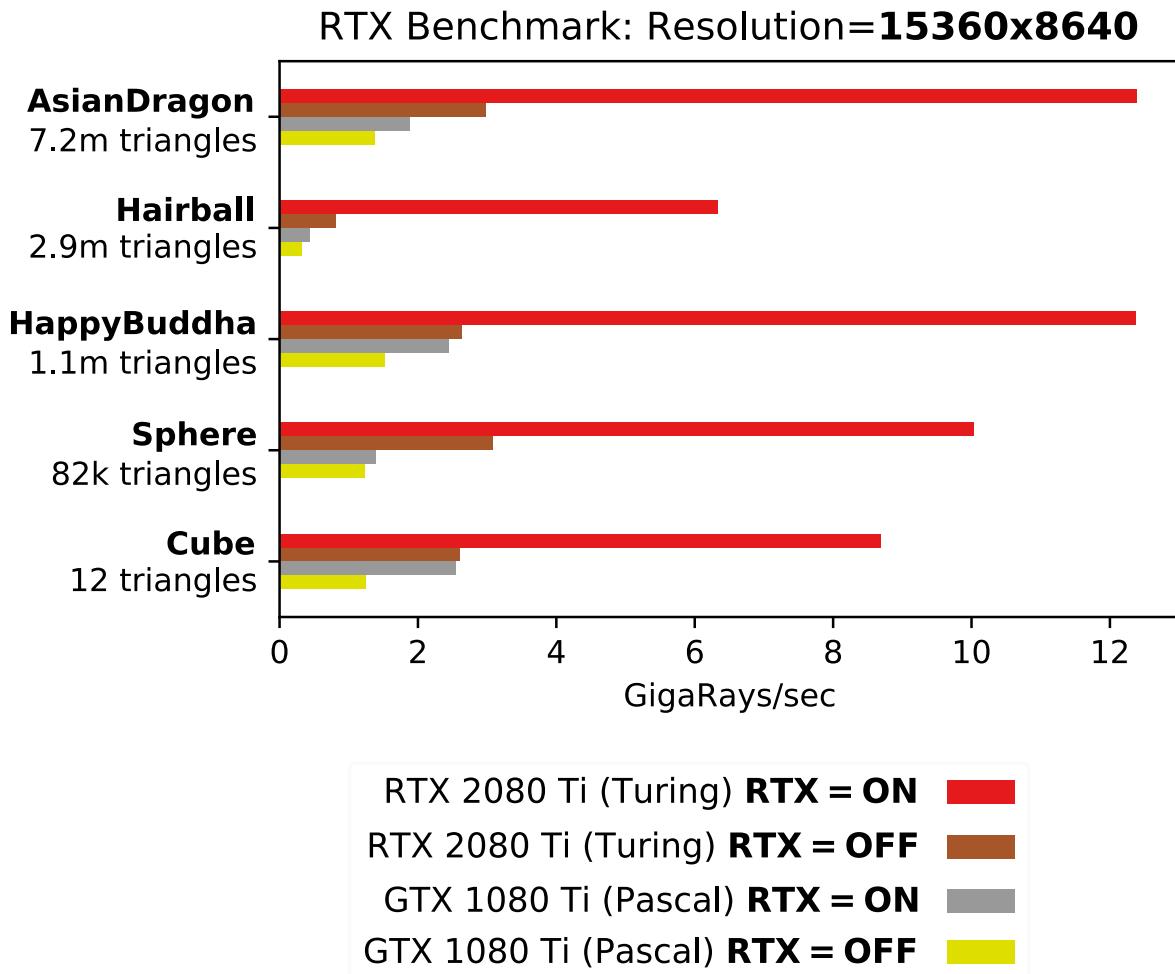
(d) Hairball, 2.9m



(e) Asian Dragon, 7.2m

- Developed a simple benchmark to evaluate the raw ray tracing performance of RT cores
- Renders frames of a 3D triangle mesh scene as fast as possible
- Each thread handles a single ray and writes the computed pixel colour to an output buffer, which is then interpreted as an image
- Five 3D models were selected to use as rendering targets, from a trivial 12 triangles, to over 7 million triangles.

# Raytracing speedup using the RT cores and OptiX



- 4.6X speedup on average for the Turing GPU
- Over 12 GigaRays/sec for the Happy Buddha and Asian Dragon models
- The Hairball model is the most geometrically complex, achieves an 11.8X speedup.

