Heterogeneous Parallel Computing for Rendering Large-Scale Data

Sung-eui Yoon

Associate Professor KAIST

http://sglab.kaist.ac.kr



Acknowledgements

Collaborators

 My students, M. Gopi, Miguel Otaduy, George Drettakis, SeungYoung Lee, YuWing Tai, John Kim, Dinesh Manocha, Peter Lindstrom, Yong Joon Lee, Pierre-Yves Laffont, Jeong Mo Hong, Sun Xin, Nathan Carr, Zhe Lin

Funding sources

- Boeing, Adobe, Samsung
- AMD, Microsoft Research Asia
- Korea Research Foundation
- MSIP, IITP





Past: Rendering Massive Geometric Data



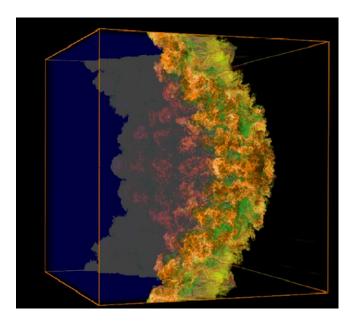
Boeing 777, 470 M tri.



Large-scale virtual world, 83 M tri.



Scanned model, 372 M tri. (10 GB)



Over 3 Terabytes of geometric data



Present: Scalable Ray Tracing, Image Search, Motion Planning

 Designing scalable graphics and geometric algorithms to efficiently handle massive models on commodity hardware



Photo-realistic rendering



Image search

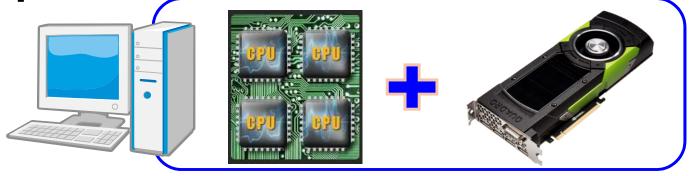


Motion planning



Recent Hardware Trends

- Multi and many cores
 - CPUs and GPUs are increasing the # of cores
- Heterogeneous architectures
 - Intel Sandy Bridge, AMD Fusion, and Nvidia Tegra embedded chips



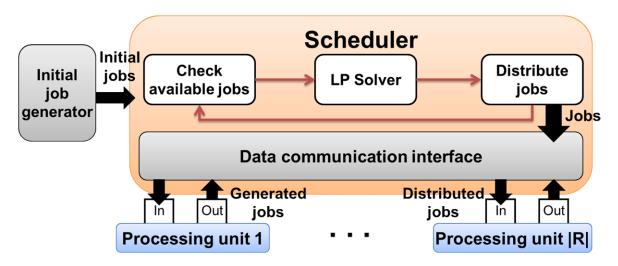
Images from NVIDIA

- Previous approaches
 - Utilize either multi-core CPUs or GPUs



Hybrid Parallel Computation for Proximity Queries

- Our initial work: manually assign jobs of continuous collision detection to CPUs and GPUs
 - Received a best paper award at Pacific Graphics, 09
- A general, job distribution algorithm for CPUs and GPUs [Kim et al., TVCG 13, Spotlight paper]



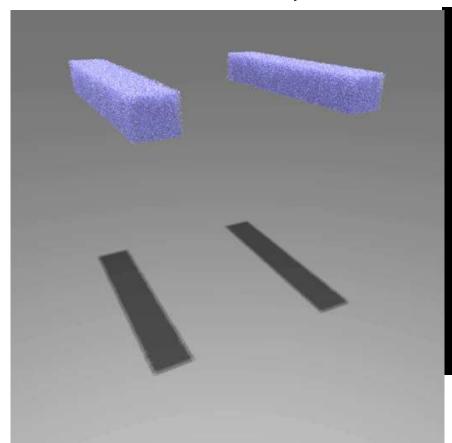


Motion planning

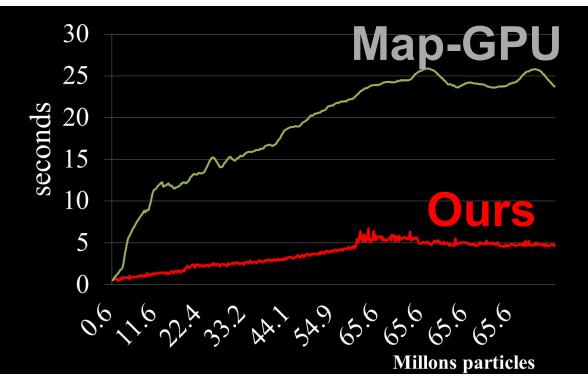
[Lee et al., ICRA 12]

Out-of-Core Proximity Computation for Particle-based Fluid Simulations [Kim et al., HPG 14]

Two hexa-core CPUs w/ 192 GB RAM GeForce GTX 780) with 3 GB video RAM



Up to 65.6 M Particles Maximum data size: 13 GB



NVIDIA mapped memory Tech
- Map CPU memory space

into GPU memory space



Heterogeneous Parallel Computing for Rendering

- T-ReX: Interactive Global Illumination of Massive Models on Heterogeneous Computing Resources, IEEE TVCG 2014
 - Manually assign tasks to CPUs and GPUs
 - Source codes are available
- Timeline Scheduling for Out-of-Core Ray Batching, High Performance Graphics (HPG), 2017
 - Automatic task assignment for high performance



T-ReX: Interactive Global Illumination of Massive Models on Heterogeneous Computing Resources

