



山东大学
SHANDONG UNIVERSITY



THE OHIO STATE UNIVERSITY



A Case Study for Ray Tracing Cores: Performance Insights with Breadth-First Search and Triangle Counting in Graphs

Zhixiong Xiao*, Mengbai Xiao*, Yuan Yuan*, Dongxiao Yu*, Rubao Lee†, Xiaodong Zhang‡

*: Shandong University

†: Freelance

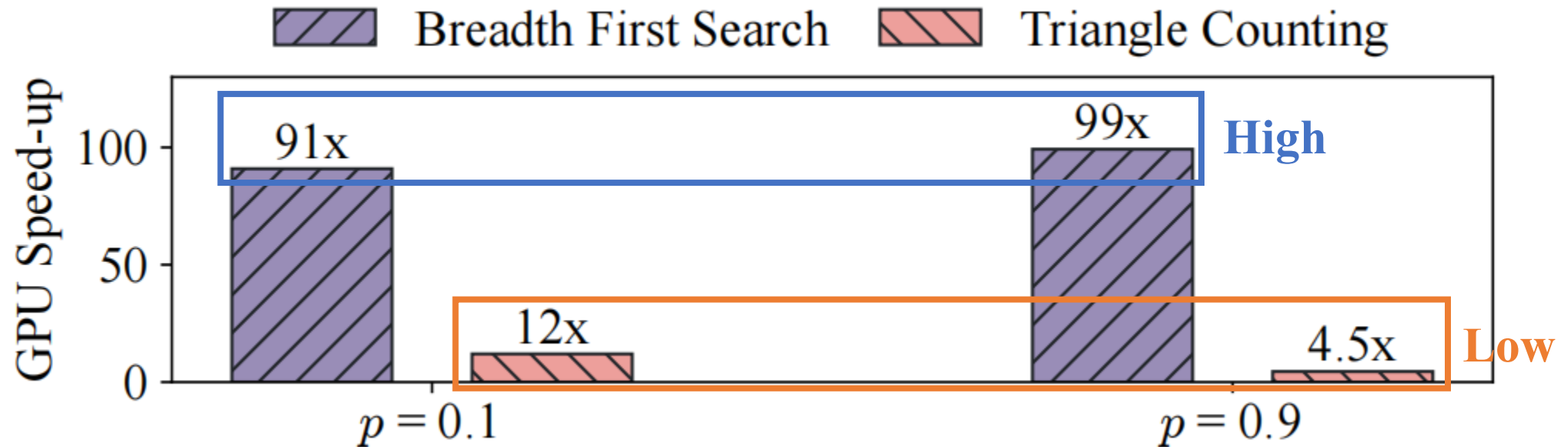
‡: The Ohio State University

June 12, 2025

Stony Brook, New York, USA

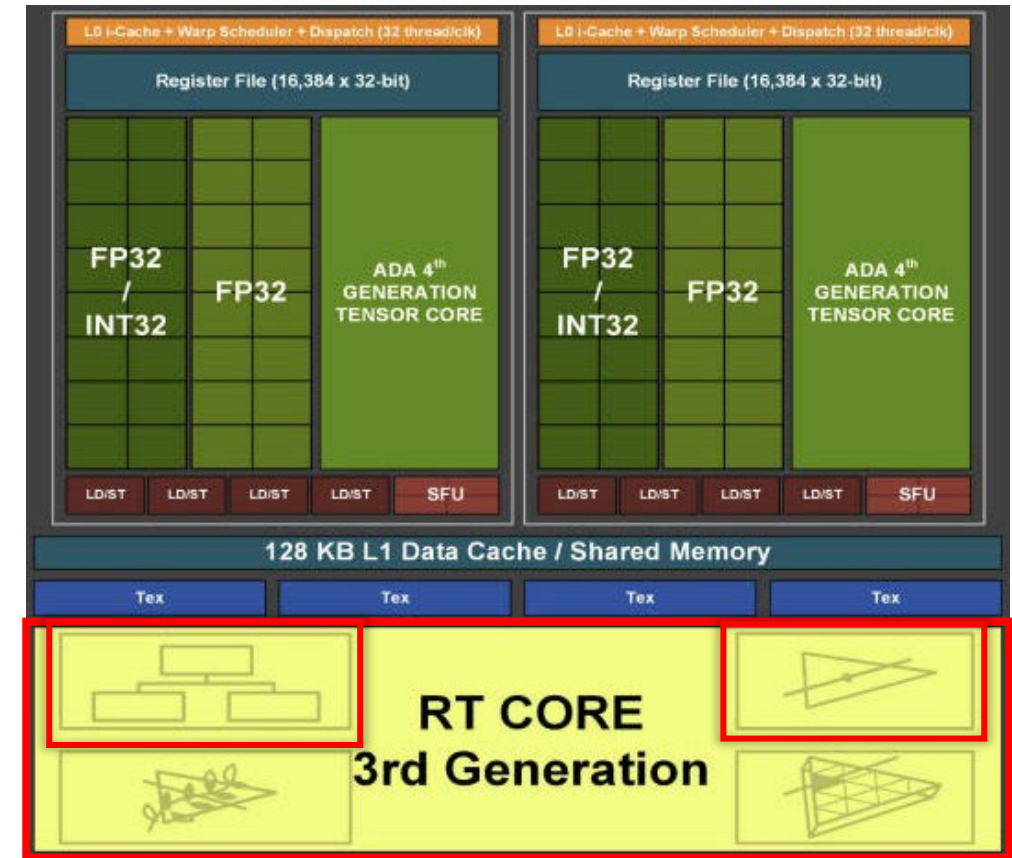
GPU Performance Gains Vary across Graph Algorithms

- **Data-parallel** graph algorithms (e.g., BFS) fit SIMT execution model well
- **Control flow-intensive** graph algorithms (e.g., TC) hardly obtain performance gains



Introduction of Ray Tracing (RT) Cores on GPU

- A **specialized hardware** for ray tracing pipeline
- Hardware-accelerated **bounding volume hierarchy (BVH)** traversal
- Hardware-accelerated **ray-triangle intersection**
- Potential for control flow-intensive workloads:
 - RTNN
 - RTScan

