A Fast Cache-Oblivious Mesh Layout with Theoretical Guarantees





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Visualization & Massive Data Sets



TERA-10

CEA supercomputer

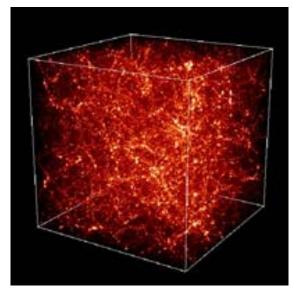
19th Top500

9968 Itanium2

60 Tflops

30 TB of memory

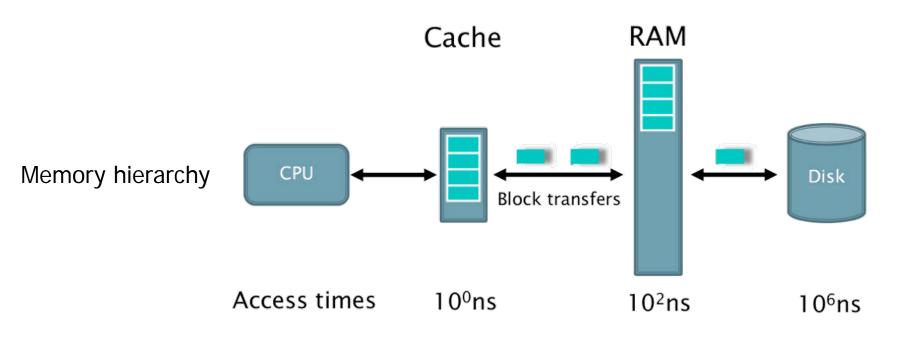
1 PB of disk space



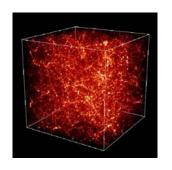
Simulation of half the observable universe 50 TB mesh

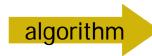
We need good I/O performance to visualize such data sets

Cache-Efficient Ordering of a Mesh



How to lay out a mesh in memory to minimize cache misses?







Outline

- Cache-aware and cache-oblivious models example: matrix multiplication
- Previous work on mesh layouts
- Our algorithm
- Experiments

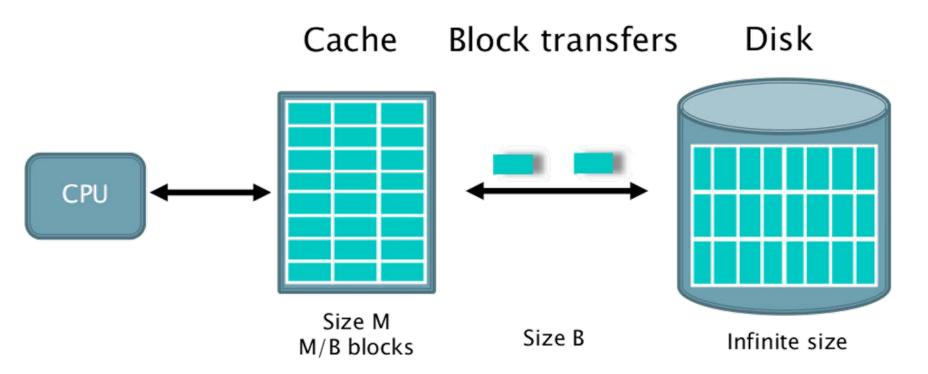
Cache-Aware Model (CA)

W: #operations

Q: #block transfers

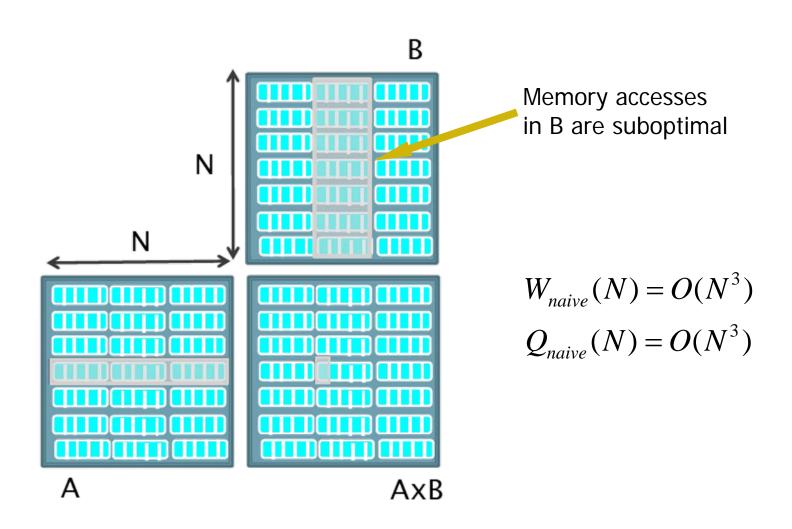
[Aggarwal & Vitter 1988]