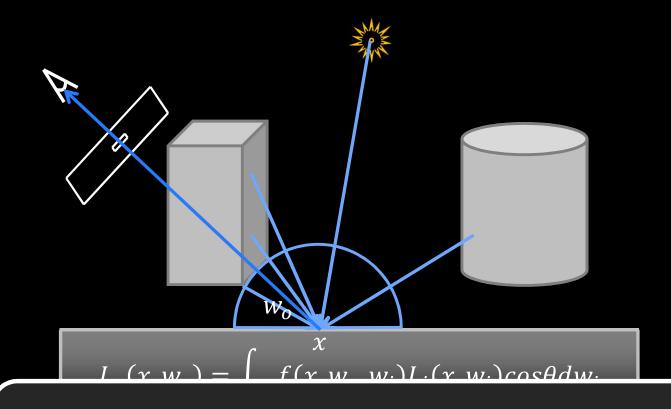
Tae-Joon Kim*, Xin Sun§, and Sung-Eui Yoon*

KAIST*, Microsoft Research Asia§

IEEE Transactions on Visualization and Computer Graphics (TVCG), 2014

Project Homepage with Codes: http://sqlab.kaist.ac.kr/T-ReX

Global Illumination



Enormous computation is necessary

Interactive Global Illumination



- Utilize GPU
- Use sparse voxel octrees
- Model complexity < 10 M tris.</p>

Massive Models

 Due to advances of modeling, simulation, and data capture techniques



CAD oil tanker, 82 M tri. (4 GB)



Boeing 777, 366 M tri. (20 GB)



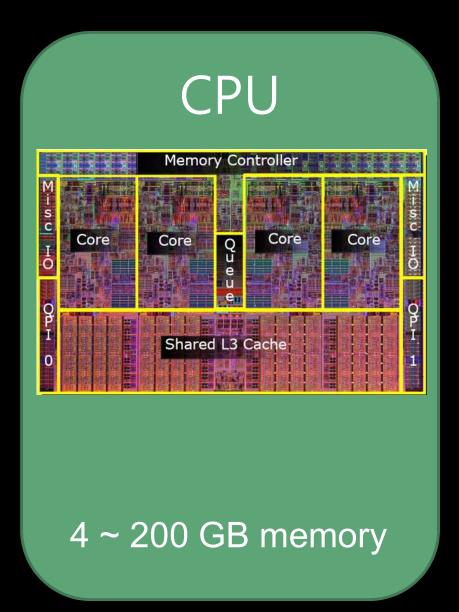
Scanned model, 372 M tri. (10 GB)

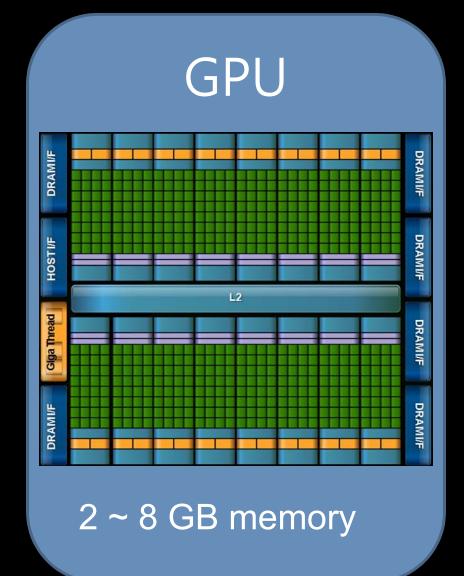
Long data access time and low I/O performance

Motivation

- Global illumination of small models can be done interactively
 - Thanks to advance of GPU architecture
- Interactive global illumination with massive models is still challenging
 - Maximize computation throughput
 - Minimize I/O requirement