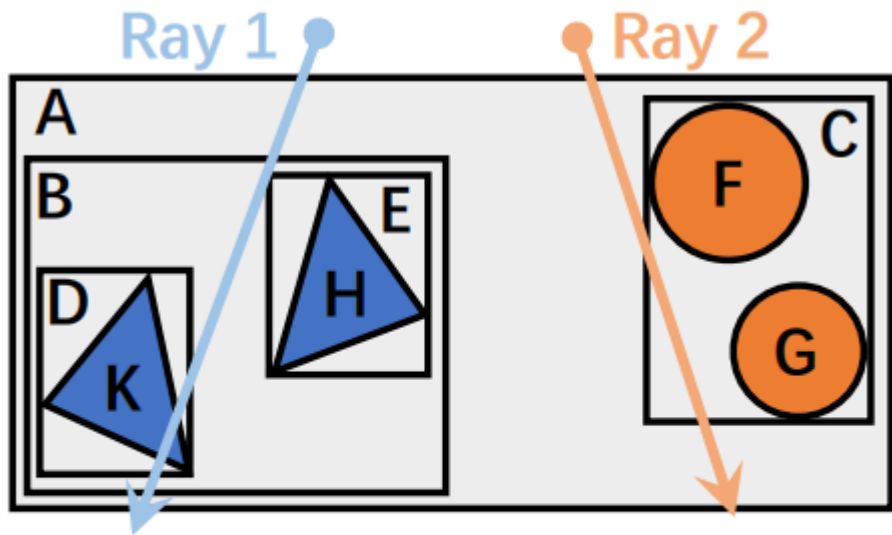


RTNN: Accelerating Neighbor Search Using Hardware Ray Tracing, PPOPP'22

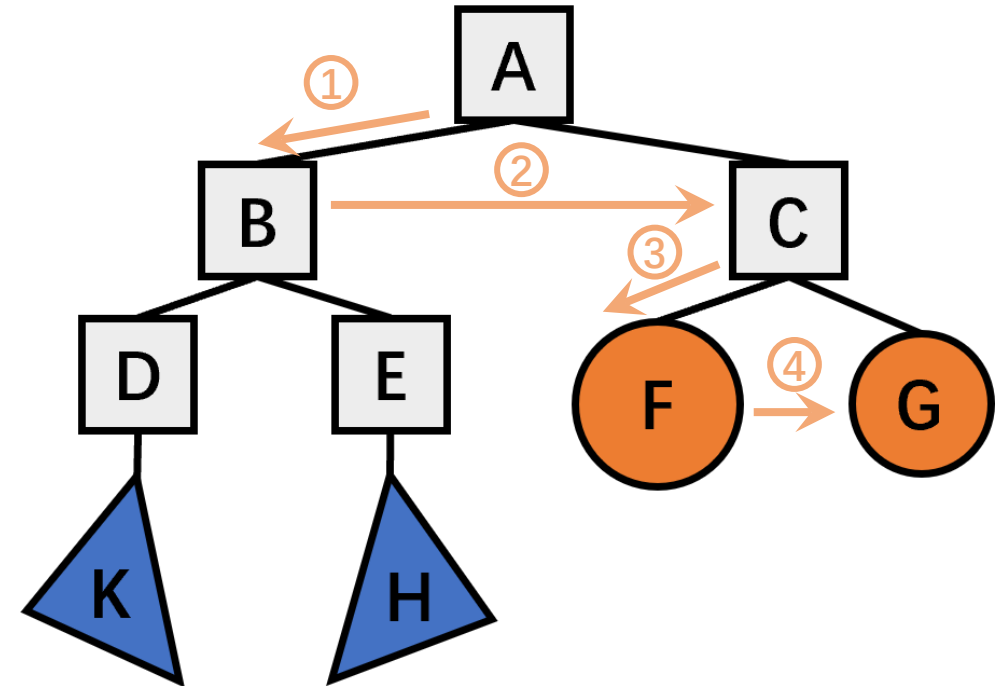
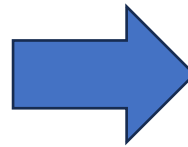
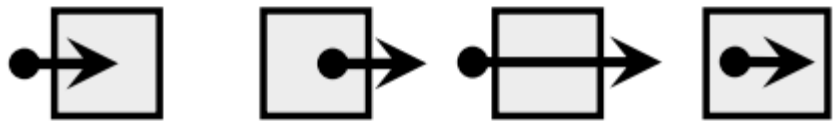
RTScan: Efficient Scan with Ray Tracing Cores, VLDB'24

BVH Construction and Traversal

- BVH is a **tree index** where primitives are recursively wrapped by axis-aligned bounding boxes (AABBs).

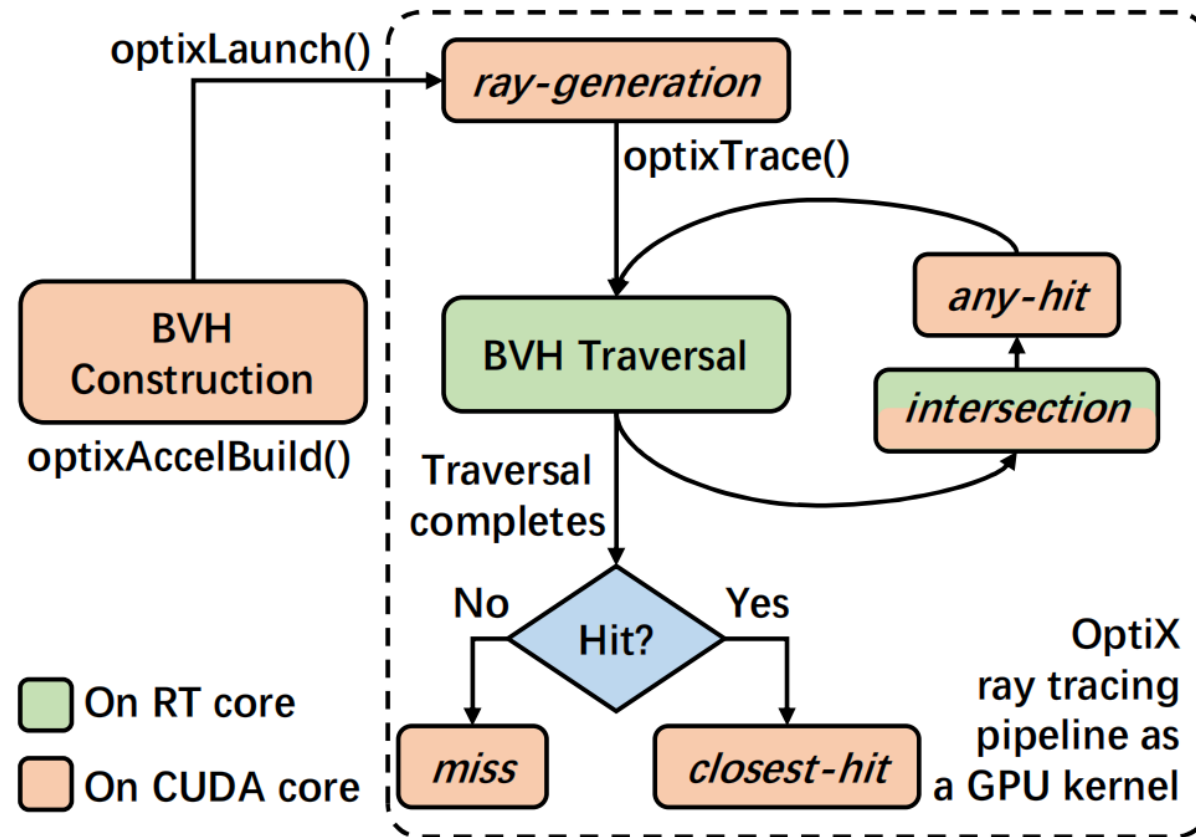


Intersection cases:

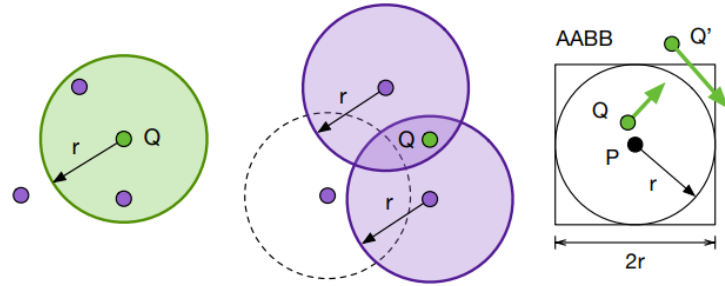


Ray Tracing Pipeline

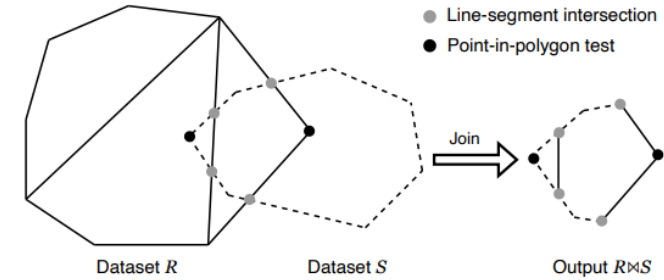
- Utilizing RT cores with Nvidia OptiX APIs.