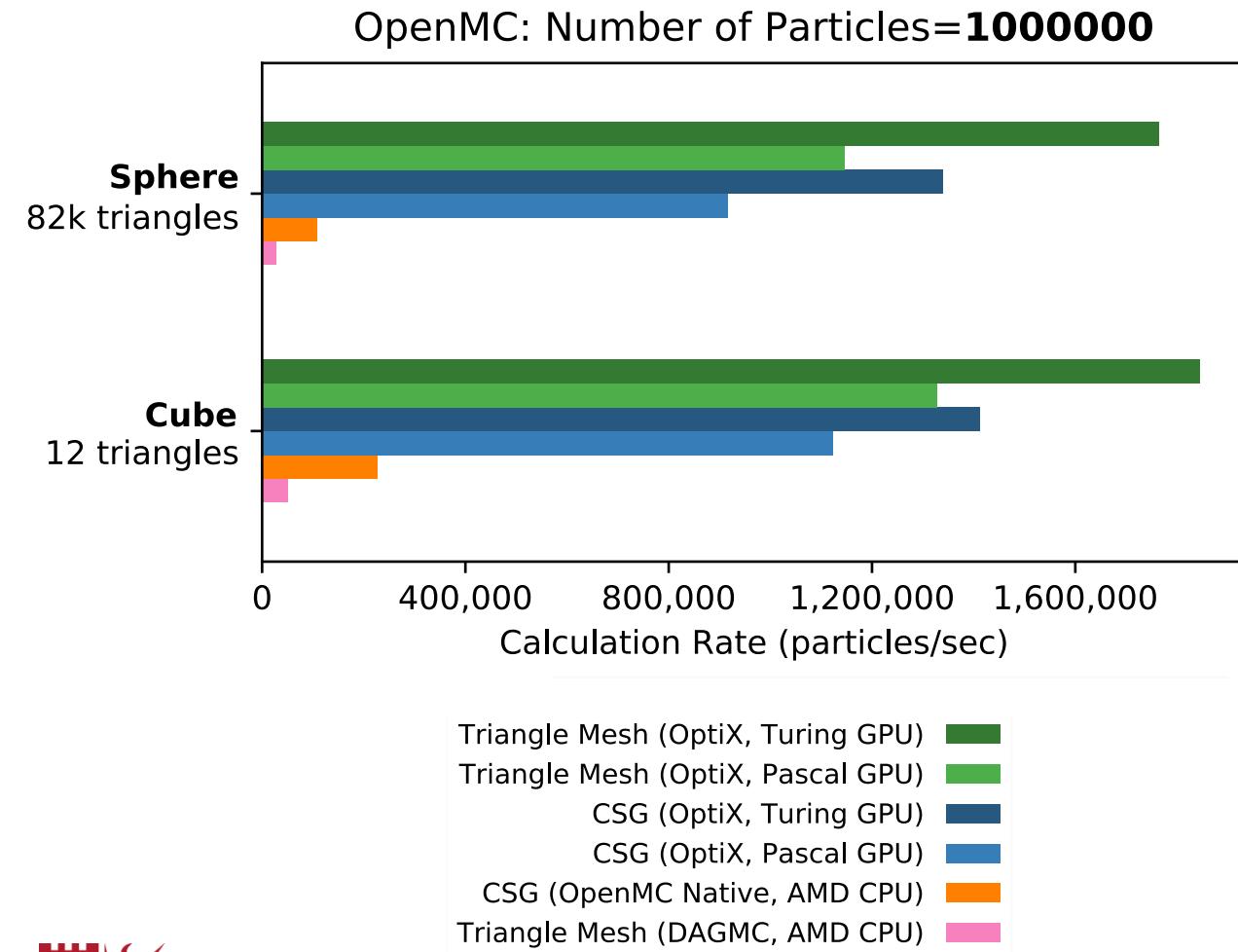
# Porting OpenMC to GPUs

- Ported the main kernel to CUDA and to OptiX for comparisons
- In the absence of a real fusion reactor model, we used the same five models from the RT benchmark
- Each model is filled with a fissionable material ( $^{235}\text{U}$ ) and is surrounded by a bounding cube filled with a vacuum
- Particles are terminated if they hit the edge of the bounding cube
- The particle source is set inside the model
- Each model was simulated for 2 generations using 2 batches of N particles, where N ranges from  $10^3$  up to  $10^7$ .

