

## Exploring C++ Standard Parallelism Features for GPU Programming in a Particle-In-Cell Application

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### Outline

- 1. Performance Portability
- 2. C++ Standard Parallelism
- 3. MiniPIC
- 4. Methodology
- 5. Benchmarking Results
- 6. Conclusion
  Strength
  Limitations
- 7. GPU Atomicity: Early Insights
- 8. Looking Ahead



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## 1. Performance Portability



### pre-exascale systems 2016 2018

2020

2022

exascale systems

2024

LANL/SNL Trinity



Cray(Intel)/Intel KNL

LLNL Sierra



IBM power9/NVIDIA Volta

**ORNL Summit** 



Riken Fugaku



ARM CPUs with SVE

**ORNL Frontier** 



AMD CPU AMD GPU

2023

ANL Aurora



Xeon CPU/Intel GPU

LLNL El Capitan



AMD CPU/ AMD GPU

Jupiter



Rheal (ARM) /NVIDIA



## 1. Performance Portability



#### **Constructors**

#### Low-Level Language

- Advanced optimizations with fine-grained control
- Very high performance
- Limited portability

#### **Examples:**

- •CUDA (for NVIDIA GPUs)
- •HIP (for AMD GPUs)

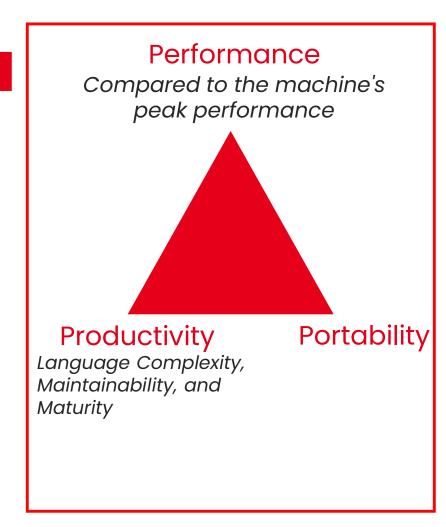
#### Portability in a Scientific Context

#### **High-Level Language**

- Good performance
- Less manual optimization
- Better portability
- Increased productivity

#### **Examples:**

- Directive-based models: OpenMP, OpenACC
- C++-based libraries: Thrust, Kokkos
- Evolution of C++

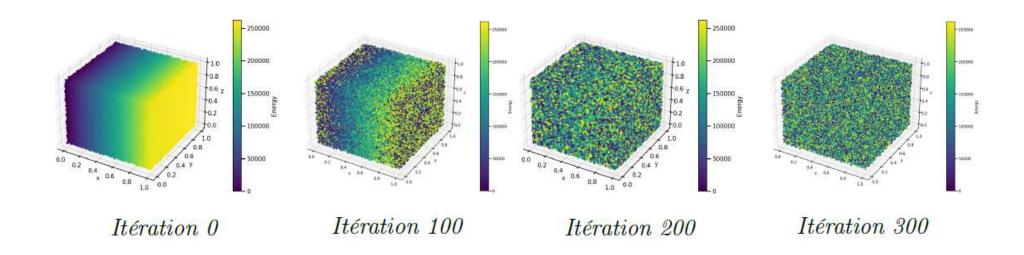




## 1. Performance Portability



Does the C++17 Standard Parallelism model offer a good compromise between Performance, Portability, and Productivity especially when running large-scale plasma physics simulations on a single GPU?