or external memory out-of-core disk access machine I/O model



Multiplying in the CA Model

NxN matrices in row-major order: naive algorithm doesn't work



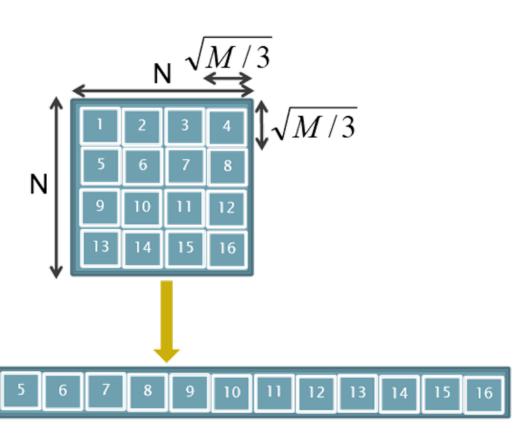
Multiplying in the CA Model

NxN matrices in submatrices

Technique used in BLAS

$$Q_{naive}(N) = O(N^3)$$

$$Q_{CA}(N) = O\left(\frac{N^3}{B\sqrt{M}}\right)$$

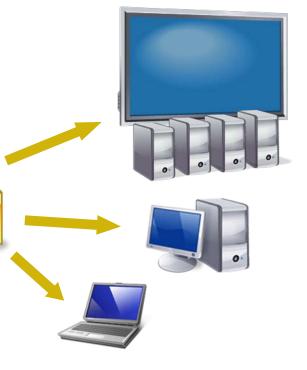


Drawbacks of the CA Model

- Only two levels of the memory hierarchy
 - · At least 4 levels on any modern CPU
 - Even deeper with multiprocessors computers
- + Efficient with all levels of the memory hierarchy

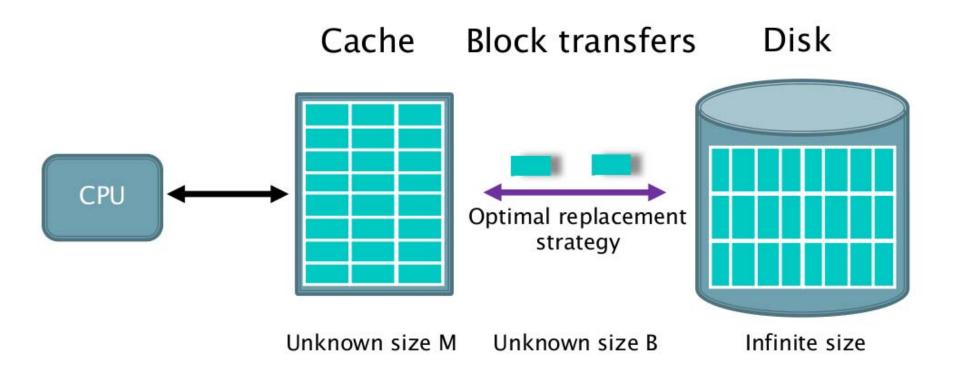
+ Architecture independent

- Architecture dependent
 - Difficult to find optimal B and M (ATLAS)
 - GPU memory hierarchy is complex
 - B and M can vary over time
 - Need to recompute the layout



Cache-Oblivious Model (CO)

[Frigo & al 1999]



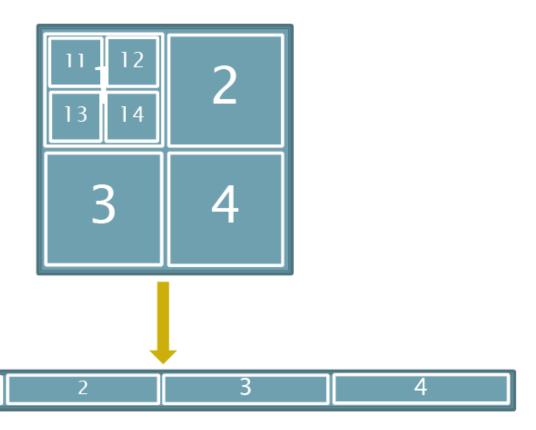
Multiplying in the CO Model

D&C matrix multiplication using a <u>recursive</u> layout

$$Q_{naive}(N) = O(N^3)$$

$$Q_{CA}(N) = O\left(\frac{N^3}{B\sqrt{M}}\right)$$

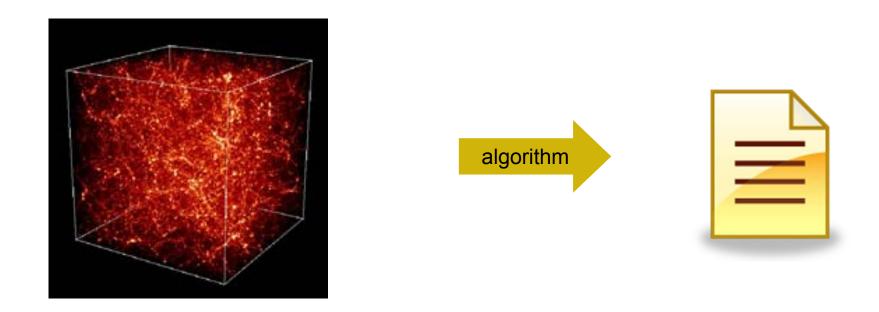
$$Q_{CO}(N) = O\left(\frac{N^3}{B\sqrt{M}}\right)$$