# Methodology

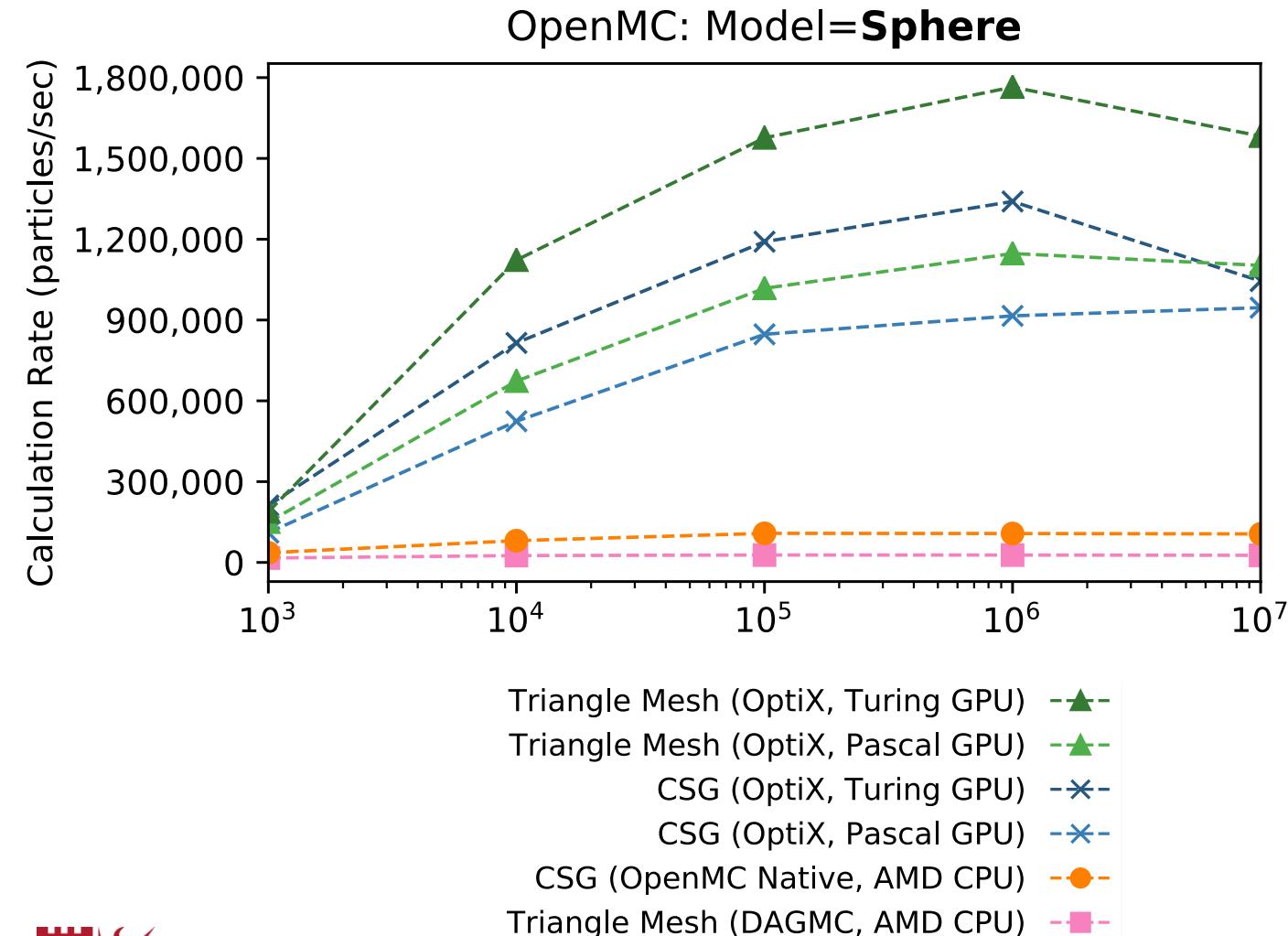
- Ran native OpenMC on a 16-core AMD Ryzen 7 2700 CPU using GCC 7.4
  - Only using the Cube and Sphere models as needed to be CSG for CPU
- Also used the same two GPUs as before: RTX 2080 Ti (Turing) and GTX 1080 Ti (Pascal)
- We collected the particle calculation rate (measured in particles/s) and wallclock time spent in particle transport.