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### 2. C++ Standard Parallelism

#### General Overview

No language extensions

Example: CUDA: \_\_host\_\_, \_\_device\_\_, \_\_global\_\_

No special directives

Example: OpenACC: #pragma

No new libraries

Example: Thrust

- Uses familiar C++ objects
- Same abstraction model for memory

Example: std::vector

Implicit memory management





### 2. C++ Standard Parallelism

### **Execution Policies**

Serial (C++98)	Parallel (C++17)		
<pre>std::vector<t> x{}; std::transform(x.begin(), x.end(), x.begin(), [](int x) { return x + 1; });</t></pre>	<pre>std::vector<t> x{}; std::transform(std::execution::par_unseq, x.begin(), x.end(), x.begin(), [](int x) { return x + 1; });</t></pre>		

#### C++ defines four execution policies:

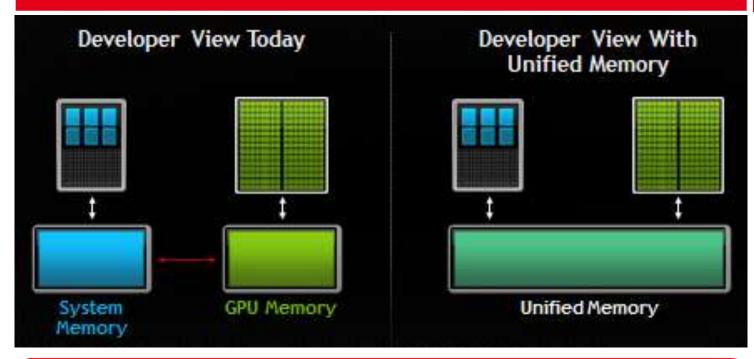
- **std::execution::seq:** Sequential execution. No parallelism is allowed.
- **std::execution::unseq:** Vectorized execution on the calling thread, but not parallel
- **std::execution::par:** Parallel execution on one or more threads
- std::execution::par\_unseq: Parallel execution on one or more threads, with each thread potentially vectorized



# 2. C++ Standard Parallelism NVIDIA Stdpar implementation



### **Memory management**



#### nvc++ compiler flags

#### **CPU**

nvc++ -stdpar=multicore

#### **GPU**

nvc++ -stdpar=gpu nvc++ -stdpar

Still physically separate memory spaces, BUT a unified virtual address space

#### Clang compiler flags

clang++ --hipstdpar



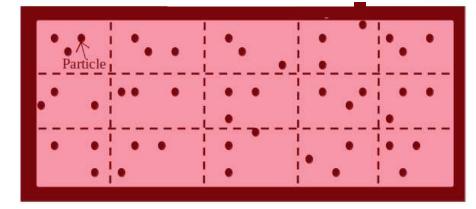


# 3. MiniPIC Description

- MiniPIC: A Mini Application
- Based on Smilei, a simulation code for plasma physics
- Uses the **Particle-in-Cell (PIC)** model
- Goal: Explore new programming models
- Several Programming Models Implemented: SYCL, Thrust, Kokkos, OpenMP, OpenMP Target, OpenACC
- A Domain Consists of Two Data Structures:
  - > Particles: Represent matter
  - Grid: Represents electromagnetic fields
- Interaction between particles and the grid