Outline

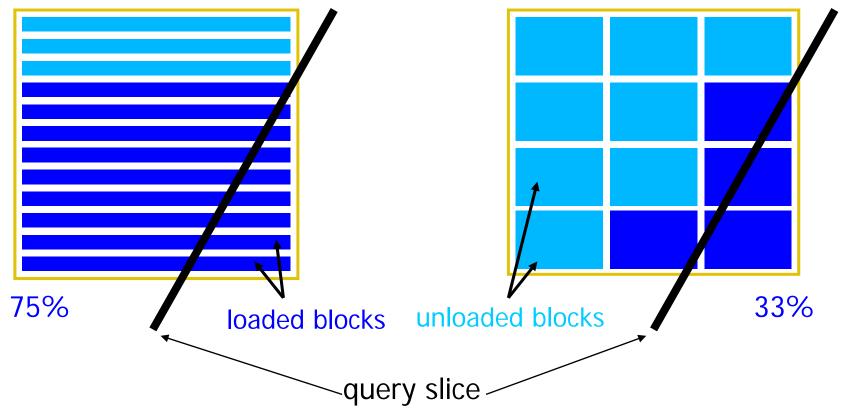
- Cache-aware and cache-oblivious models example: matrix multiplication
- The mesh layout problem and previous work
- Our algorithm
- Experiments

How to lay out a mesh efficiently in memory?

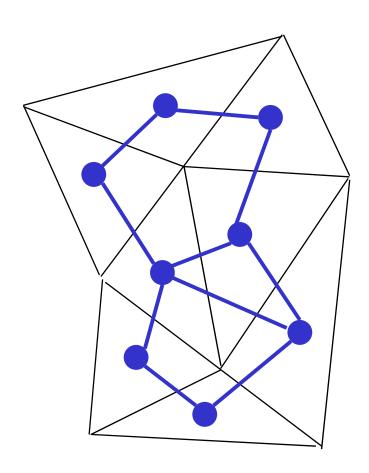


Idea

Triangles (or vertices) that are most likely to be accessed sequentially should be stored nearby