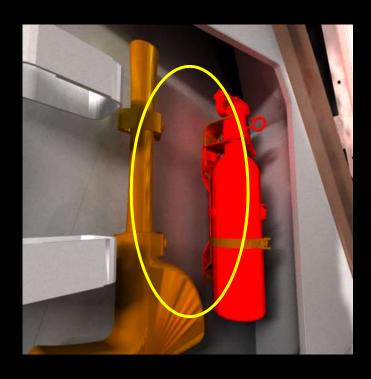
Heterogeneous Computing Resources





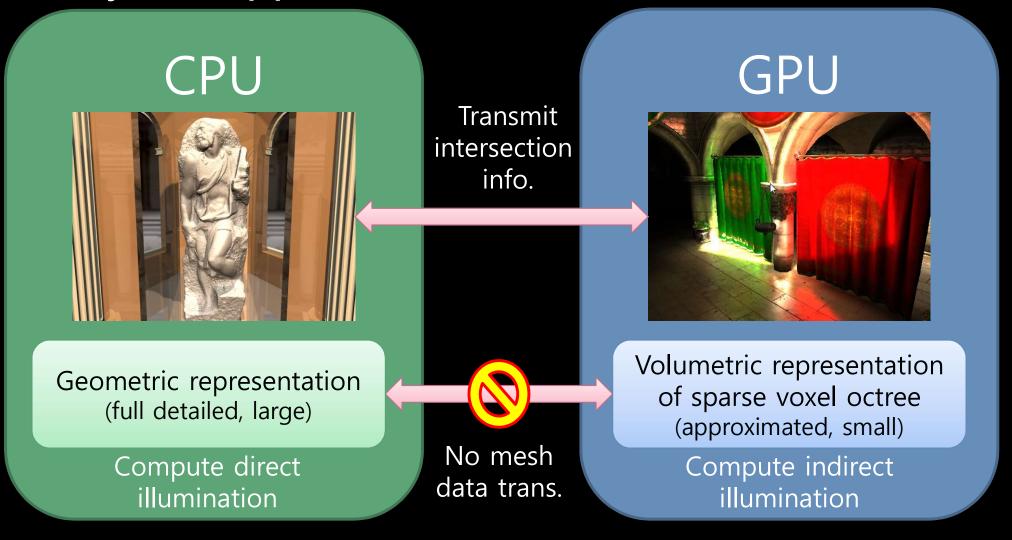
Observation

 Global illumination effect is less sensitive to geometry details

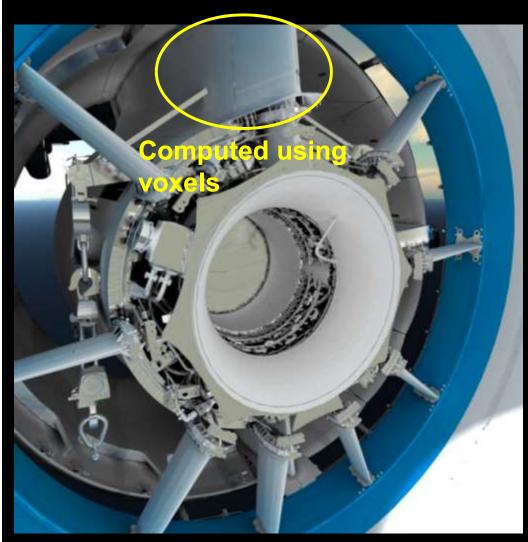


Our Approach

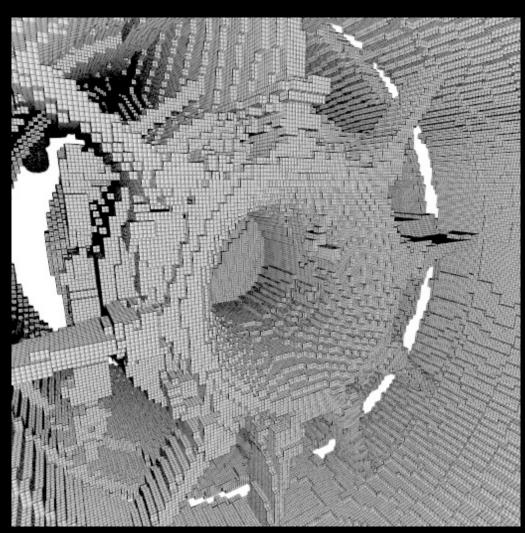
Hybrid approach



Approximated Illumination



Raw Aftesh shadahigation



Approximated voxel representation

Results

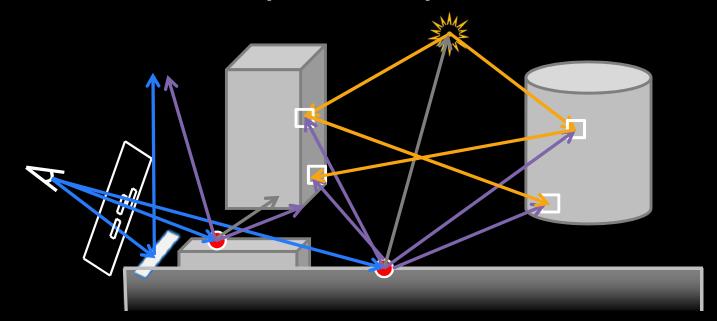
Outline

 Use photon mapping for rich visual effects e.g., color bleeding

- Classify rays into fitting processors
 - Each class of ray uses representation

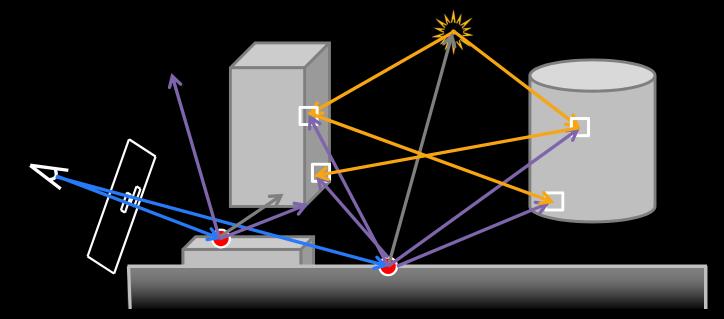
Ray Classification

- C-rays
 - More sensitive to geometry details
 - Generates high-frequency visual effects
 - The primary rays and their secondary rays reflected on perfect specular materials



Ray Classification

- G-rays
 - Less sensitive to geometry details
 - Generates low-frequency visual effects
 - Any rays other than C-rays (e.g., gathering rays, shadow rays)