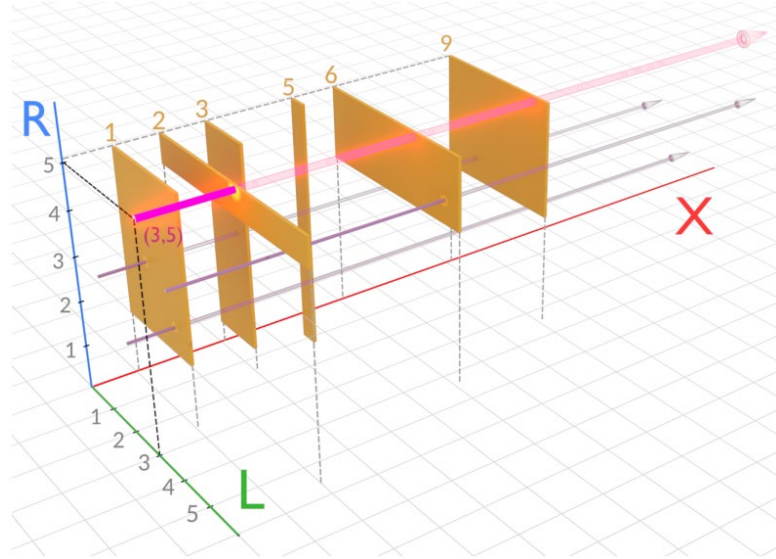
General Computing on RT Cores



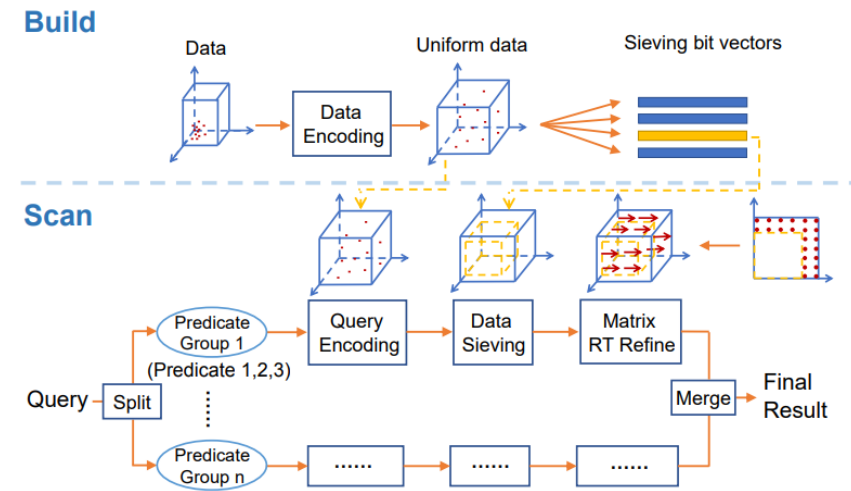
Accelerating Low-Dimensional Neighbor Search