- C++ Code
- No distributed-memory parallelism

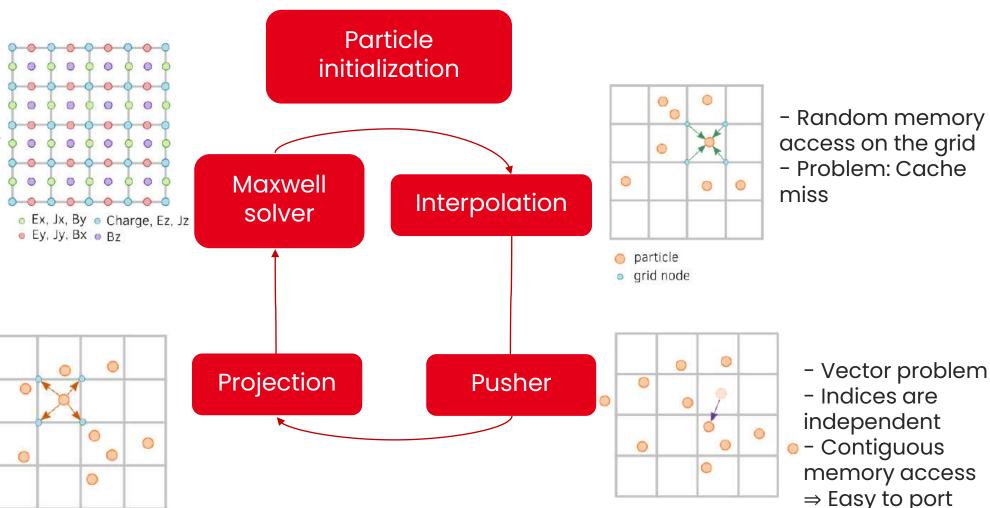
#### <u>GPU</u> Domain





### 3. MiniPIC PIC loop: The four main operators

- Stencil problem
- Less computationally expensive
- ⇒ Parallelizes efficiently on GPU



Random access

Memory contention



## 4. Methodology

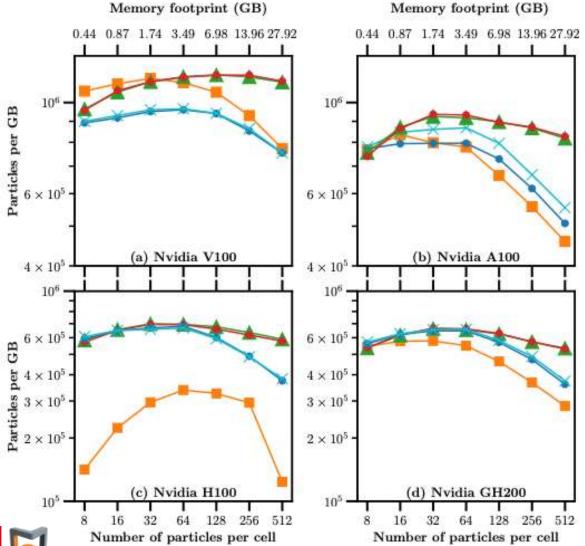
- The case studied: Thermal plasma
- It consists of electrons and ions with a uniform charge distribution
- 64 cells per direction

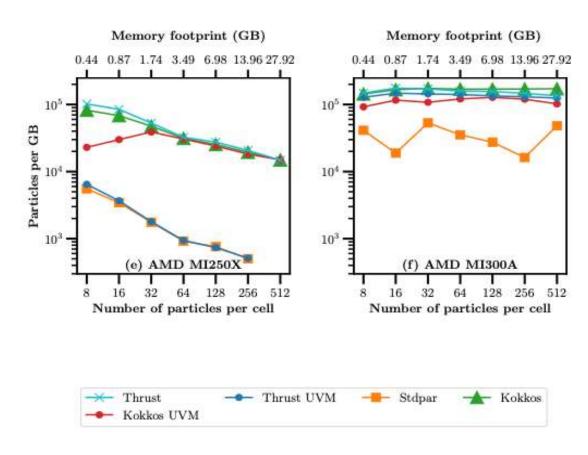
- Comparison with:
  - Thrust
  - Kokkos
  - Thrust with UVM
  - Kokkos with UVM

GPU Name	CUDA/ ROCM Version	Compiler Version	Memory Size/Type I	$egin{array}{l} Memory \ Bandwidth \ (MB/s) \end{array}$	Double- Precision (TFLOPS)
NVIDIA V100 (PCIe 32GB)	12.2.1	nvc++ 23.7	32GB/HMB2	900	7
NVIDIA A100 (SXM 40GB)	12.2.1	nvc++ 23.7	40GB/HBM2	1555	9.7
NVIDIA H100 80GB	12.2.1	nvc++ 23.7	80GB/HBM3	3350	34
NVIDIA GH20 480GB	0 12.3	nvc++ 24.1	96GB/HBM3	4000	34
AMD MI250X	rocm-6.1.2	$^{\mathrm{clang}++}_{17.0.0}$	128GB/HBM2	2e 3200	47.9
AMD MI300A	rocm-6.1.2	$_{ m clang++} $ $_{ m 17.0.0}$	128GB/HBM3	3 5300	61.3