# CUDA OpenMC results on simple geometries



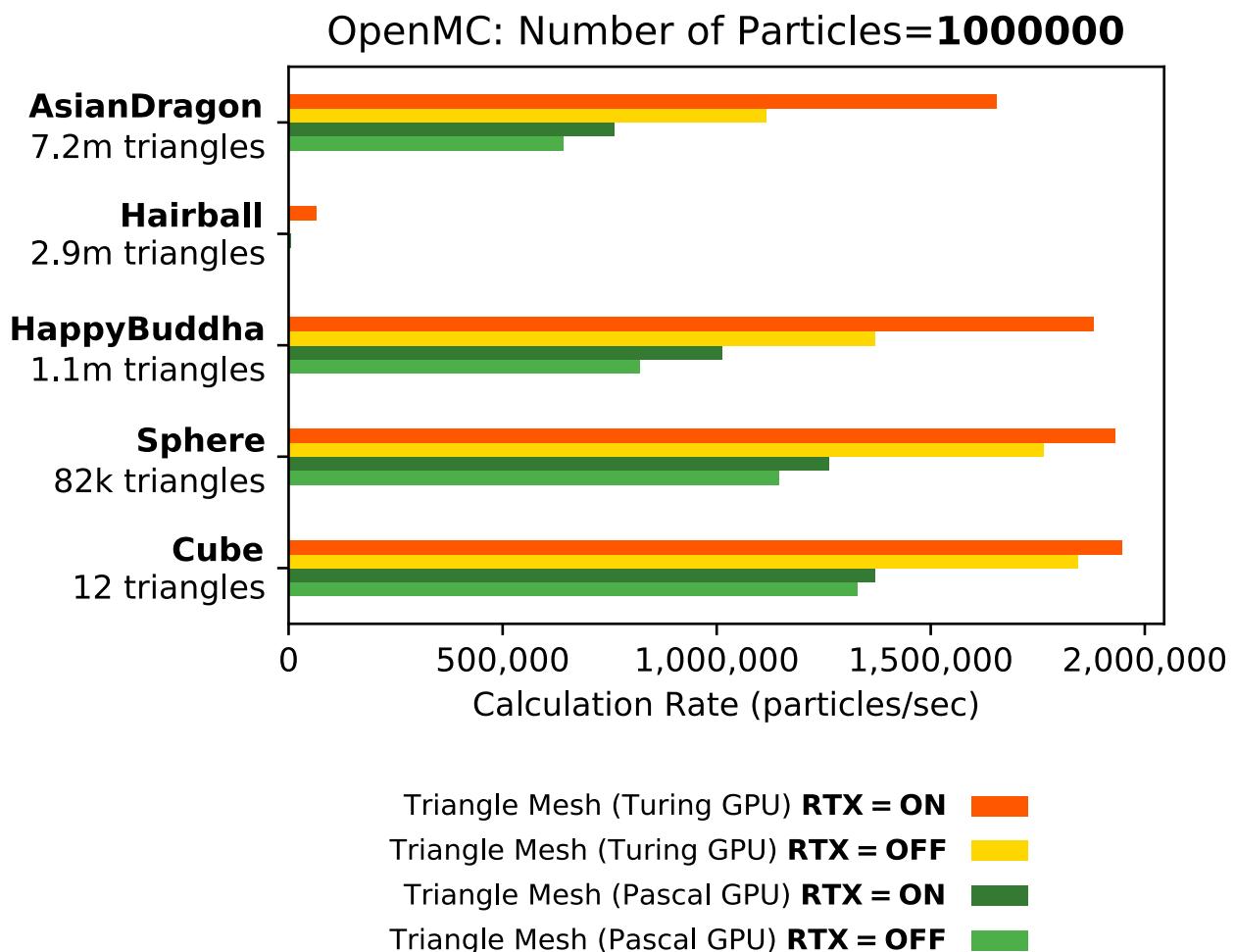
- GPU version ~13X the native CPU version
- Turing ~1.4X faster than Pascal

# Increasing particle counts



- Peak speedup is 16.4X over the CPU
  - Peak GPU performance at 1M particles

# RT (OptiX) OpenMC results



- RTX mode on the Turing GPU is the fastest in all cases, being 30-50% faster on the larger geometries
- The Hairball model shows the biggest difference, being 20.1x faster with RTX mode on Turing.

# Conclusions

- Monte Carlo-based particle transport can port well to GPUs
- Ray tracing hardware holds promise for accelerating this application, with results from 1.3X to 20X over CUDA alone
- Currently hard to program these cores – have to go through a graphics API to do it
- AMD, Intel, Arm and others are also adding RT hardware
- Potentially other uses of RT hardware that can be explored

# For more information

**M. Martineau and S. McIntosh-Smith. Exploring on-node parallelism with neutral, a Monte Carlo neutral particle transport mini-app.** In Cluster Computing (CLUSTER), 2017 IEEE International Conference on, 2017. DOI: 10.1109/CLUSTER.2017.83

**On the Porting and Optimisation of Physics Simulations for Heterogeneous Parallel Processors.**  
Matt Martineau, PhD thesis, University of Bristol, January 2019.

**Bristol HPC group:** <https://uob-hpc.github.io/>

**Isambard:** <http://gw4.ac.uk/isambard/>

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