Accelerating Spatial Join



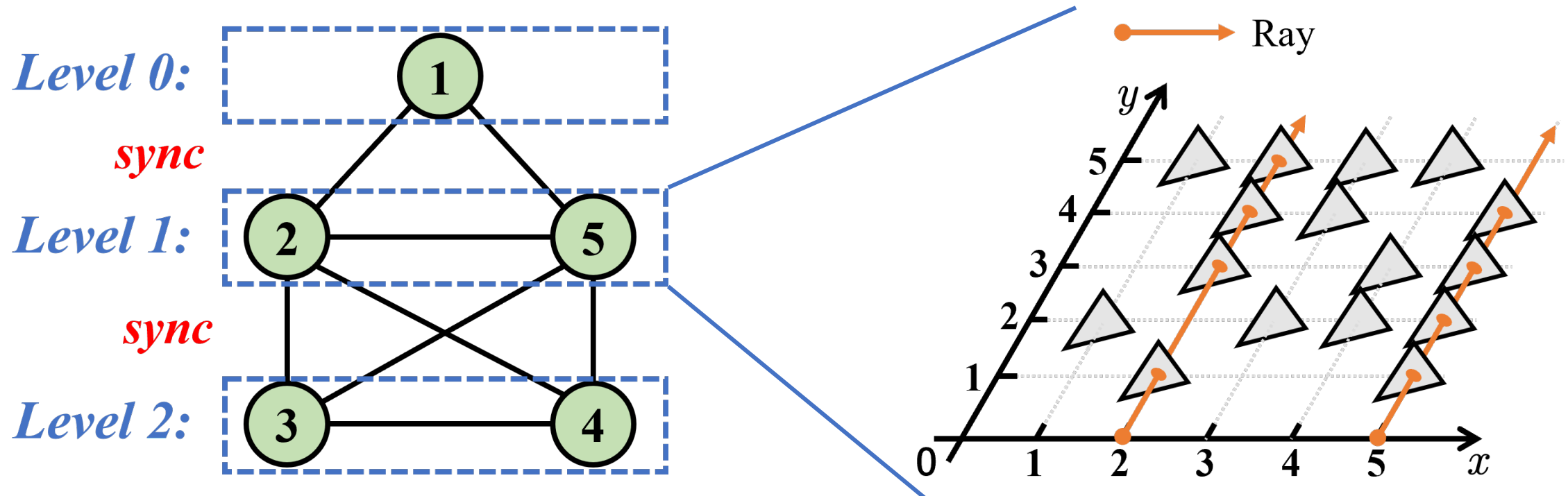
Accelerating Range Minimum Queries



Accelerating Scans in Databases

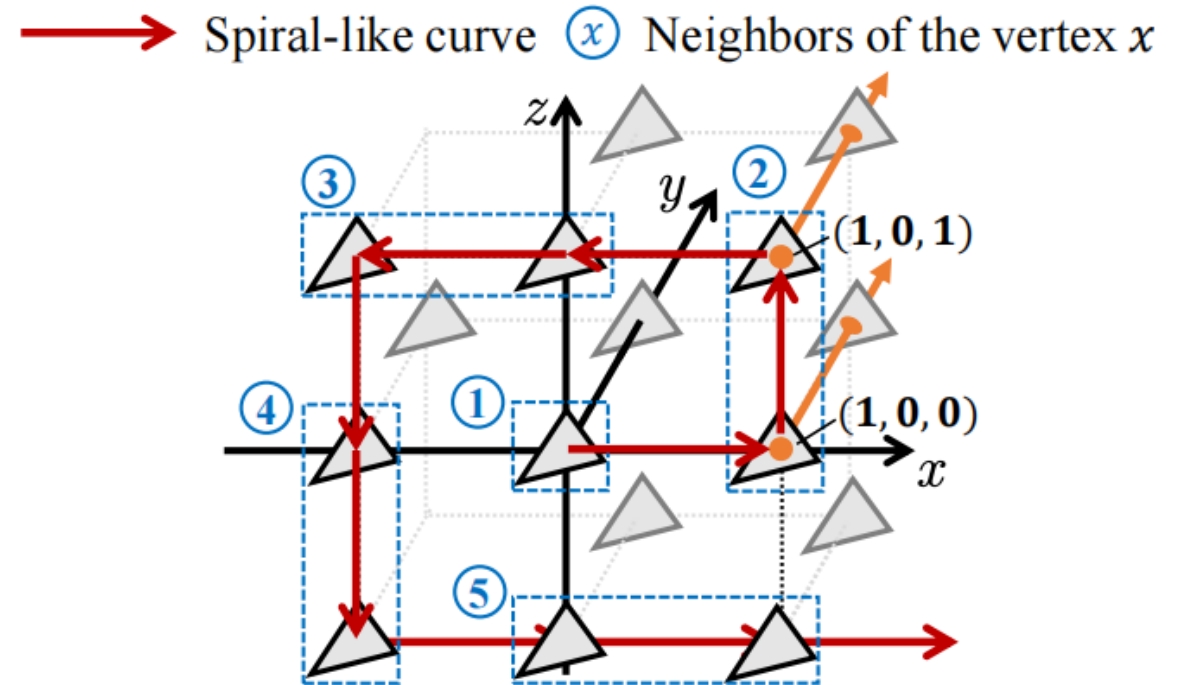
RT-based Breadth-First Search

- Adopting a parallel strategy based on [hop-by-hop](#) traversal
- Arranging primitives in a form of [adjacency matrix](#)
- Replacing [expansion](#) on CUDA cores with hardware-accelerated BVH



Further Optimizations

- **Deficiencies** of the naive method
 - Inaccuracy
 - High BVH traversal cost
 - Workload imbalance
- **Optimizations:**
 - Spiral-like curve layout of primitive queues
 - Compacting placement of primitives
 - Splitting long rays



Vertex Encoding

- Representing a neighbor with a triangle primitive is **memory-inefficient**



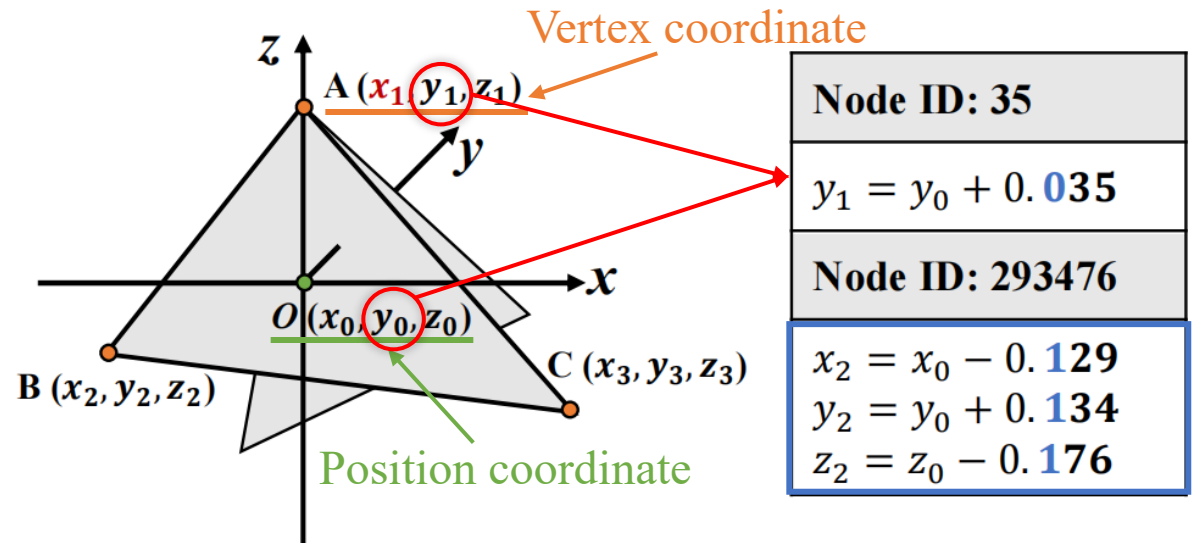
- Encoding IDs into triangle coordinates
 - Difference** between vertex coordinates and position coordinates
 - Splitting a large ID into **multiple parts**



- Reducing memory footprint and I/O

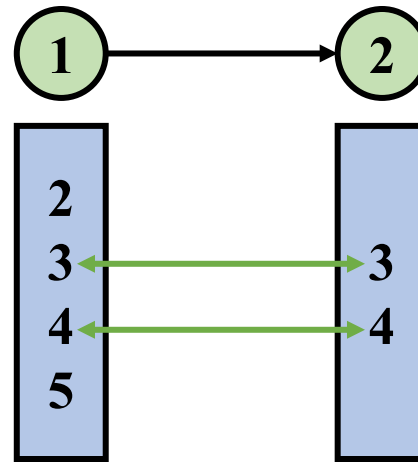
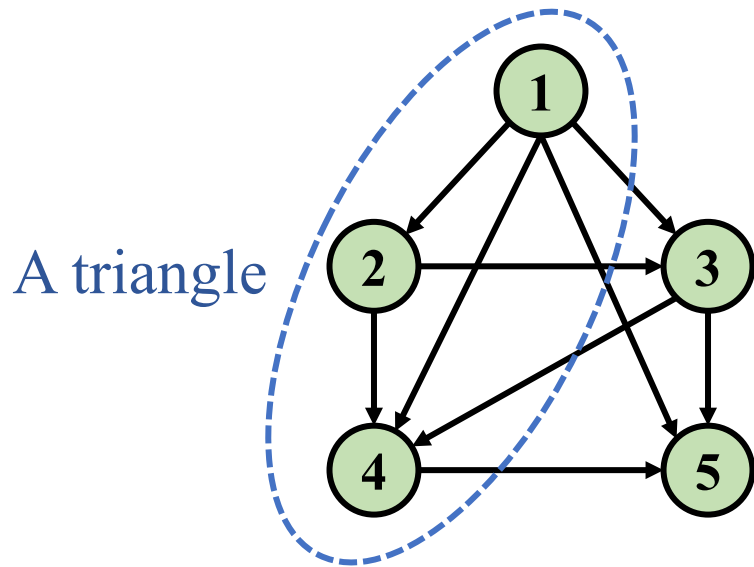
Positive encoding coordinates: x_3, y_1, y_2, y_3, z_1

Negative encoding coordinates: x_2, z_2, z_3

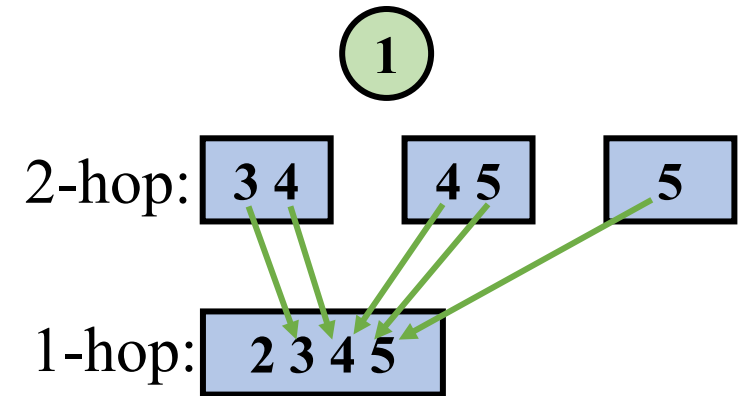


RT-based Triangle Counting

- Triangle in graph is a **clique of three vertices**
- Common strategies:
 - Edge-wise set intersection
 - Vertex-wise search



