Spatial Data Structures

CS425: Computer Graphics I

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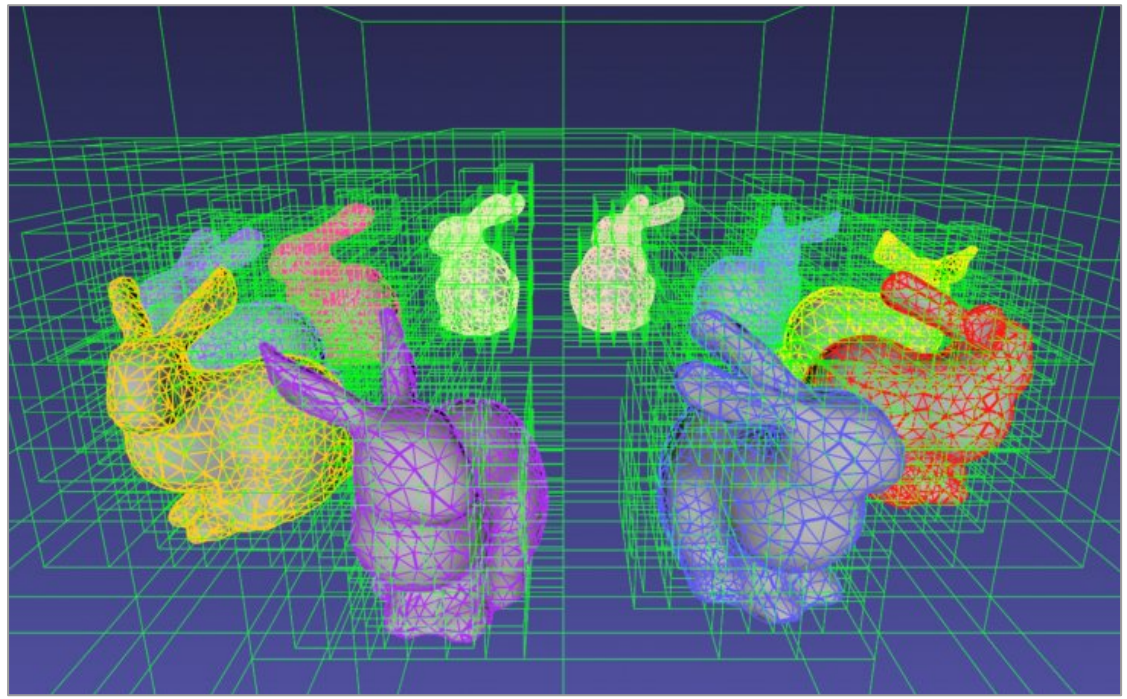
Overview

- Spatial data structures
  - Uniform grid
  - Nested grids
  - Quadtree / octree
  - K-d tree
  - Bounding Volume Hierarchy
  - Efficient sparse voxel octrees
  - High resolution sparse voxel DAGs

How to efficiently organize objects?

- 3D data contains spatial information.
- How to perform queries when there are thousands / millions of objects (points, polygons)?
  - Ray-scene intersection.
  - Proximity queries
  - Point in polygon.
  - Range query.

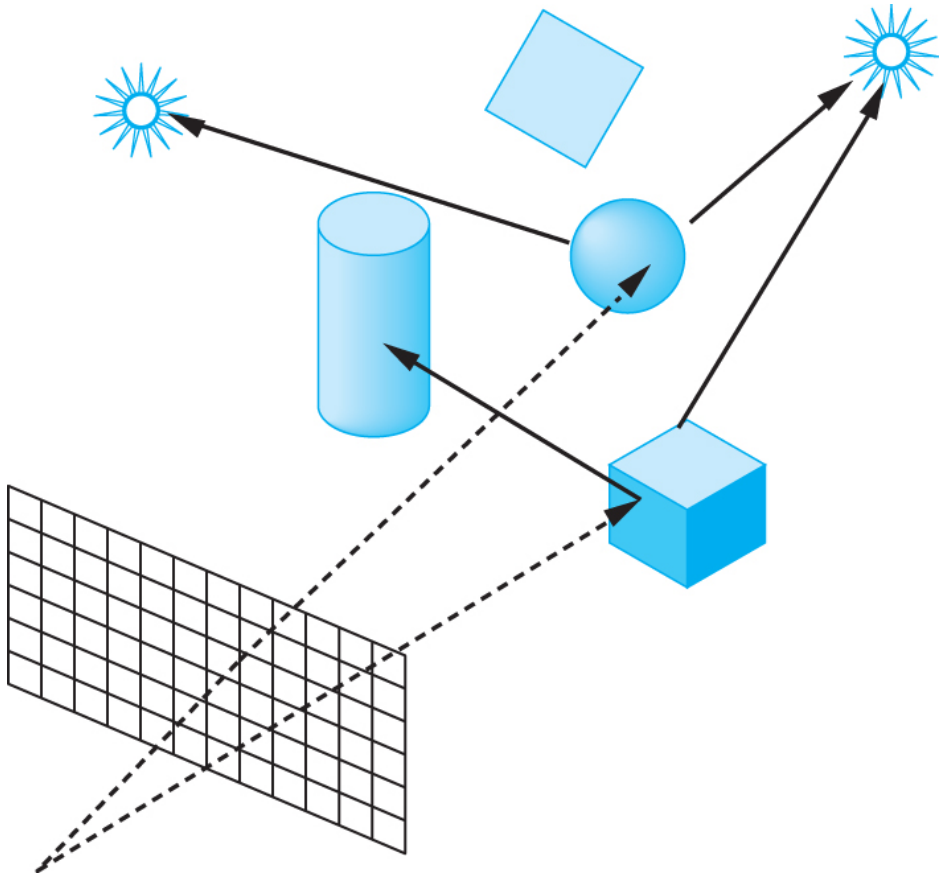


Ray-scene intersection

- Given a scene with n primitives and a ray r, find the closest point of intersection of r with the scene.

```
function intersectObjects(ray, scene) {
    for(var i=0; i < scene.objects.length; i++) {
        var object = scene.objects[i];
        var dist = intersection(ray, object);
        // ...
    }
}
```

- Complexity: $O(n)$
- How to do better?