Data Representations

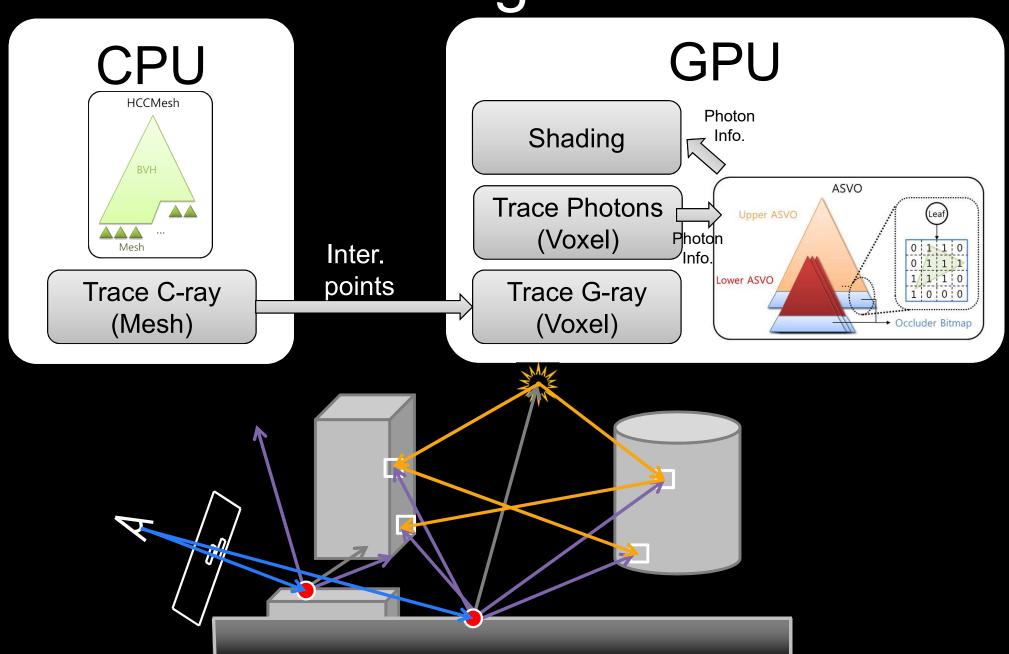
Augmented Sparse Voxel Octree (ASVO)

- GPU side volumetric representation for G-ray
- Efficiently traversed in GPU
- Approximated geometry & photon map

HCCMeshes [Kim et al. Eurographics'10]

- High quality geometry for C-ray
- Random-accessible compression (7:1 ~ 20:1)
- Supports high performance decompression

Rendering Process



Results

- Interactive responsiveness
 - About 30 ms response time for dynamic changes on cameras, materials, and lights
- High performance
 - 3 M ~ 20 M rays/s
- High complexity
 - Up to 470 M triangles

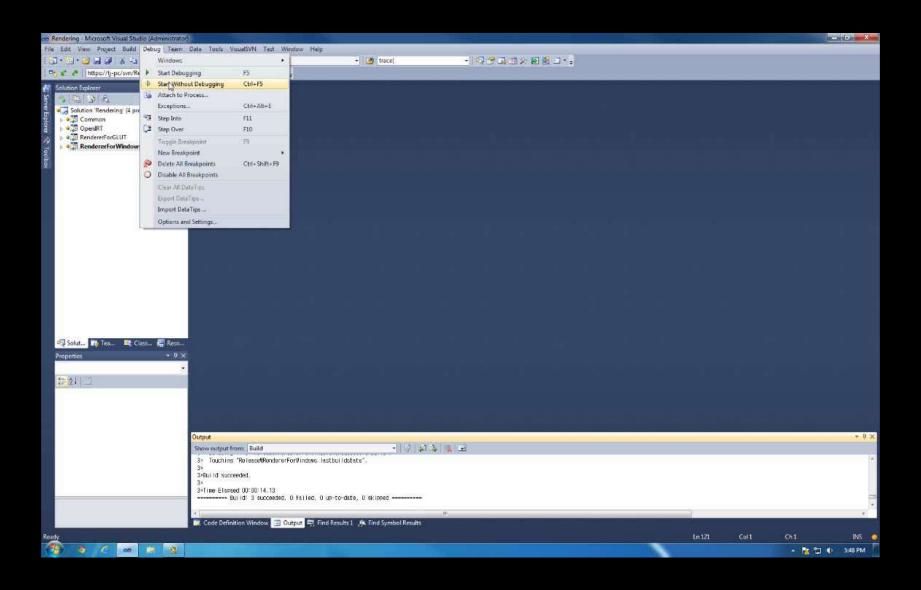
Results

- Test environment (PC)
 - Intel Core i7 CPU (hexa-core) w/ 8 GB RAM
 - NVIDIA GTX 680 card with 2 GB DRAM
 15% of GPU memory was allocated for upper ASVO
- Boeing 777 model benchmark
 - 366M Triangles
 - 15.6 GB mesh + 21.8 GB BVH for raw model
 - 6.55 GB for HCCMesh
 - 11 area lights (generated 5 M photons each)

Comparison

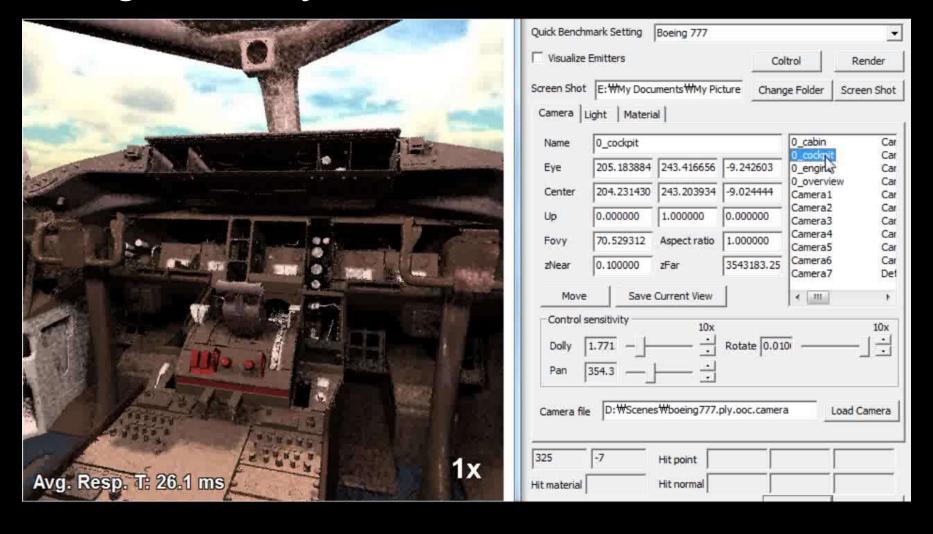
- 3.9 times improvement over CPU-only implementation
 - Same algorithm, but running on CPU only
 - Main memory holds both representations (HCCMeshes, ASVOs)
- 135 times improvement over simple photon mapping on CPU
 - Using HCCMeshes only