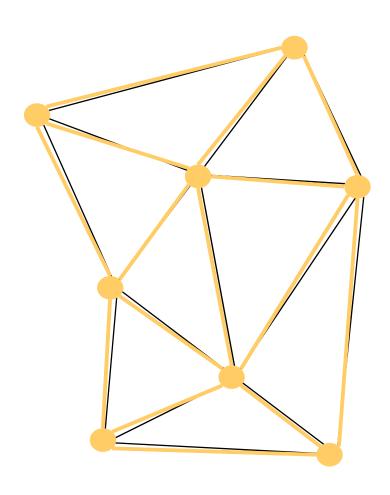


Graph of Sequential Accesses



G₁: sequential access between triangles

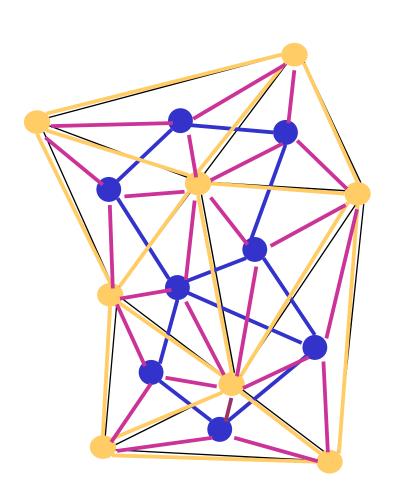
Graph of Sequential Accesses



G₁: sequential access between triangles

G₂: sequential access between vertices

Graph of Sequential Accesses



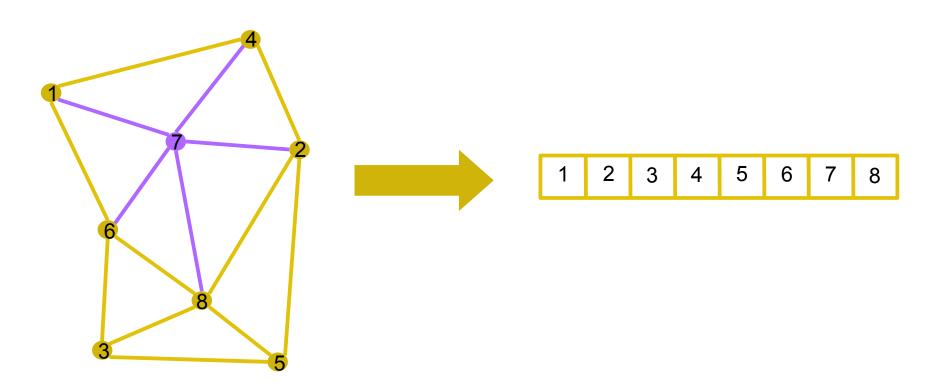
G₁: sequential access between triangles

G₂: sequential access between vertices

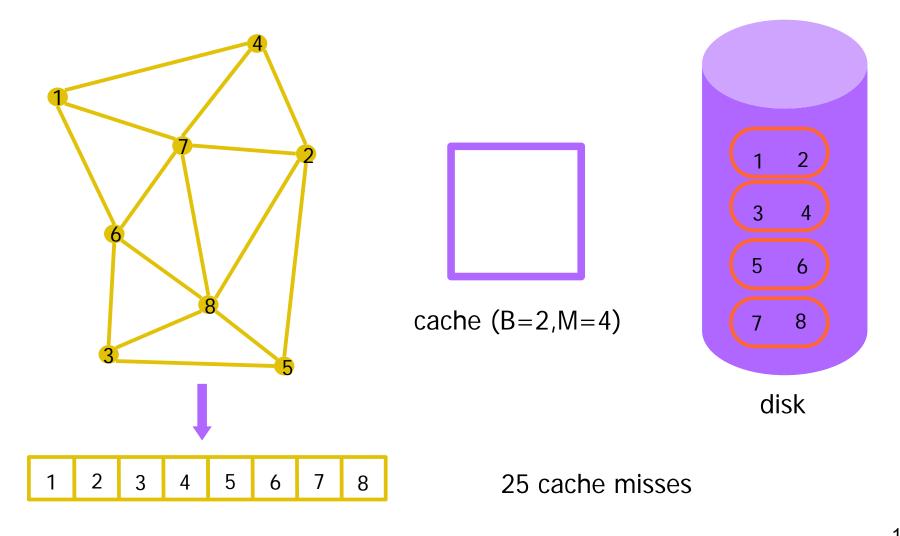
G₃: both

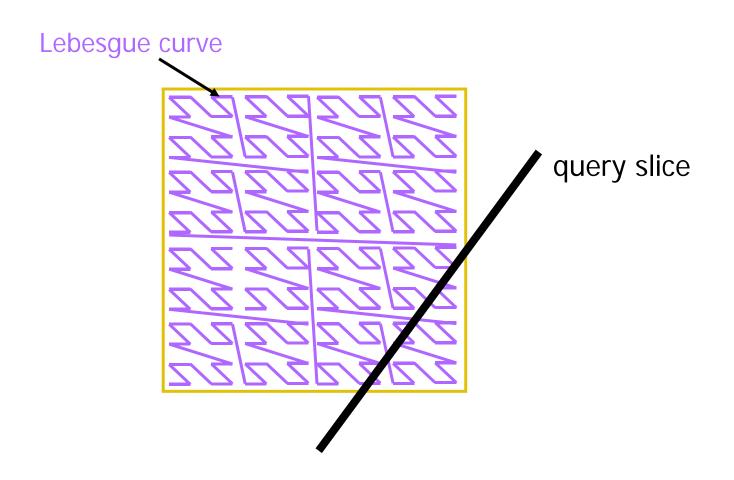
Mesh Layout Problem

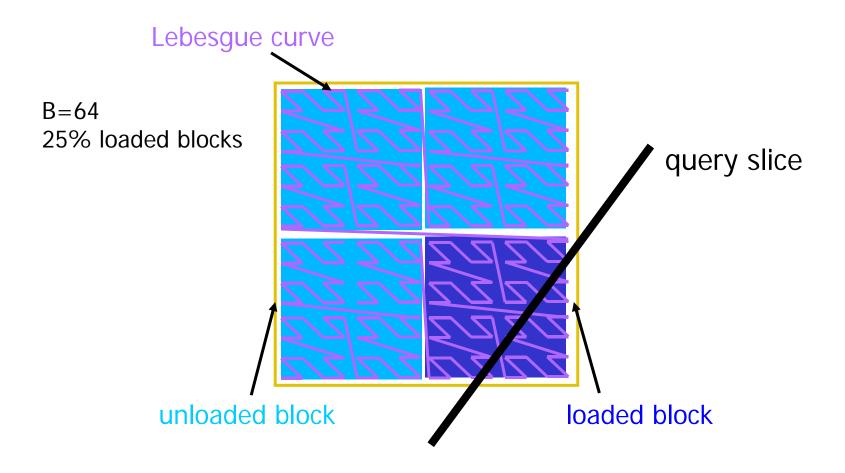
Minimize # of cache misses if each node touches all its neighbors