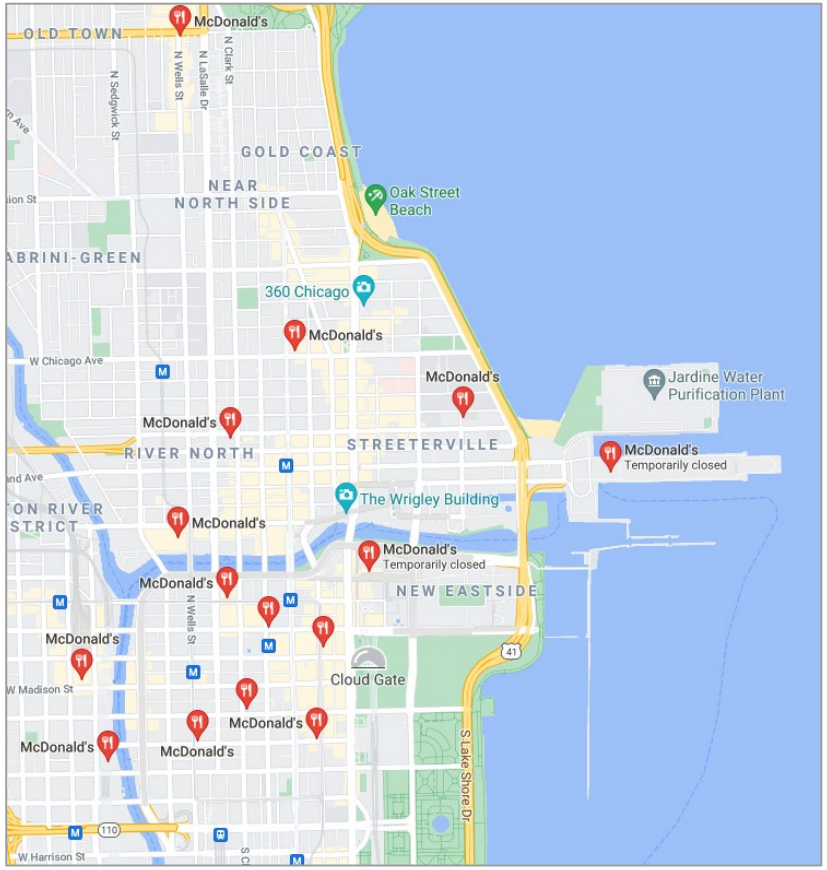


Proximity query

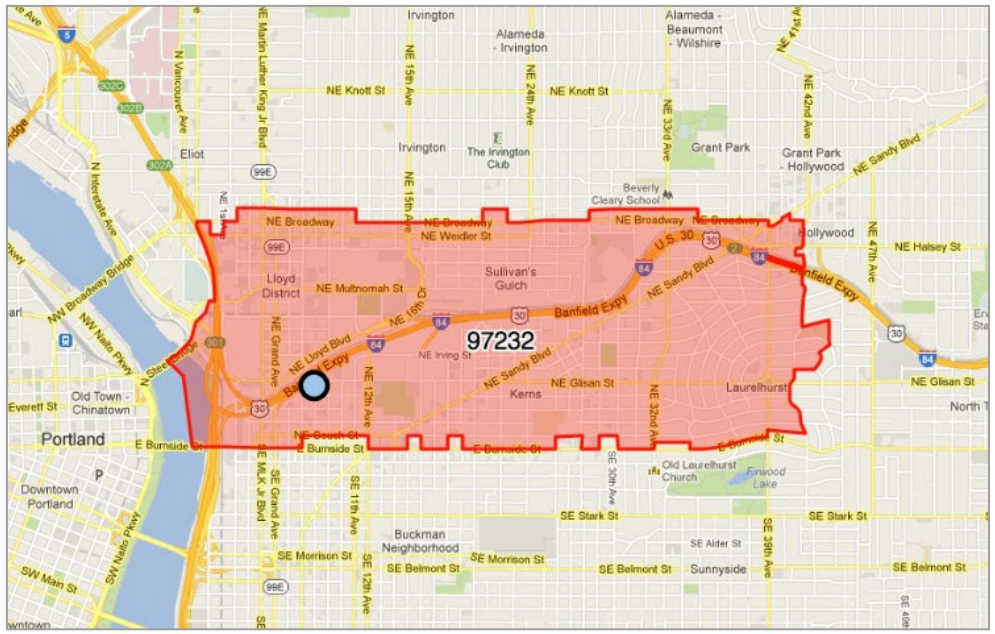
- Query based on proximity.
- “What is the closest McDonald’s?”

```
function findPlaces(query, scene) {
    for(var i=0; i < scene.places.length; i++) {
        var place = scene.places[i];
        var dist = satisfyQuery(query, place);
        // ...
    }
}
```

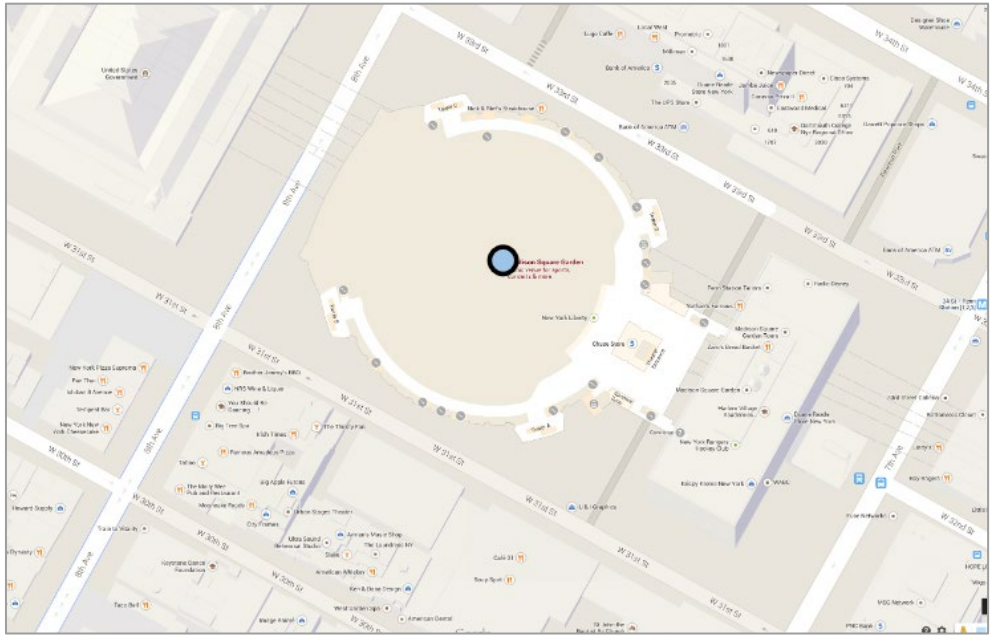
- Complexity: $O(n)$
- How to do better?



Point in polygon



What is the zip code for this complaint?



Am I inside a specific building?

Time complexity

- Ray-scene intersection: $O(n)$
- Proximity query: $O(n)$
- Point in polygon: $O(n)$

How to reduce the time complexity?

Motivation

- Expensive operations (ray tracing, query).
  - Complex scenes (millions of objects).
  - Large number of operations (hundreds of millions per second).
- Reduce complexity through pre-processing data.
  - Spatial data structures: structures of objects in space.
  - Eliminate candidates as early as possible.
  - Reduce complexity to $O(log\ n)$ on average.
  - Worst case complexity still $O(n)$.
    - Can you come up with a worst case example?

Spatial data structures

- Data structures to accelerate queries of the kind:

  “I’m here, which object is around me?”

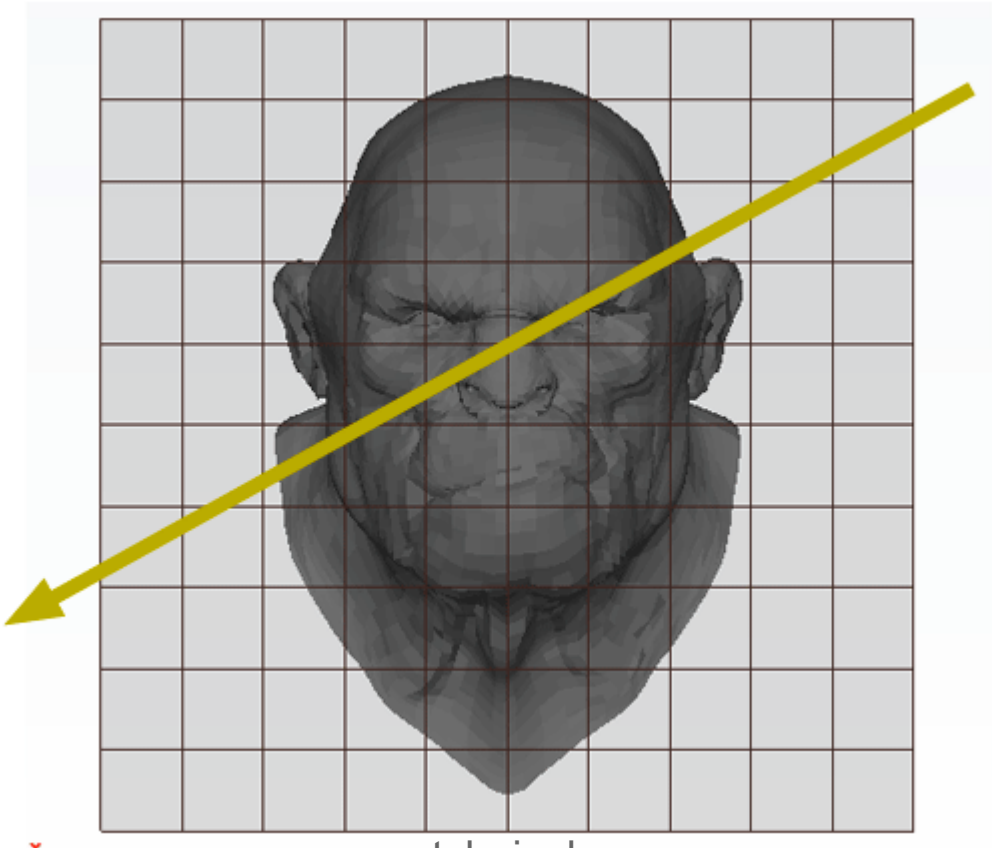
- Partition space or set of objects.
- Tasks:
  1. Construction / update:
     - Pre-processing for static parts of the scene.
     - Update for moving parts of the scene.
  2. Access:
     - Optimize so it is done as fast as possible.