Demonstration



Progressive Rendering

Progressively refine the frame



Render

Screen Shot

Car

Car

Car

Car

Car Car

Car Car

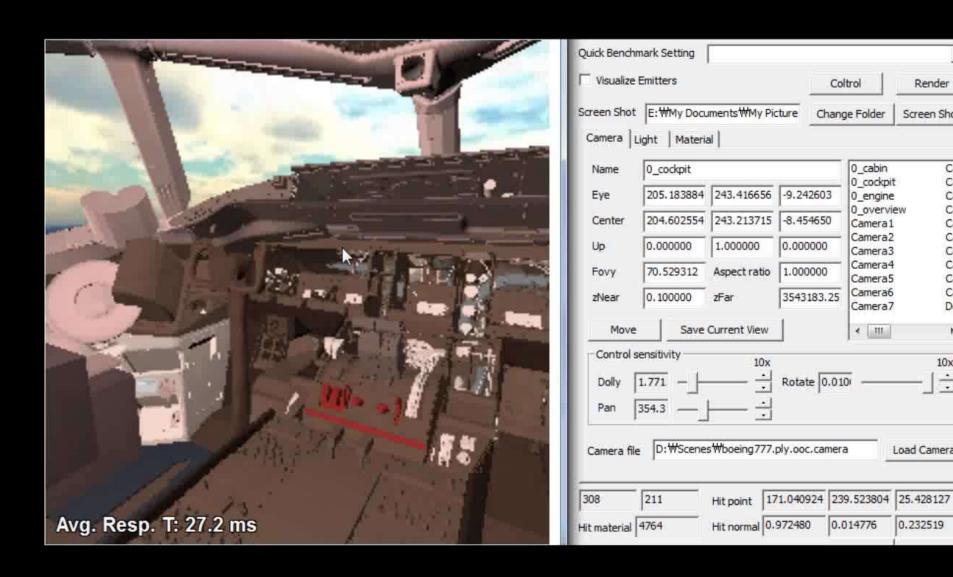
Def

10x

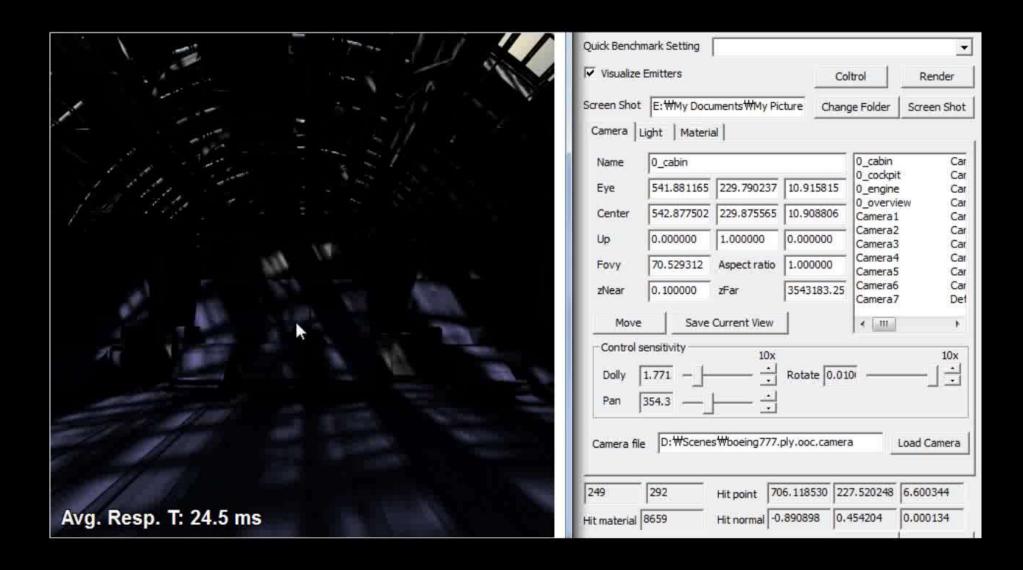
Load Camera

0.232519

Materials Changes



Lights Changes

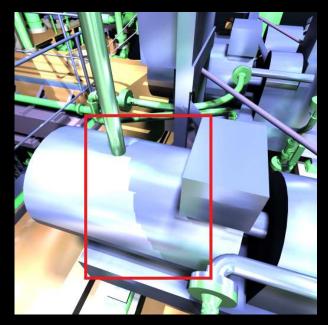


Conclusion

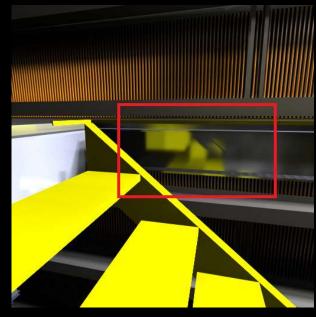
- Present an integrated progressive rendering framework for global illumination of massive models
 - Use a decoupled representation: HCCMeshes in CPU and ASVOs in GPU for handling large-scale models
- Reduce expensive transmission costs and achieve high utilizations for CPU and GPU