Edge-wise set intersection

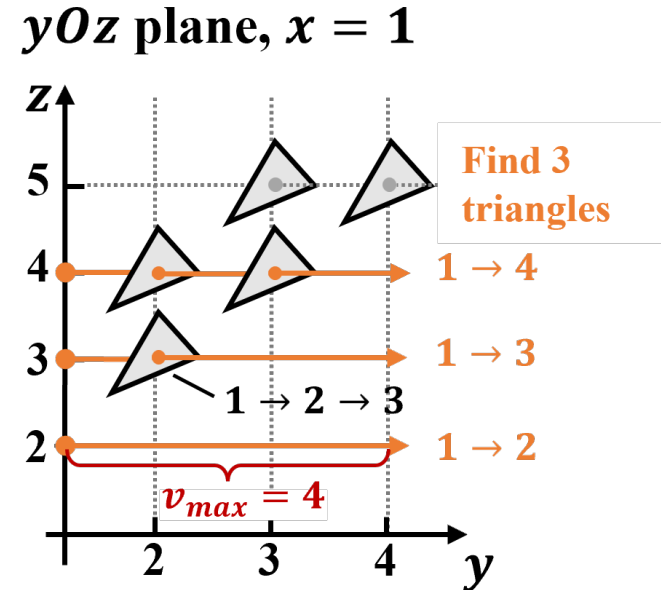
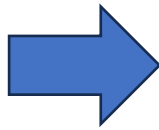
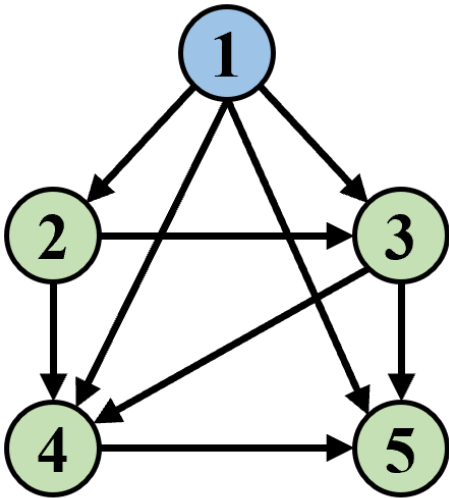


Vertex-wise search

The 1-among-2 Method

- Finding 1-hop neighbors among 2-hop neighbors
 - **Rays** represent **1-hop** neighbors, while **primitives** represent **2-hop** neighbors
- The number of rays: $|E|$, the number of primitives: $\sum_{u \in V} \sum_{v \in N(u)} \text{degree}(v)$

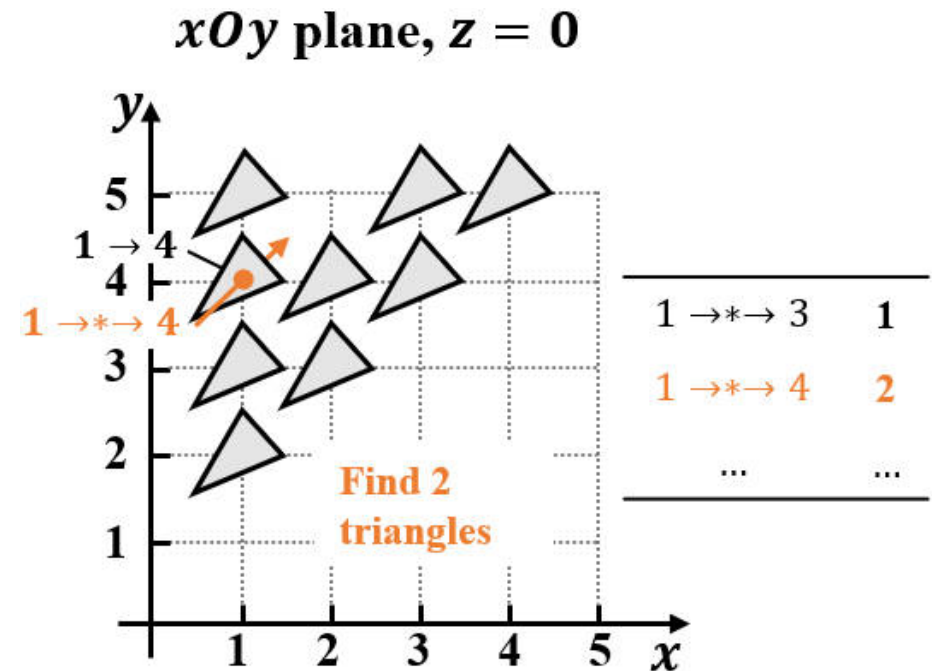
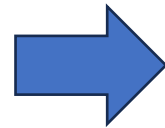
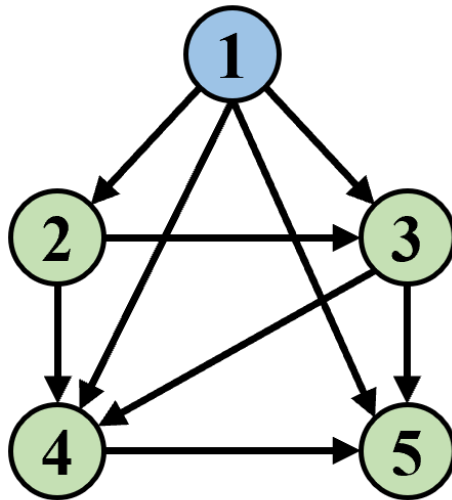
Take vertex 1 as an example:



The 2-among-1 Method

- Checking if an edge exists between a vertex and its 2-hop neighbor
 - **Rays** represent **2-hop** relationships, while **primitives** represent **1-hop** relationships
- The number of rays: $\sum_{u \in V} \sum_{v \in N(u)} \text{degree}(v)$, the number of primitives: $|E|$

Take vertex 1 as an example:

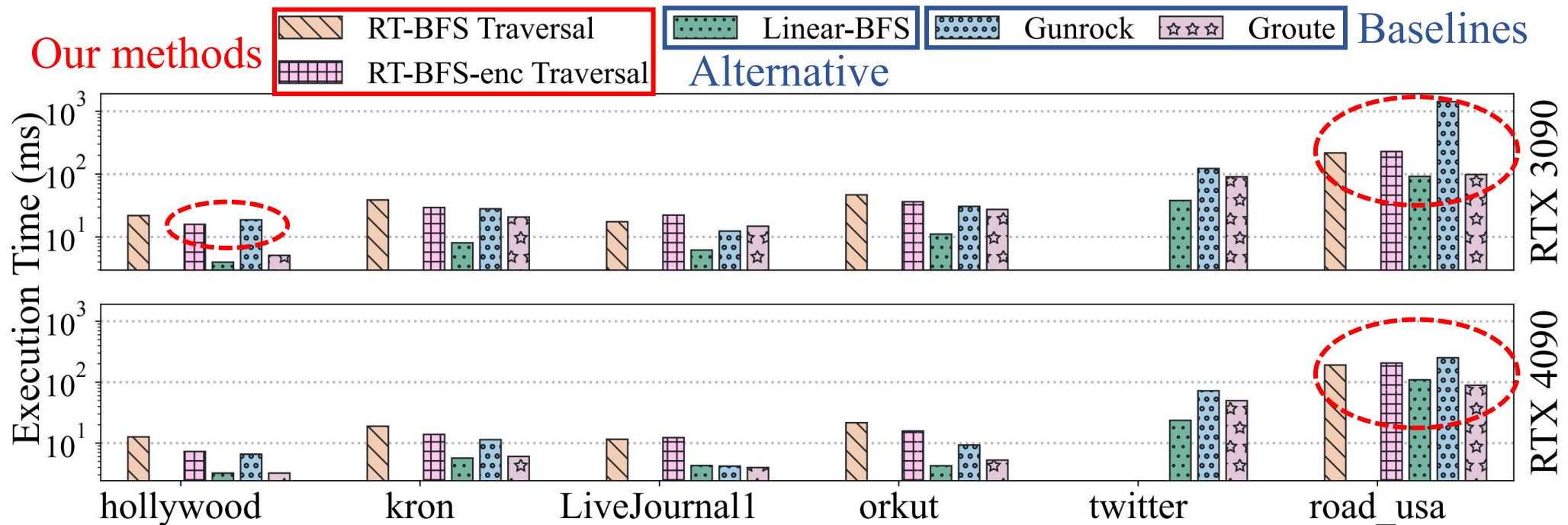


Experimental Setup

- We conduct experiments on two machines:
 - RTX 3090: 82 SMs, 10,496 CUDA cores, 24 GB memory, and 82 the 2nd-generation RT cores
 - RTX 4090: 128 SMs, 16,384 CUDA cores, 24 GB memory, and 128 the 3rd-generation RT cores
- Datasets consist of real-world graphs
- We use execution time as the primary performance metric and additionally report executed instructions, memory footprint, and power

Inferior Performance of RT-based BFS Against Baselines

- RT cores can **hardly accelerate** BFS even with encoding technique
- **3.94× slower** than Linear-BFS, a CUDA variant of the same workflow without RT cores



Understanding the Inferior Performance of RT-based BFS

- RT-BFS issues $4.03\times$ more memory instructions
- BVH traversal incurs **higher memory access** overhead than classic CUDA methods in **sequential access workloads** such as BFS

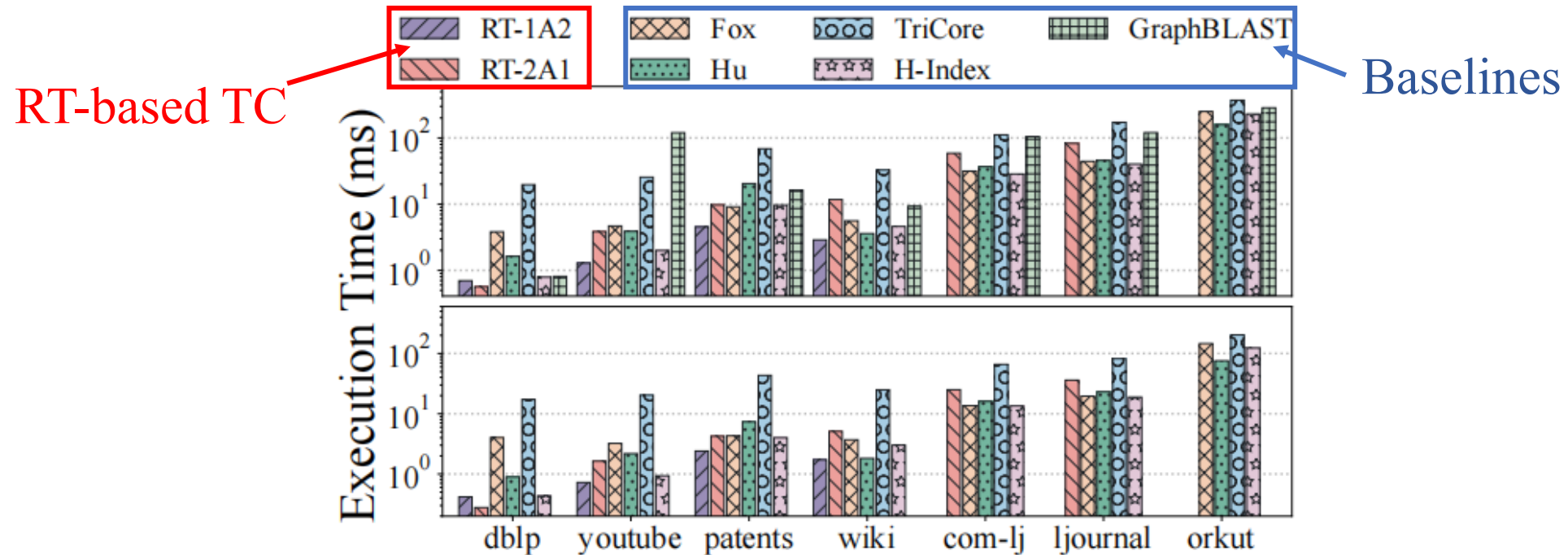
Dataset	Load/Store Instructions		
	RT-BFS	RT-BFS-enc	Linear-BFS
hollywood	54.14M	20.55M	12.55M
kron	102.15M	39.84M	23.67M
LiveJournal1	46.03M	22.39M	11.71M
orkut	123.07M	48.05M	24.93M
twitter	N/A	N/A	77.88M
road_usa	98.21M	62.49M	36.87M

$4.03\times$ memory
I/O instructions

less memory
I/O instructions

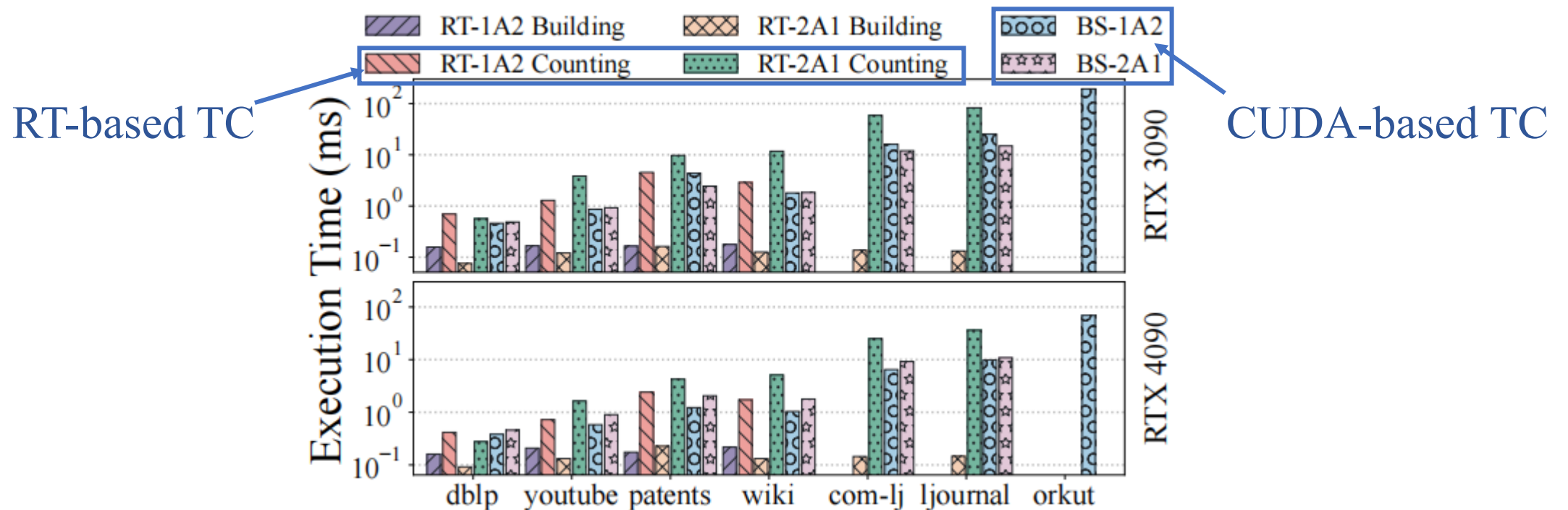
Improved Results of RT-based TC in Specific Scenarios