Spatial data structures

- Uniform grid: 2D/3D data, uniform distribution.
- Quadtree: 2D data, non-uniform distribution.
- Octree: 3D data, non-uniform distribution.
- KD-tree: 2D/3D data, avoid empty cells.

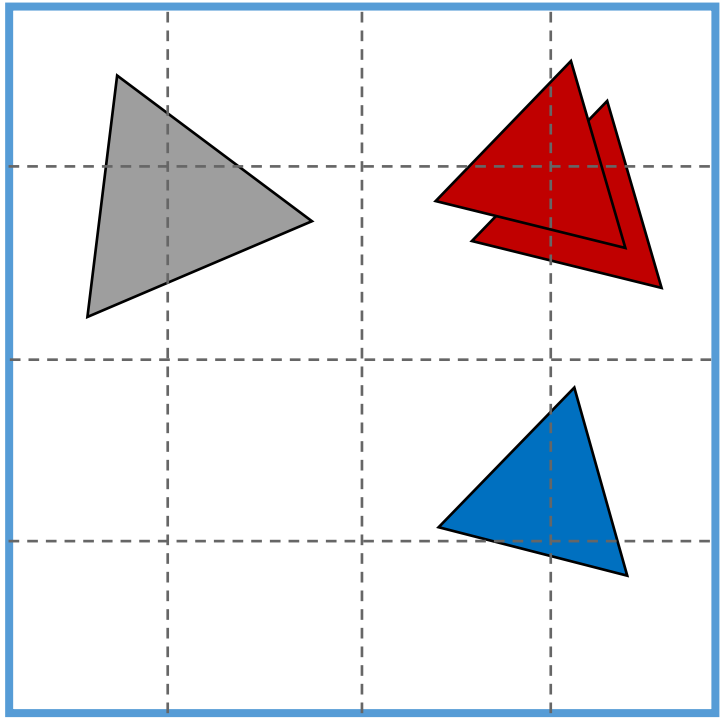
Uniform grid

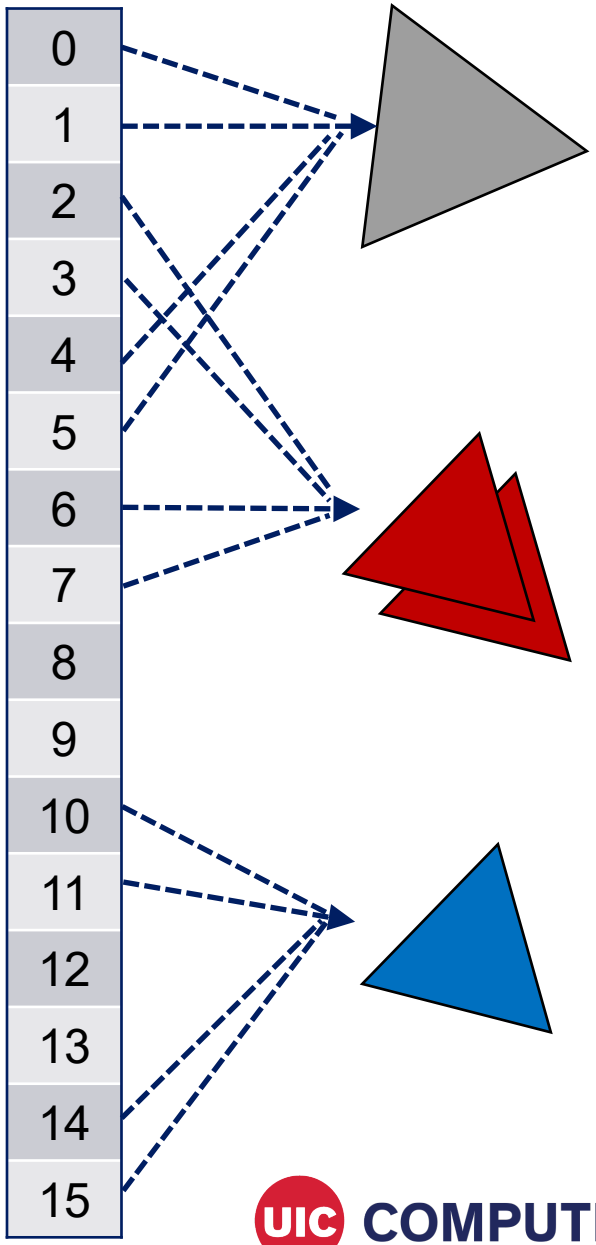
- Partition space into equal-sized volumes (i.e., voxels).
- Each cell will contain objects that overlap the voxel.
- Good for uniform data (points are evenly distributed in space).
- Fast construction and queries.



scratchpixel.com

Uniform grid





Uniform grid: construction and query

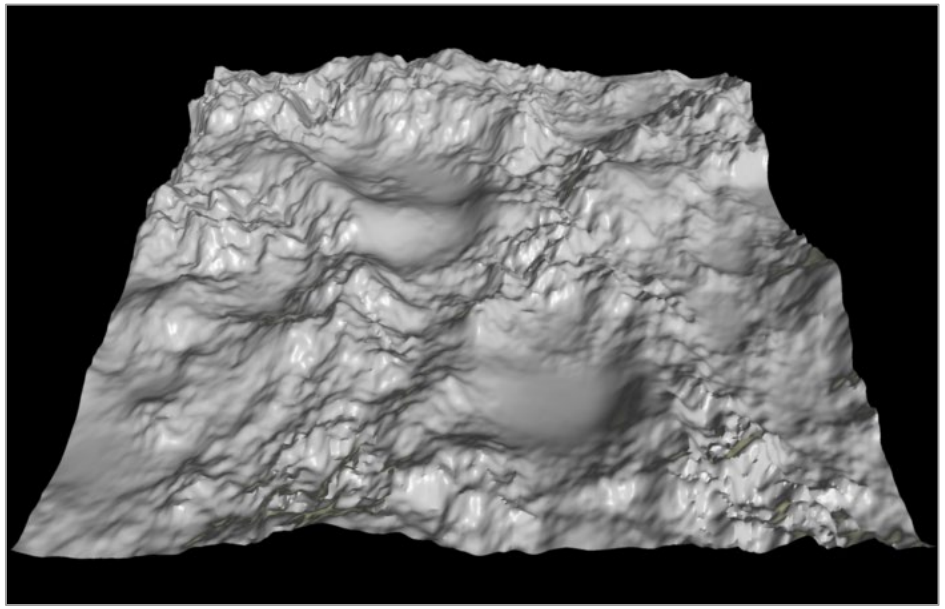
- Array of 3D voxels
  - Each voxel: list of pointers to colliding objects.
- Indexing function:
  - 3D point → cell index (constant time!)
- Construction:
  - Initialize cells for grid with size $w * h$
  - For each object $\mathrm{p}(x, y)$:
    - Compute grid cell using (x, y).
    - Store p in cell.
- Query:
  - For query rectangle $(x_1, y_1) \times (x_2, y_2)$:
    - Compute subgrid for (x_1, y_1) and (x_2, y_2).
    - For all cells inside subgrid, report all objects.
    - For all cells on the border of the subgrid, test objects against rectangle.