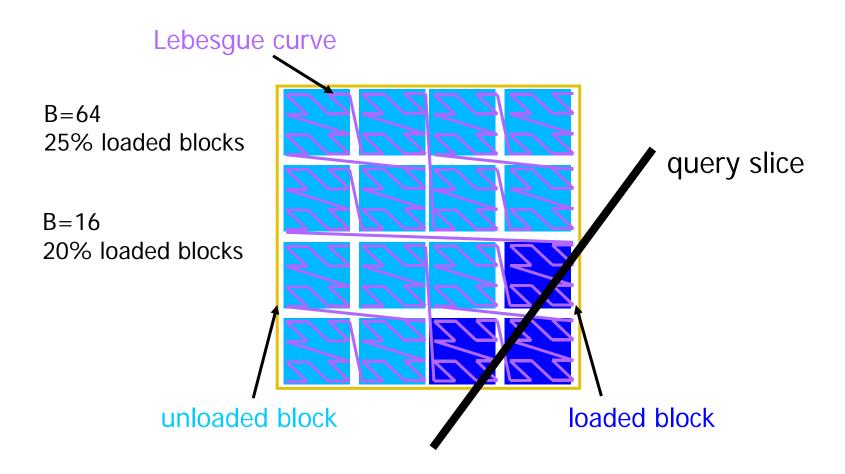


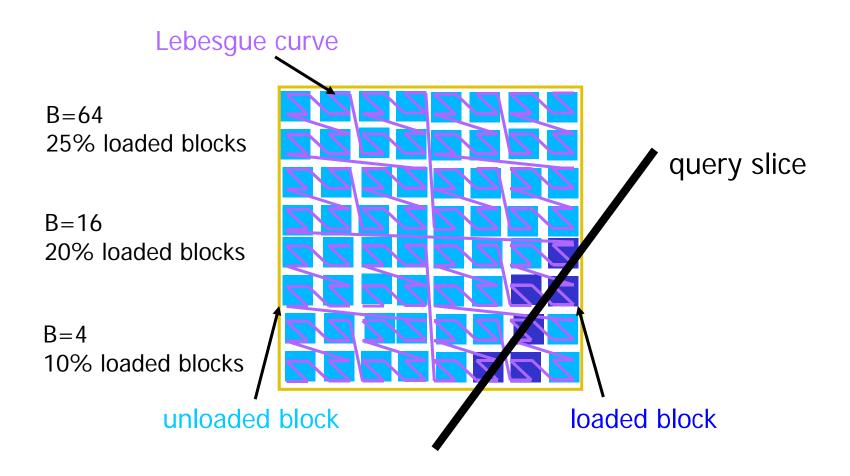
Example







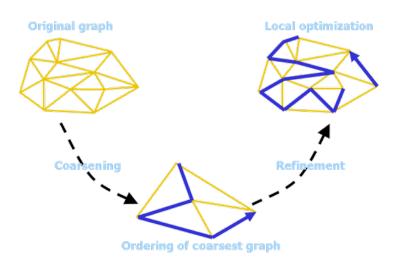




Unstructured Meshes

[Pascucci & al SIGGRAPH 2005, OpenCCL]

- Heuristic algorithm based on multi-level optimization
 - slow
 - high memory usage
- Good experimental results (2-5x improvement)
- But no guarantee on
 - time to compute the layout
 - layout quality



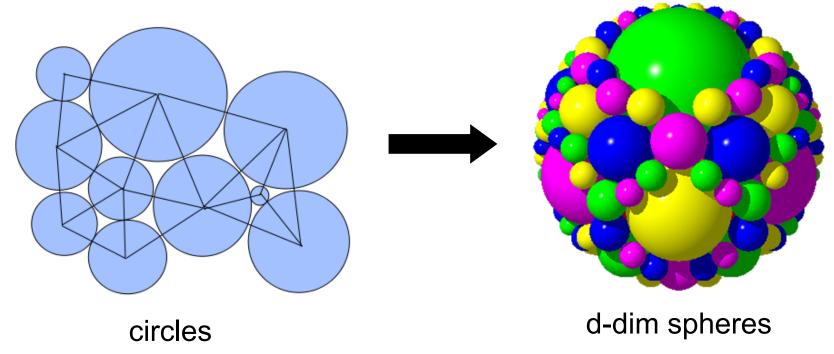
Outline

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[Miller & al 98]