- RT-based TC has superior performance on some datasets
- RT-2A1 is slower due to **high miss ratio of rays**



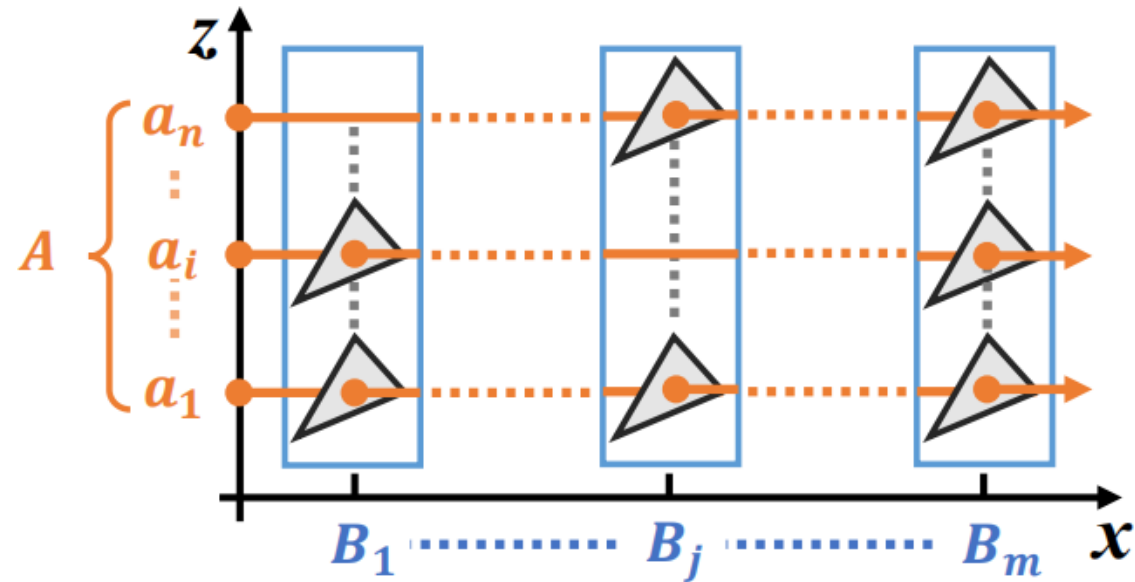
Understanding the Source of Performance Gains

- Replacing BVH traversal with **CUDA-based binary search**
- Superiority of RT-based TC results from a more load-balanced task scheduling
- BVH traversal on RT cores is **less efficient** for triangle counting



Evaluation of RT-based Set Intersection

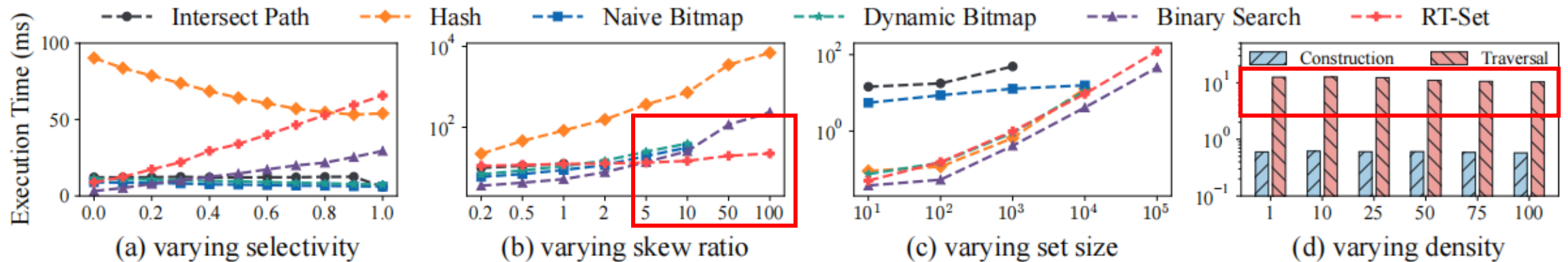
- Method description



- Intersection between two sets
- Intersection between one set and multiple sets

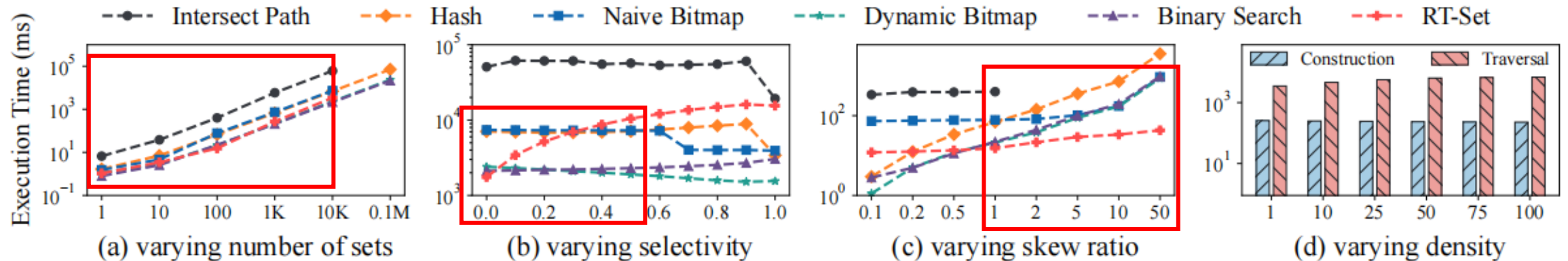
Pairwise Set Intersection

- RT-Set shows better performance when the **skew ratio is high**
- **Higher density** enhances RT-Set performance



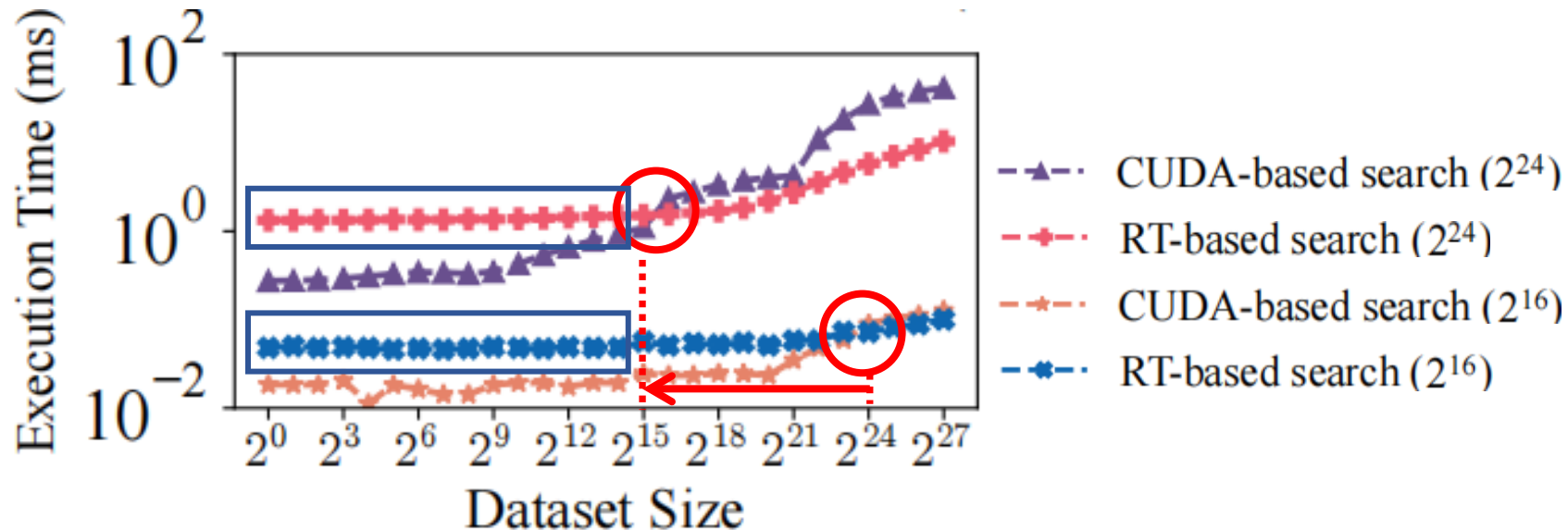
Intersection between One Set and Multiple Sets