Uniform grid: complexity

- Build time: $O(n)$
- Space: $O(w * h) + O(n)$
- Query: $O(k)$

Uniform grid: complexity

- When uniform grids work well? Uniform distribution of objects.

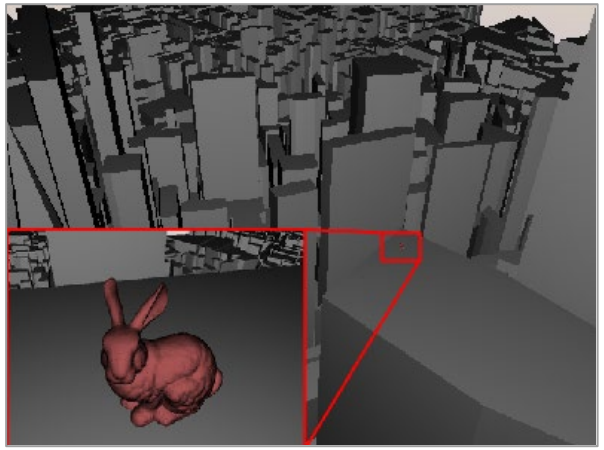


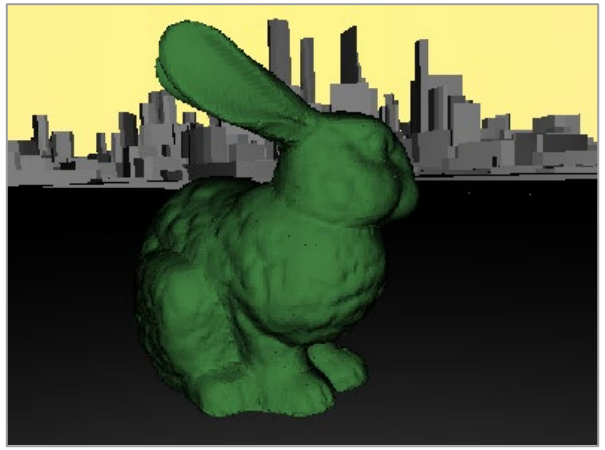
Mitsuba renderer

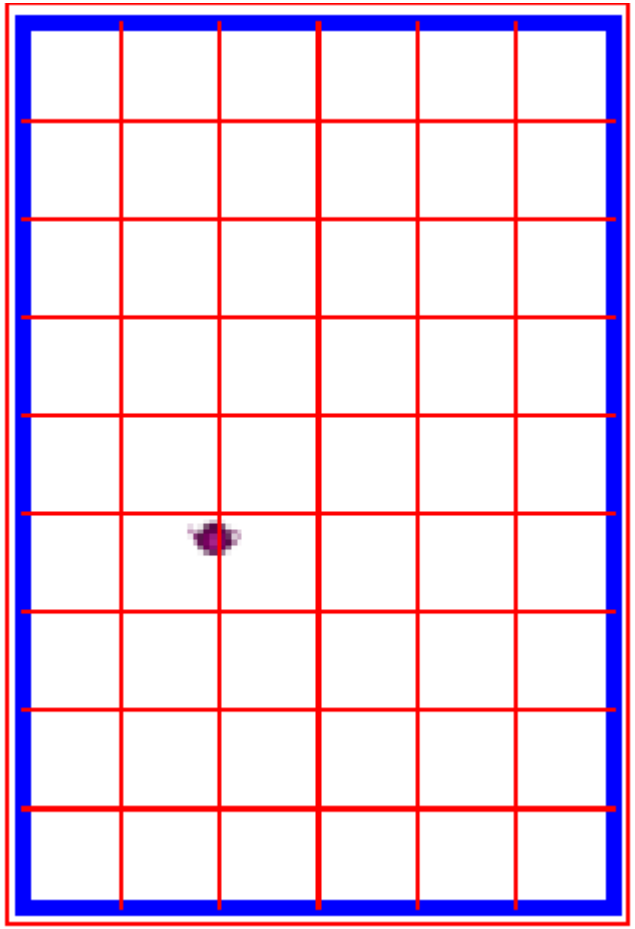
peterguthrie.net

Uniform grid: drawbacks

- When uniform grids do not perform well? Non-uniform distribution of objects.
- “Teapot in a stadium” problem: uniform grids cannot adapt to local density of objects.





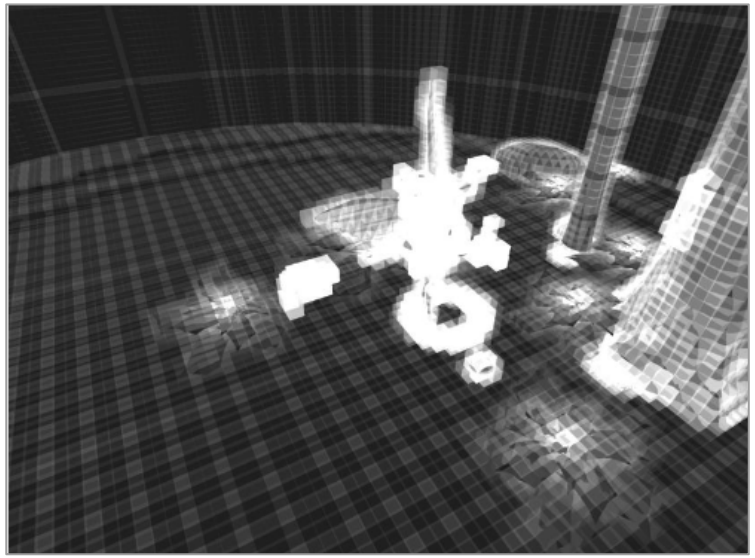


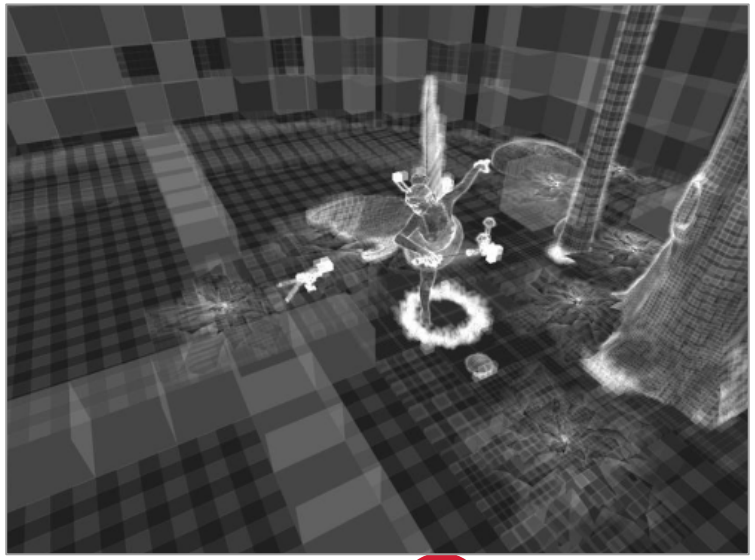
Uniform grid: drawbacks

- Assumes objects uniformly distributed in space.
- What happens when assumption does not hold?
  - Many empty cells.
  - Few cells with too many points.
- Change cell size?
  - Too small: memory occupancy too large.
  - Too big: too many objects in one cell.