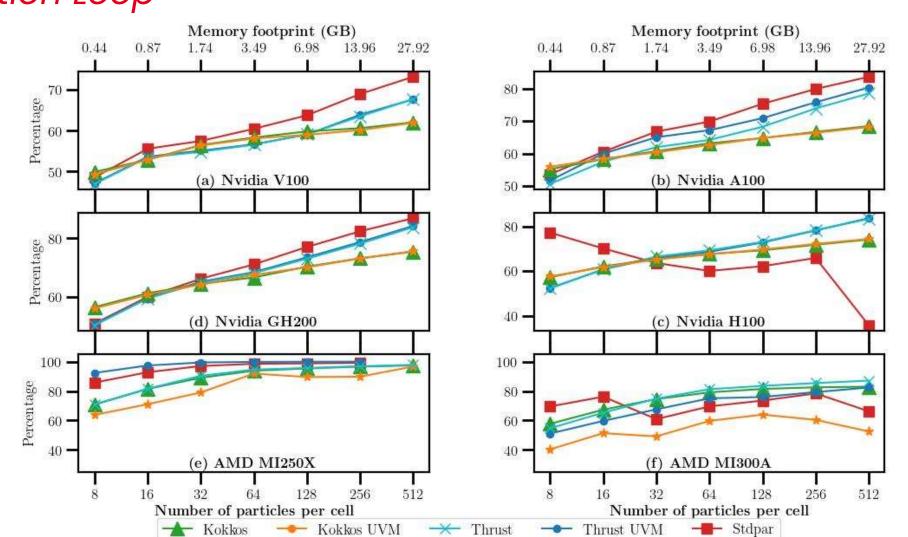
# 5. Benchmarking Results PIC Loop







# 5. Benchmarking Results *Projection Loop*







# 6. Conclusion Strengths



- Accessible to non-experts
- No major rewrite of existing C++ code
- No explicit memory management
- Compatible with NVIDIA and AMD
- Classic dynamic memory allocation
- High performance for simple loops





## 6. Conclusion *Limitations*

- Limited optimization for experts
- CPU-GPU switching without explicit management → inefficient data transfers
- Execution policy implementation depends on compiler developers
- Model still in development
- Potential for further improvement



## 7. Looking Ahead



#### **Untapped Potential**

- Investigate the performance differences between different programming models, particularly in comparison with Thrust.
- •Analyze performance variations between the H100 and GH200.

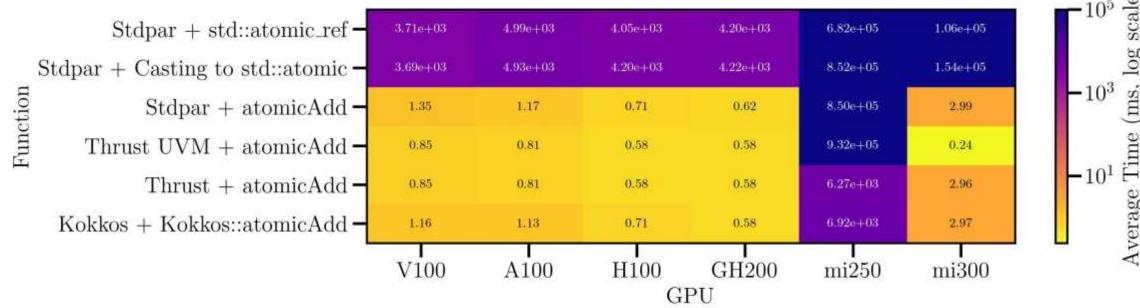
#### **Future Direction**

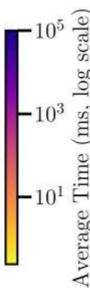
•Focus on the exploration of asynchronous programming.



## 8. GPU Atomicity: Early Insights

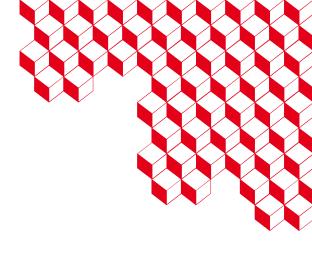












## Thank you