- RT-Set has **competitive performance** when the number of sets is increasing
- **Low selectivity** favors the indexing performance of the BVH
- The advantage when the **skew ratio is high** is further magnified



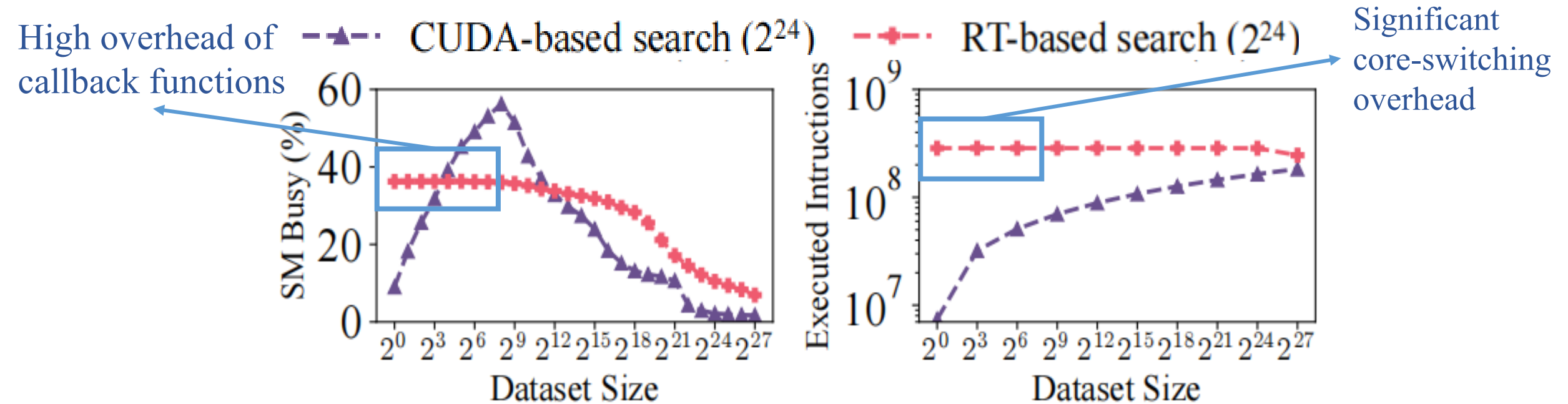
Analysis of Search Performance on RT Cores

- Comparison of RT-based search and CUDA-based search on one-dimensional sorted integers
 - Larger dataset and query size benefit RT-based search
 - RT-based search has non-trivial constant overhead independent of data size



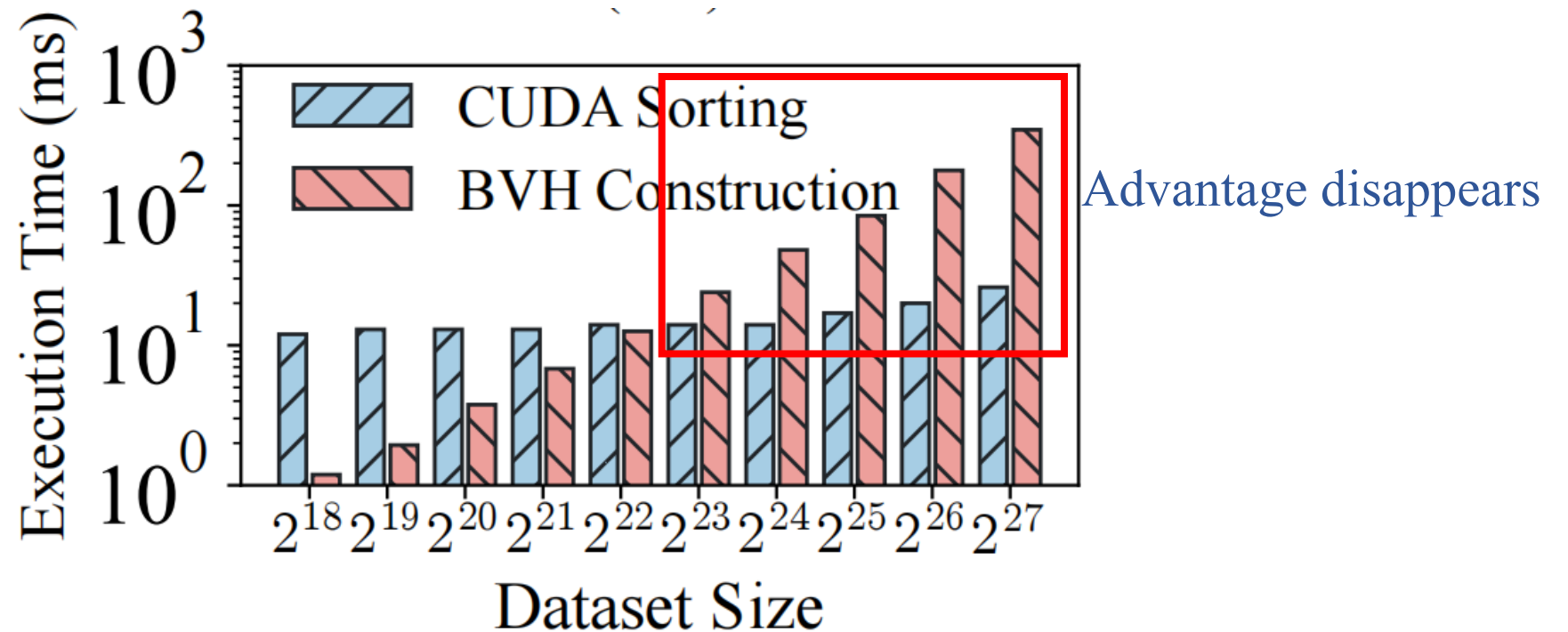
Kernel Profiling with Nsight Compute

- SM busy of RT-based search is **more stable** and callback overhead is **high**
- RT pipeline has **notable core-switching overhead**



BVH Construction Overhead Matters

- BVH Construction **scales with dataset size** while CUDA sorting remains stable



Conclusion

- RT-based methods offer little acceleration for sequential-access applications like BFS, and BVH is suboptimal for triangle counting
- For set intersection, RT cores gain advantages if the **skew ratio** between query and dataset is **high**
- The RT cores can **more efficiently** spot an element than CUDA cores, but they also incur a **constant and non-trivial overhead** from the OptiX execution pipeline
- The **high BVH construction overhead** and **high memory footprint** should be considered

Thank you!

Email: xiaozxiong@mail.sdu.edu.cn

The open-source code is available at: <https://github.com/xiaozxiong/RT-Graph>