Nested grids

- Possible solution to “teapot in a stadium” problem.
- Hierarchy of uniform grids: each cell is itself a grid.
- Fast building & traversal.

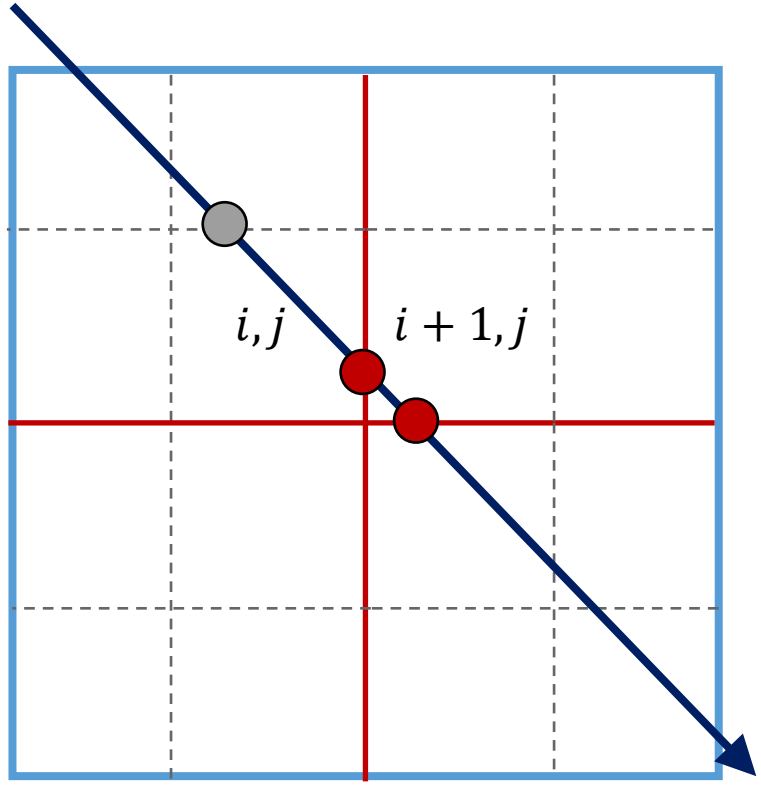




Philipp Slusallek

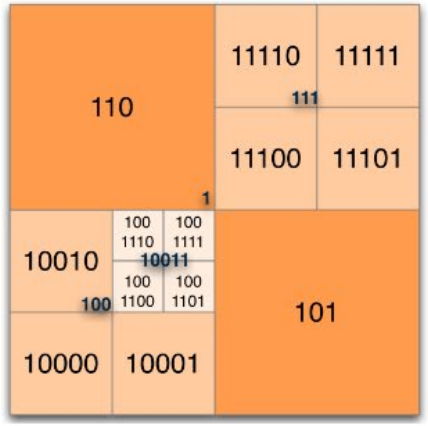
Incremental traversal

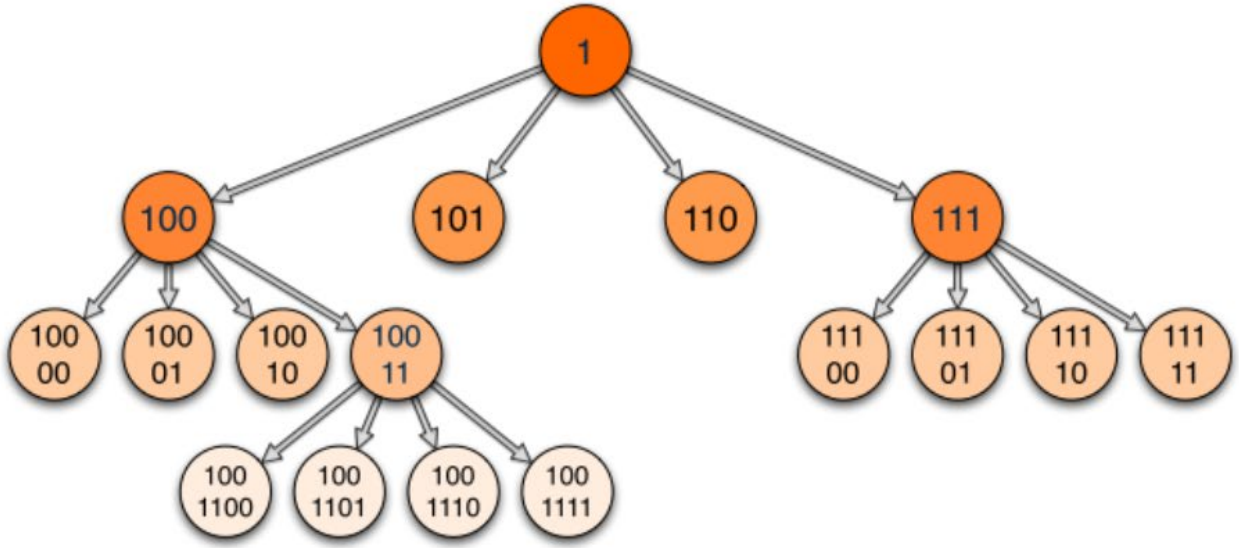
- How to traverse the voxels?
- Incremental traversal similar to line rasterization:
  - From current intersection P on cell (i, j), perform ray-plane intersection tests with the next 2 grid planes along the ray direction.
  - Next intersection point is the nearest one among the 2 candidates.
  - Update cell index according to new intersection point.
  - Repeat process until the ray intersects with a primitive or reaches the boundary of the grid.



Quadtree

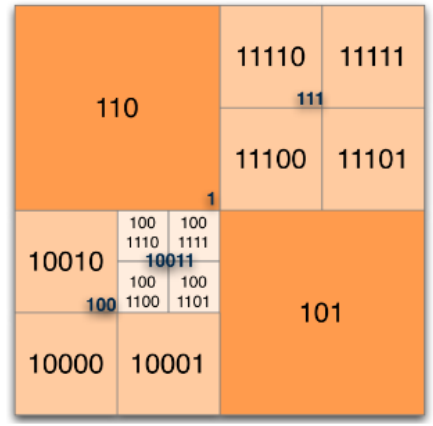
- Hierarchical structure that stores regular grids at each level.
- Adaptive subdivision: adjust depth to local scene complexity.