Limitations

- Volumetric representation
 - Biased and inconsistent
 - Spans more space than its geometric model



Point light sources



Highly glossy materials

High-Performance Graphics 2017

Los Angeles | July 28-30, 2017

TIMELINE SCHEDULING FOR OUT-OF-CORE RAY BATCHING

Myungbae Son

Sung-EuiYoon

SGVR Lab KAIST





Our Scenario

- Complex scenes
 - Out-of-core model: Too big data!
 - Cannot be stored in main / GPU memory
- Complex device configurations
 - Distributed memory cluster system
 - Client-assisted remote rendering
 - Renderfarm of heterogeneous devices



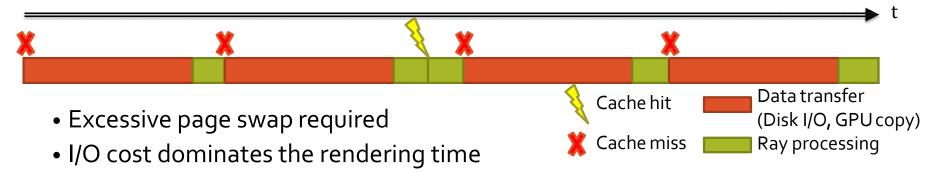
Boeing 777, 366 M tri. (20 GB)





Challenges

- Massively complex scene
 - Over **96%** of runtime is spent on I/O in naïve BDPT (Boeing777)



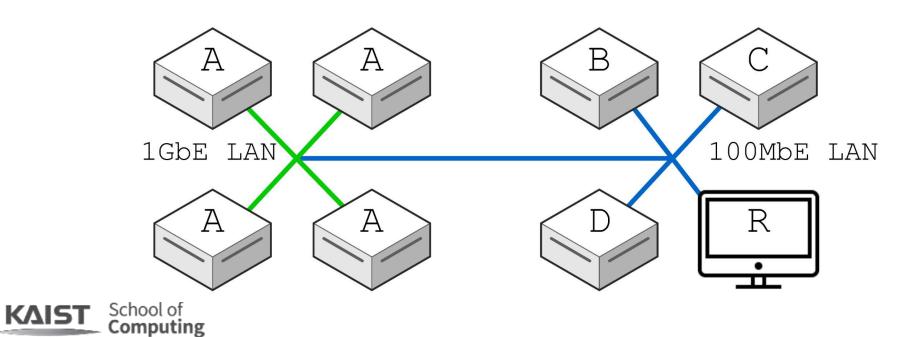
- Global Illumination with incoherent rays
 - Efficient ray scheduling is required





Challenges

Complex and heterogenenous device configurations...

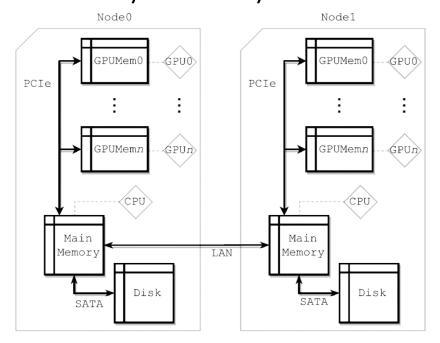




Challenges

Further down to the processor and memory hierarchy level...

- Different processors
- Different memory channels
- Different nodes and network







Goal & Contributions

Design a scheduler for global illumination

- Processes massive models
- Supports variety of computing environments
 - Complex and heterogeneous device configurations

Our contributions

- A modeling technique: device configurations and jobs
- A scheduling algorithm: Greedy Makespan Balancing (GMB)
- An adaptation to path tracer





OURAPPROACH





Our Approach

- Formulation technique for MC ray tracing jobs Device Connectivity Graph (DCG) and Timing Model
- Timeline scheduling and Greedy Makespan Balancing algorithm Simple, iterative algorithm that considers utilization and latency hiding
- Adaptation to actual renderer framework
 Out-of-core path tracer





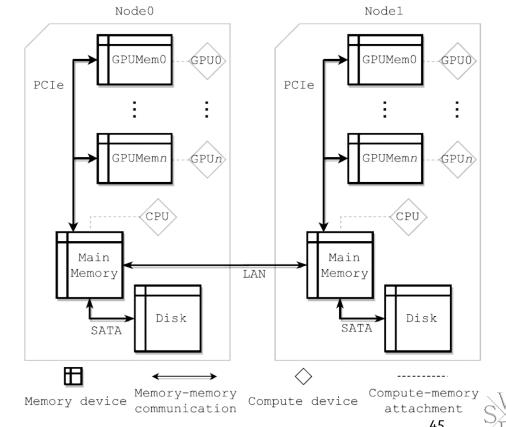
Formulation: Device Connectivity Graph

- Graph of memory devices
 - Memory
 Disk storage, RAM, GMEM
 - Connections (Channels)
 PCle (RAM ↔ GMEM)
 SATA (Disk ↔ RAM)

LAN (RAM \leftrightarrow RAM)

...