Overlap graphs

- Generalize planar graphs
- Contain well-shaped meshes

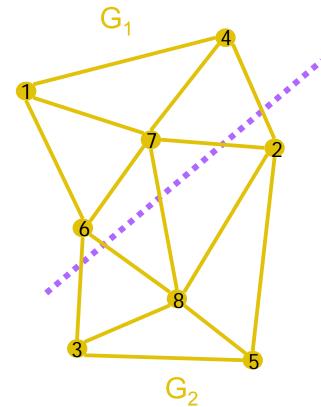


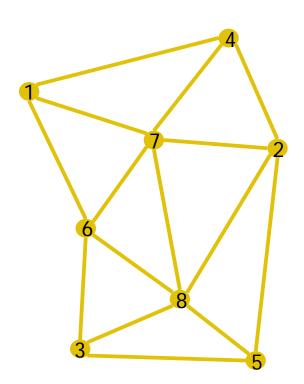
- Separate the mesh into two roughly equal-size pieces cutting few edges
- Planar graphs [Lipton-Tarjan]

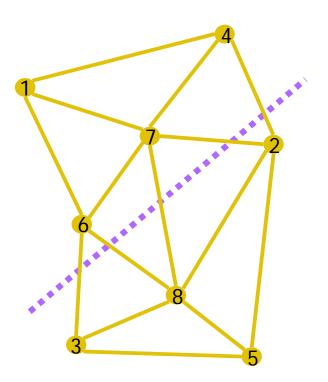
$$|G_1|, |G_2| \le \frac{2}{3}|G|$$
 $E(G_1, G_2) \le \sqrt{8|G|}$

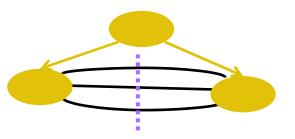
Overlap graphs (randomized linear time)

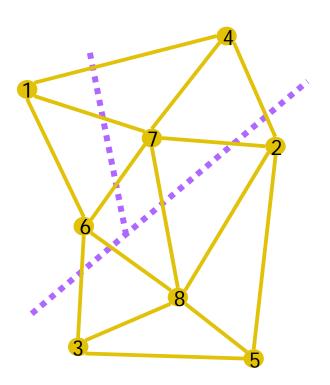
$$|G_1|, |G_2| \le \frac{d+1}{d+2}|G|$$
 $E(G_1, G_2) = O(|G|^{1-\frac{1}{d}})$

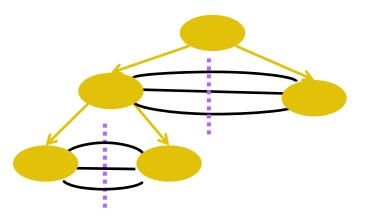


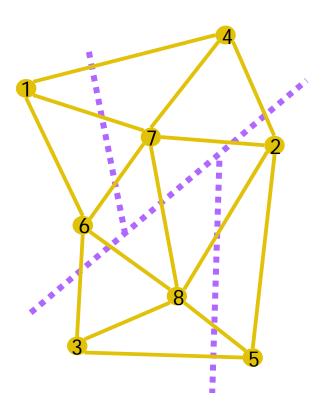


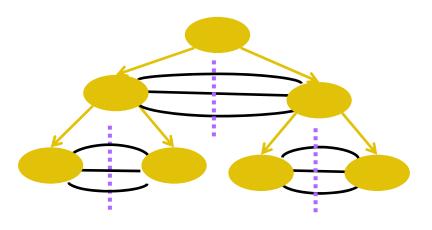




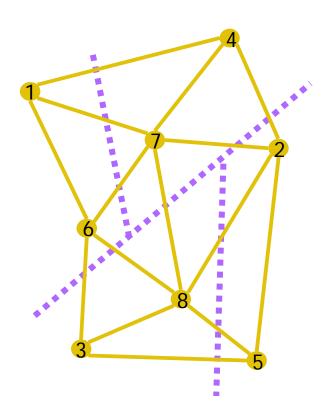


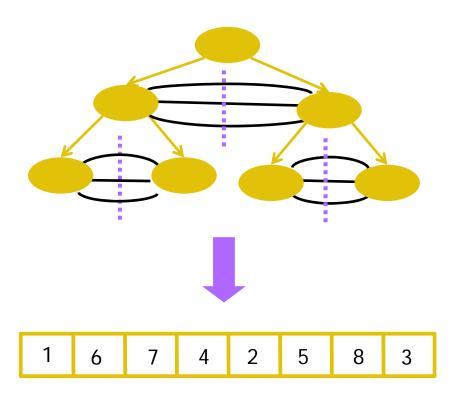






- Recursively cut the mesh $W(N) = O(N \log N)$
- The order of the leaves gives the layout