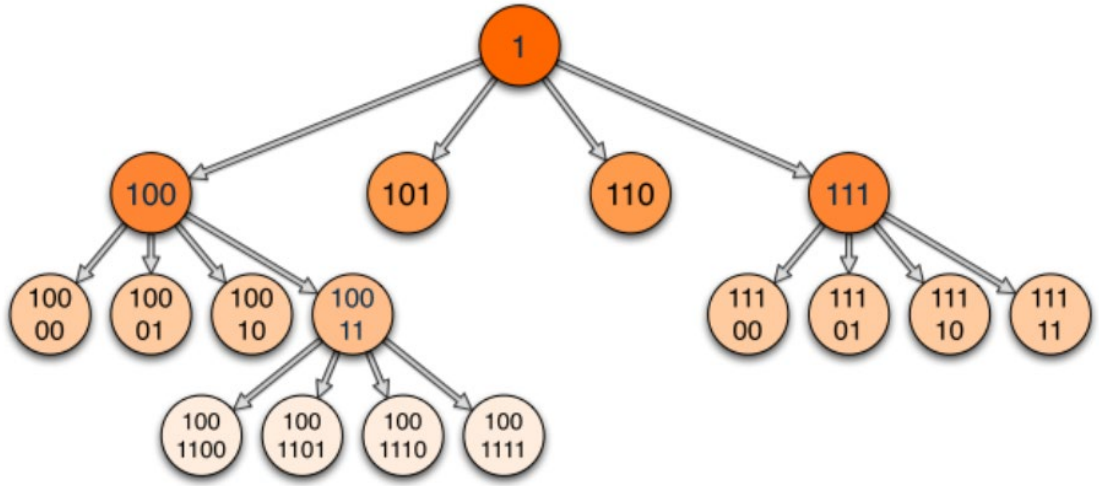




Quadtree

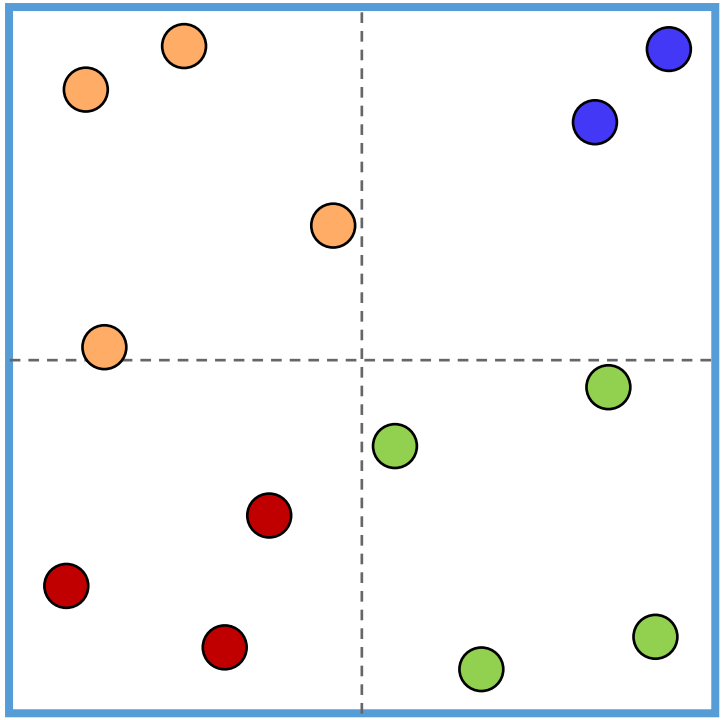
- Rooted tree in which every internal node has four children.
- Every node corresponds to a square.
- Tree: branching factor 4 or 8.
- Each node: splits into all dimensions at once (in the middle).
- Construction: continue splitting until end nodes have few objects (or limit level reached).





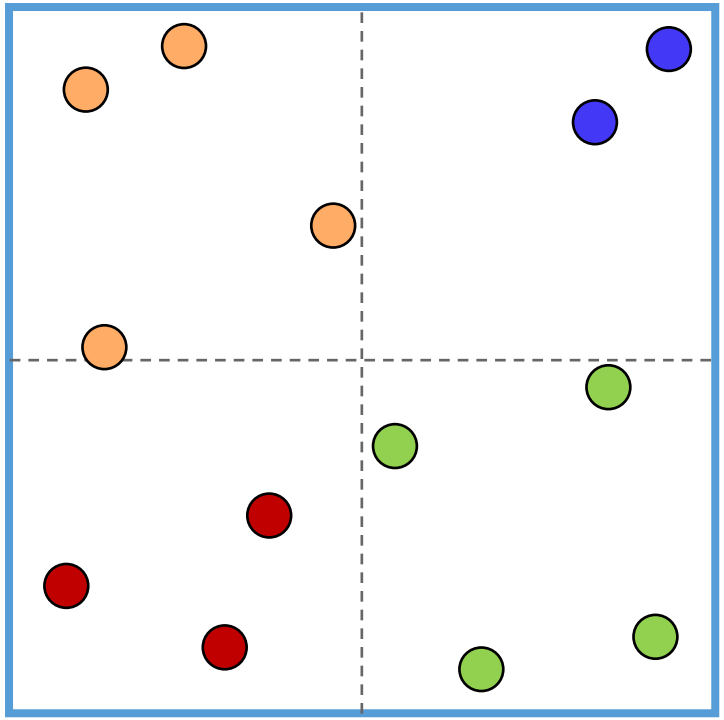
Quadtree: construction

- Split the top level.



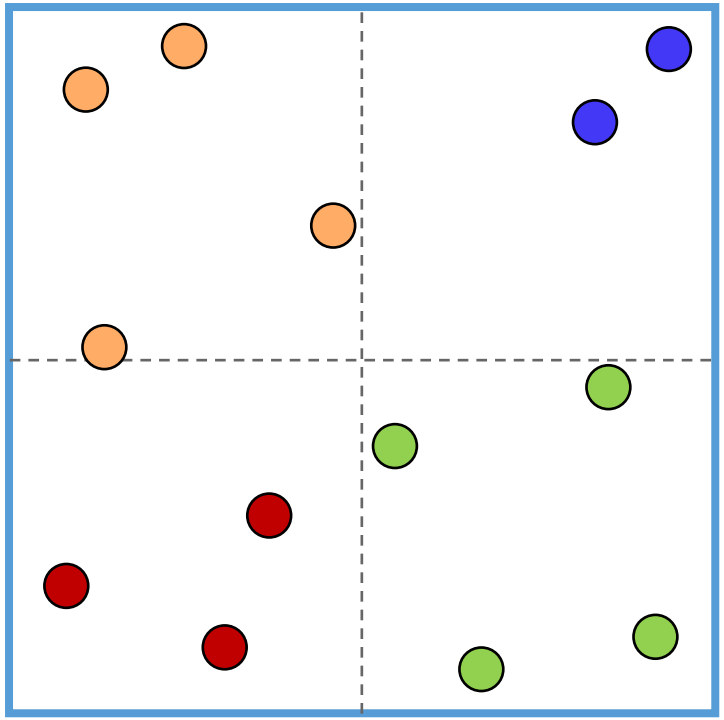
Quadtree: construction

- Split the top level.
- Can we stop?



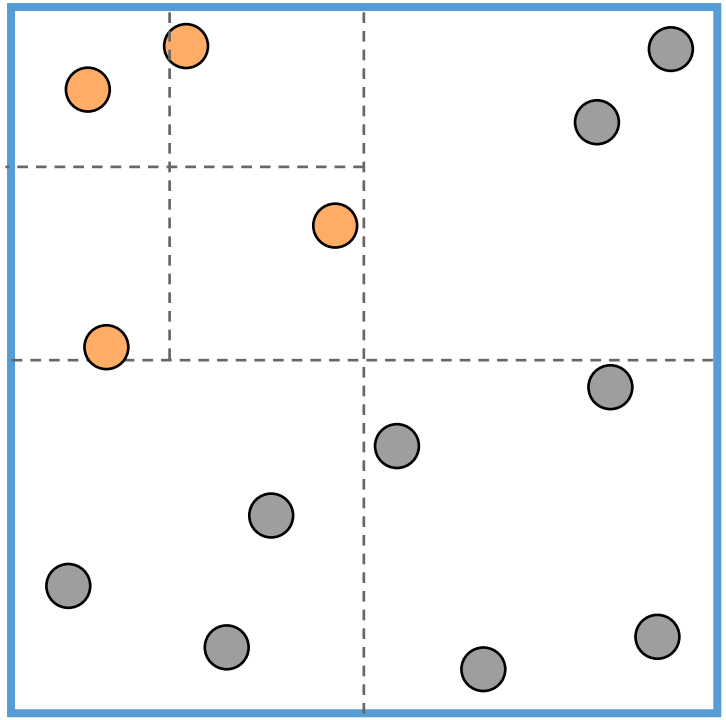
Quadtree: construction

- Split the top level.
- Can we stop? No, split the next level.