• Stores bandwidth information





Formulation: Timing Model

- Assume simple yet efficient linear model on time
 - Job execution

$$T_{EXEC}(d, j, W) = \begin{cases} 0, & if W = \emptyset \\ T_{SETUP}(d, j) \\ + T_{RATE}(d, j) \cdot (|w_1|, |w_2|, \dots), \end{cases}$$
 otherwise

• Data transfer

$$T_{TRANS}(d_i \rightarrow d_j, w) = T_{LAT}(d_i \rightarrow d_j) + \frac{|w|}{T_{BW}(d_i \rightarrow d_j)}$$

- Fitting each parameter (T_{SETUP} , T_{RATE} , T_{LAT} , T_{BW})
 - Use least squares method on test run





Our Approach

- Formulation technique for MC ray tracing jobs
 Device Connectivity Graph (DCG) and Timing Model
- Timeline scheduling and Greedy Makespan Balancing algorithm Simple, iterative algorithm that considers utilization and latency hiding
- Adaptation to actual renderer framework
 Out-of-core path tracer

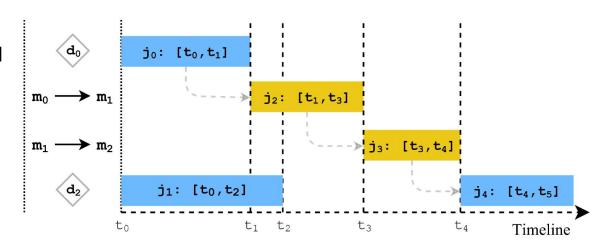




Timeline Scheduling

- A representation of schedule with timing constraints

 - For ←→ memory channels
 Data transfers are allocated
 - Dependencies between jobs and fetches

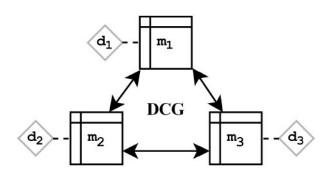


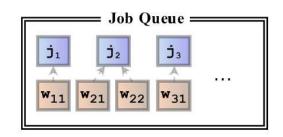
<u>Def.</u> schedule: a set of timelines that jobs and fetches are allocated

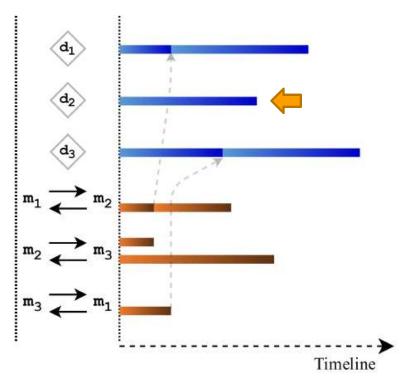




Greedy Makespan Balancing Algorithm





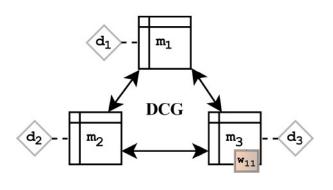


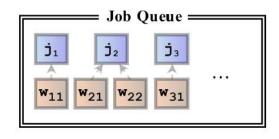


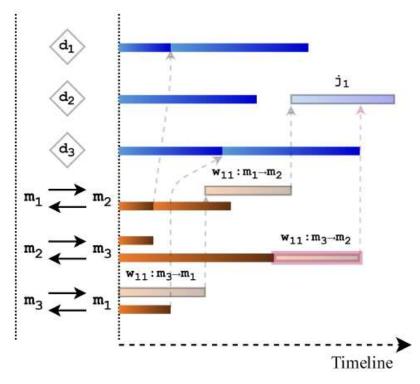
1. Choose least occupied compute device \boldsymbol{d}



Greedy Makespan Balancing Algorithm





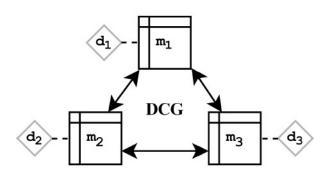


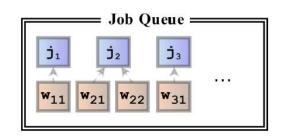


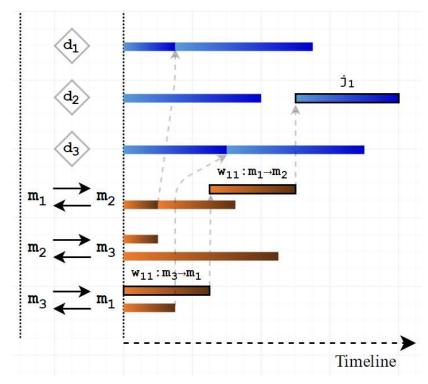
2. Find job j_i that can be run at d as soon as possible



Greedy Makespan Balancing Algorithm









4. Repeat until devices are occupied enough



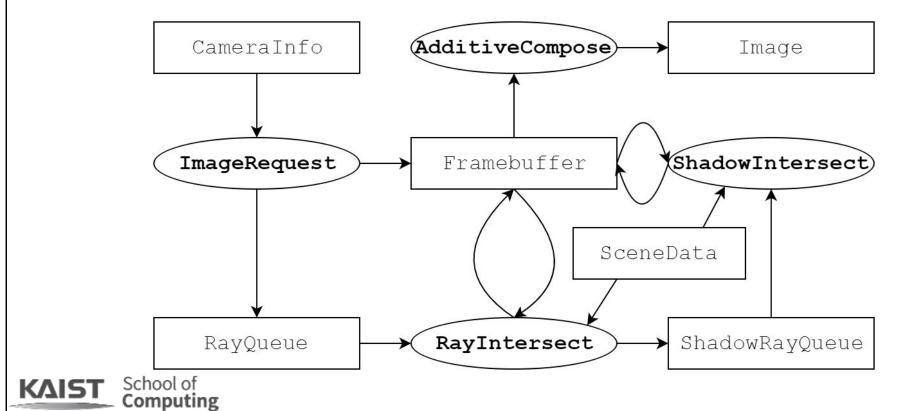
Our Approach

- Formulation technique for MC ray tracing jobs
 Device Connectivity Graph (DCG) and Timing Model
- Timeline scheduling and Greedy Makespan Balancing algorithm Simple, iterative algorithm that considers utilization and latency hiding
- Adaptation to actual renderer framework
 Out-of-core path tracer