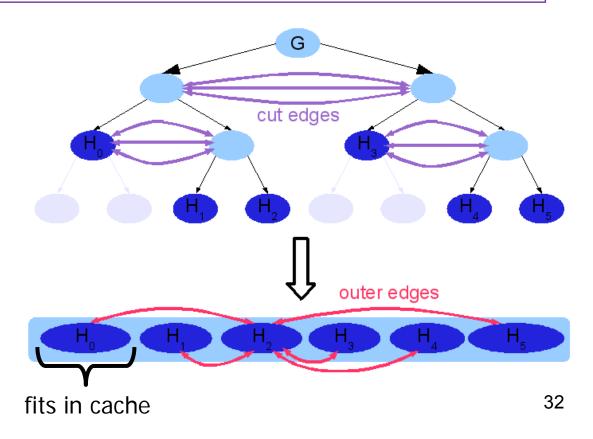


Guarantee on the Quality of the Layout

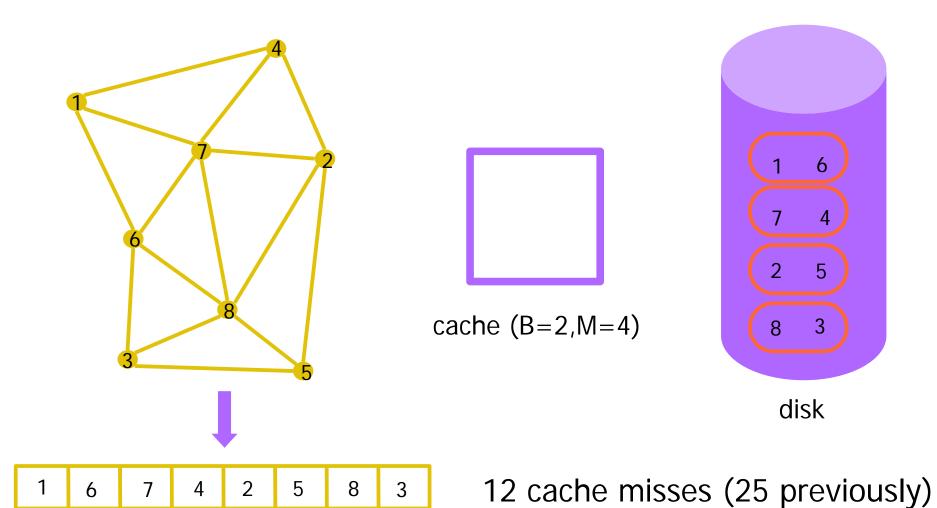
Theorem:

Our layout guarantees that a traversal of an O(N)-size d-dim mesh causes less than O(N/B+N/M^{1/d}) cache misses

- Each subgraph fits in cache
- Edges inside a subgraph do not cause a cache miss
- Cache misses are bounded by the number of edges between two subgraphs (outer edges)
- One can show that there are few outer edges



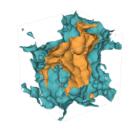
Back to the Example

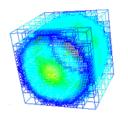


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Experiments









	Plasma	Reactor	Skull	Neptune
Туре	Structured	AMR	Unstructured	Unstructured
#Vertices	274k	84k	37k	2M
#Cells	1.3M tetra	78k hexa	156k tetra	4M tri
Size	47MB	8 MB	6 MB	169 MB

Opteron 875 2,2Ghz

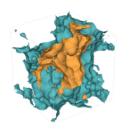
L1 = 64K

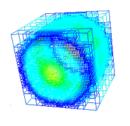
L2 = 1024K

Cache lines = 64B

32G of RAM

Layout Computation







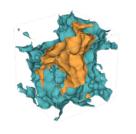


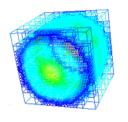
	Our Layout		OpenCCL*	
	Time (s)	Memory (MB)	Time (s)	Memory (MB)
Plasma	107	124	282	6,843
Reactor	8.8	15	27.6	458
Skull	10.6	16	26.9	814
Neptune	410	269	843	20,500

At least twice as fast using 30 times less memory

^{*} www.cs.unc.edu/~geom/COL/OpenCCL/

Layout Quality







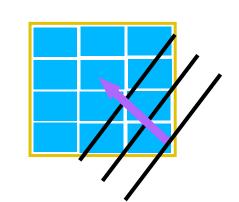


	Original Layout	Our Layout	OpenCCL
	Time (s) (Dev.)	Time (s) (Dev.)	Time (s) (Dev.)
Plasma	32.4 (0.25)	33.4 (0.22)	32.5 (0.22)
Reactor	3.3 (0.05)	3.4 (0.09)	3.3 (0.10)
Skull	5.1 (0.04)	4.96 (0.02)	4.95 (0.02)
Neptune	121 (2.5)	110 (1.3)	110 (1.0)