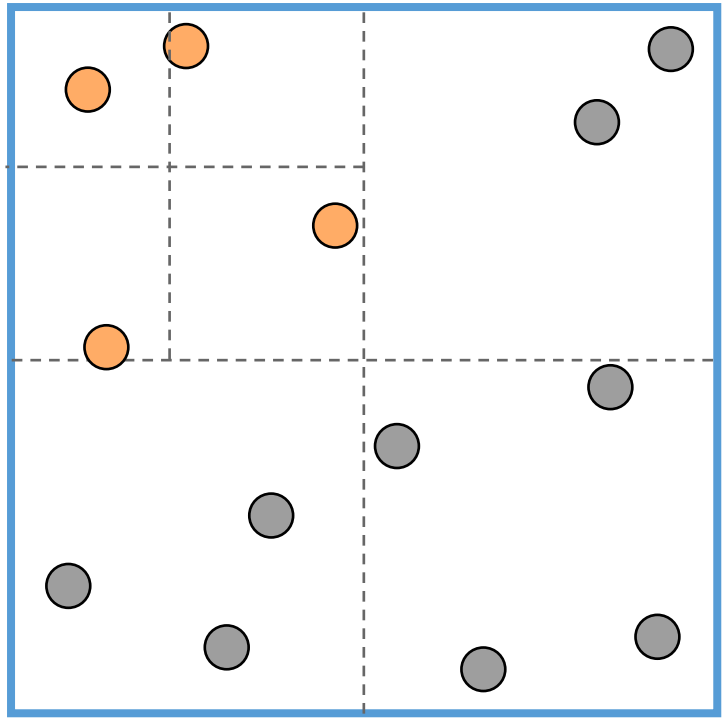
Quadtree: construction

- Split the top level.
- Can we stop? No, split the next level.
- Split top-left.



Quadtree: construction

- Split the top level.
- Can we stop? No, split the next level.
- Split top-left.
- Can we stop top-left?



Quadtree: construction

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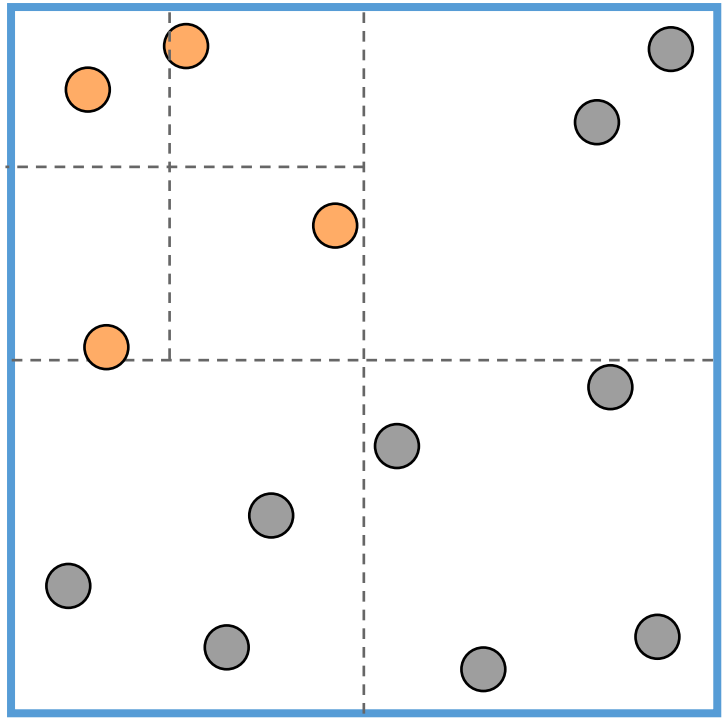
- Can we stop top-left? Yes.



Quadtree: construction

- Split the top level.
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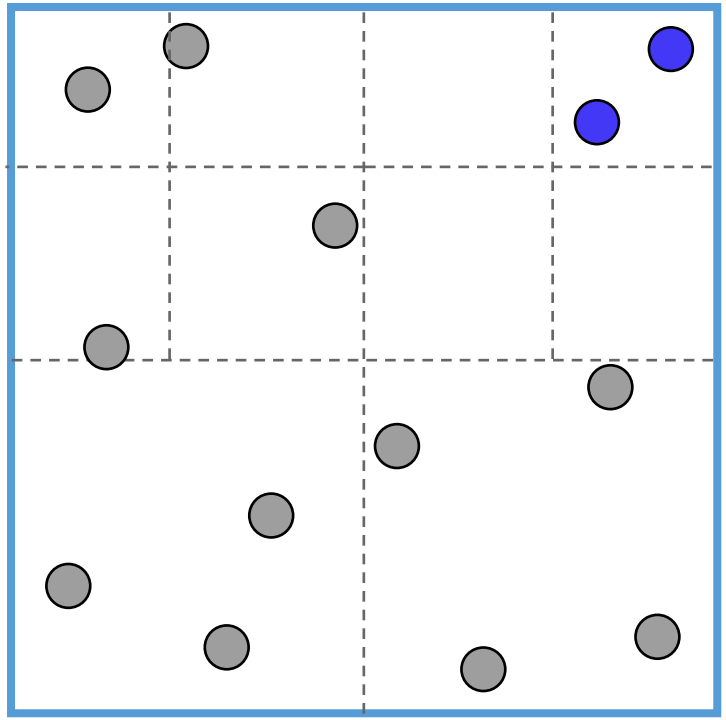
- Split top-left.
- Can we stop top-left? Yes.
- Split top-right.



Quadtree: construction

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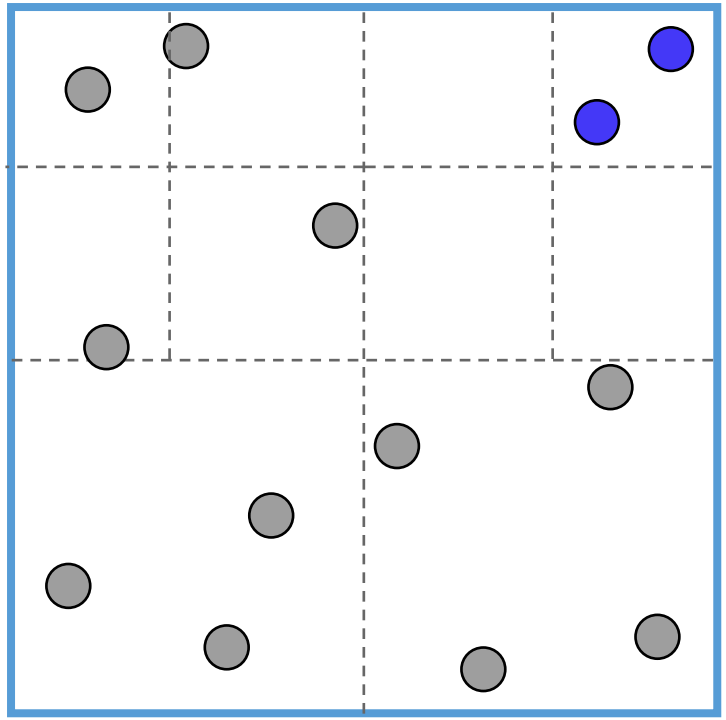
- Can we stop top-left? Yes.
- Split top-right.
- Can we stop top-right?