Out-of-core Path Tracer Jobs





RESULTS





Benchmark scene



Boeing777 (26.5GB, 496M tri, 5.2sec/img)



SponzaMuseum (12.3GB, 245M tri, 34.8 sec/img)

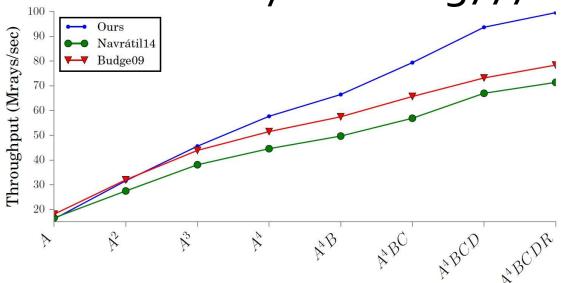
 $(800 \times 800 \times 32spp \times 60frames)$

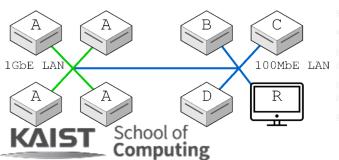
- Model preparation
 - Even-sized median-split kdtree, 27 / 26 subdivision, respectively





Horizontal Scalability – Boeing777

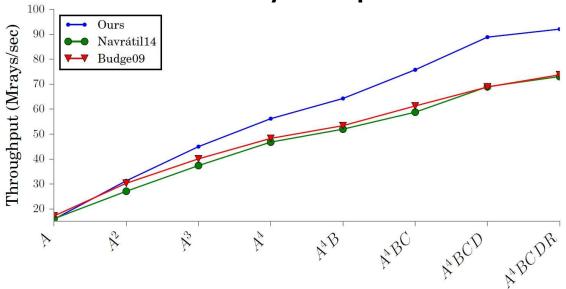


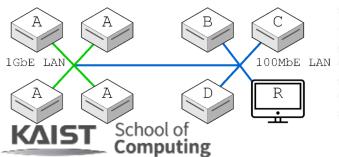


Type	CPU	Main memory	GPU Memory	GPU	Note
A	i7-4770K 3.5GHz octa-core	DDR3 8GB	6GB	GTX Titan	1GbE LAN, 4 nodes
В	i7-4790K 4GHz octa-core	DDR3 8GB	6GB	GTX Titan	
C	Xeon E5-2690 2.9GHz 16-core	DDR3 8GB	6GB	GTX Titan	
D	Xeon E5-2690 2.6GHz 16-core	DDR3 8GB	6GB	GTX Titan X	
R	i7-3770k 3.5GHz quad-core	DDR3 8GB	4GB	GTX980	



Horizontal Scalability – Sponza Museum



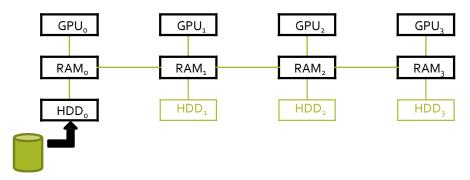


Type	CPU	Main memory	GPU Memory	GPU	Note
A	i7-4770K 3.5GHz octa-core	DDR3 8GB	6GB	GTX Titan	1GbE LAN, 4 nodes
В	i7-4790K 4GHz octa-core	DDR3 8GB	6GB	GTX Titan	6
C	Xeon E5-2690 2.9GHz 16-core	DDR3 8GB	6GB	GTX Titan	
D	Xeon E5-2690 2.6GHz 16-core	DDR3 8GB	6GB	GTX Titan X	
R	i7-3770k 3.5GHz quad-core	DDR3 8GB	4GB	GTX980	



Efficiency on Data Fetching

Central scene DB scenario

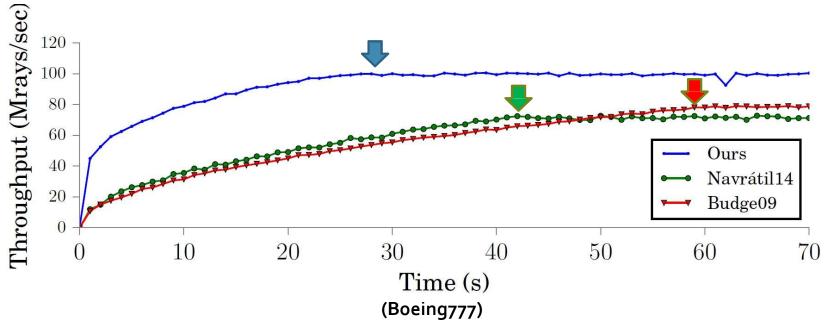


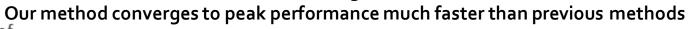
- Initially no data at slave nodes at all
- The master node gives scene data blocks on-demand





Efficiency on Data Fetching









Conclusion

- Presented specification techniques for out-of-core MC ray tracing on arbitrary hardware setup
 - DCG and timing model
- Presented a timeline based scheduling algorithm
 - GMB algorithm
- Applied to the out-of-core path tracer
 - Prediction technique for future rays





Conclusions

 Two different techniques, manual assignment and automatic approaches, for large-scale rendering

Released a free book on rendering

 Working on a journal version of our tutorial