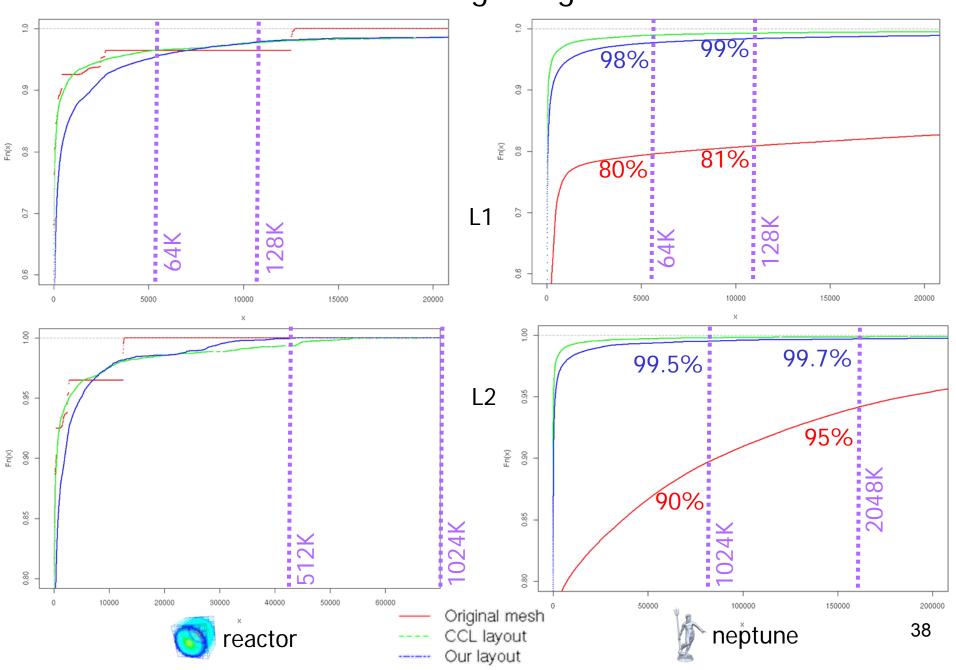
5% 00/

9%

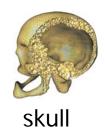
- VTK 5
- a cut plane is moved through the whole mesh
- experiment is repeated 30 times

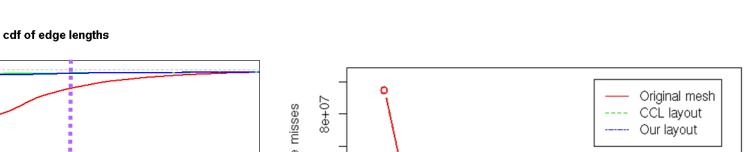


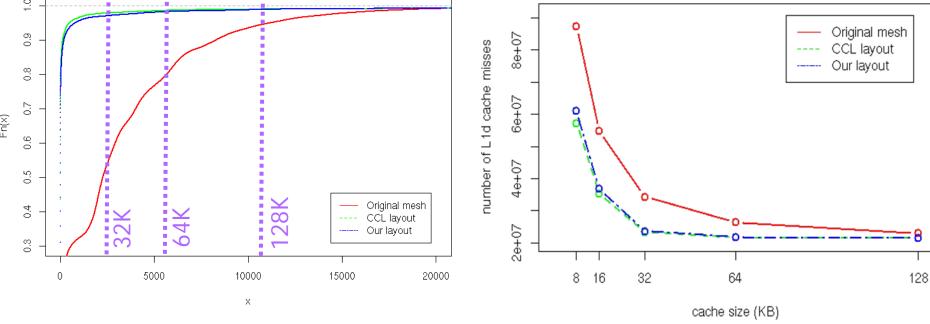
Cdf of edge lengths



Correlation Edge Lengths / Cache Misses







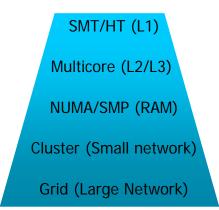
Simulation (valgrind) of the number of cache misses with varying L1 cache sizes

Conclusion & Future Work

- Our algorithm
 - Fast O(N log N)
 - Quality & time guarantees
 - Architecture independent
- Better validation
 - Big 3D unstructured meshes
 - Find the factor 2 improvement of Pascucci SIGGRAPH05
- Improve the layout
 - What if only part of the mesh is accessed?
 - Parallelism



Figure 9: **Dynamic Simulation:** Dragons consisting of 800K triangles are dropping on the Lucy model consisting of 28M triangles. We obtain 2 times improvement by using COL on average.



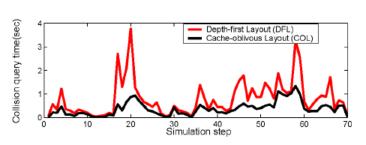


Figure 11: **Performance of Collision Detection:** Average query times for collision detection between the Lucy model and the dragon model with COL and DFL are shown. We obtain 2 times improvement in the query time on average.

Questions?

Thank you