Quadtree: construction

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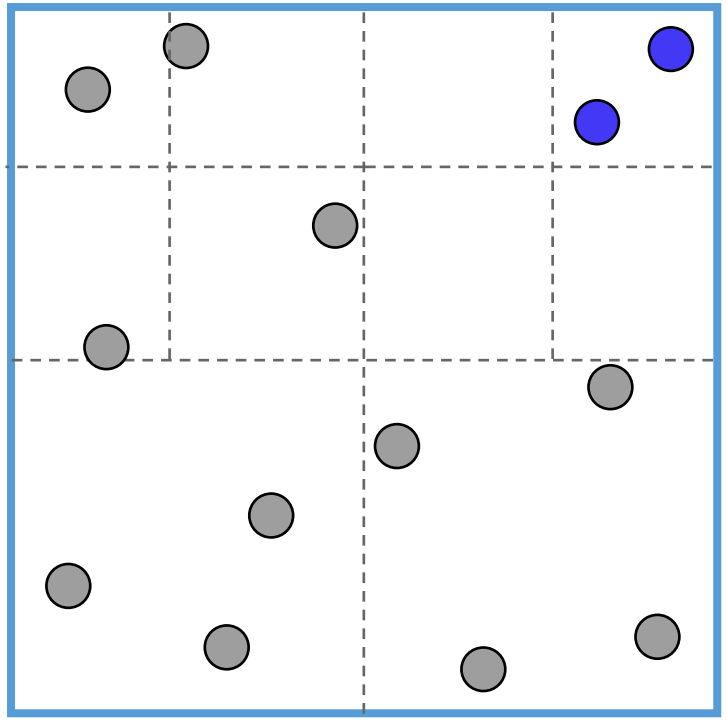

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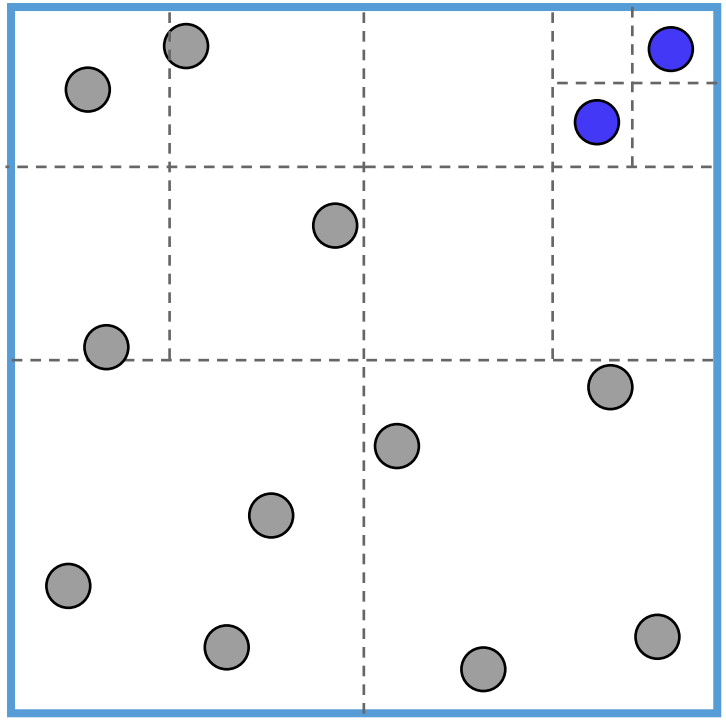

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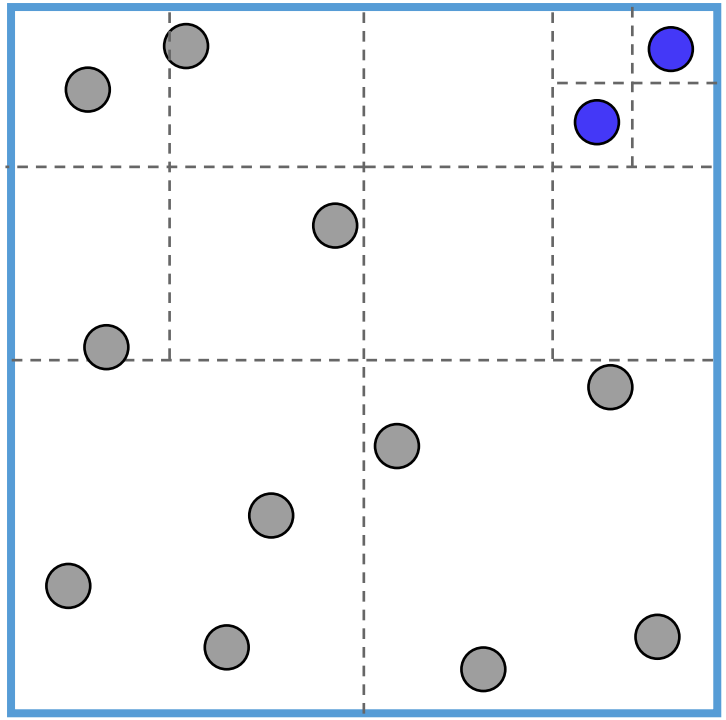

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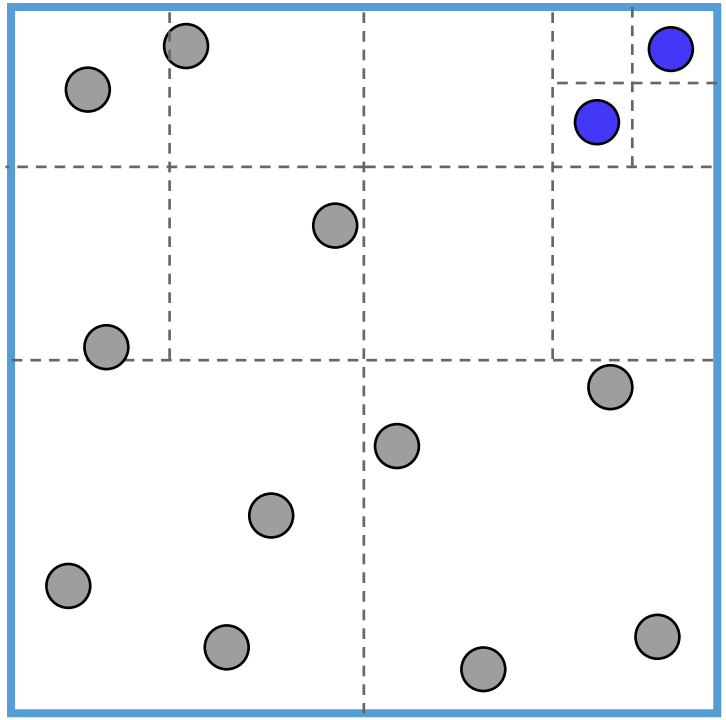

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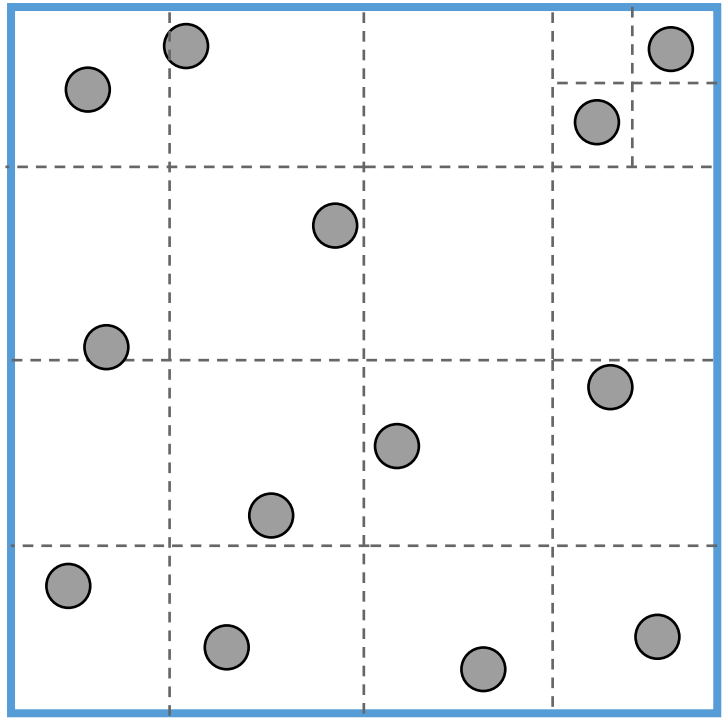
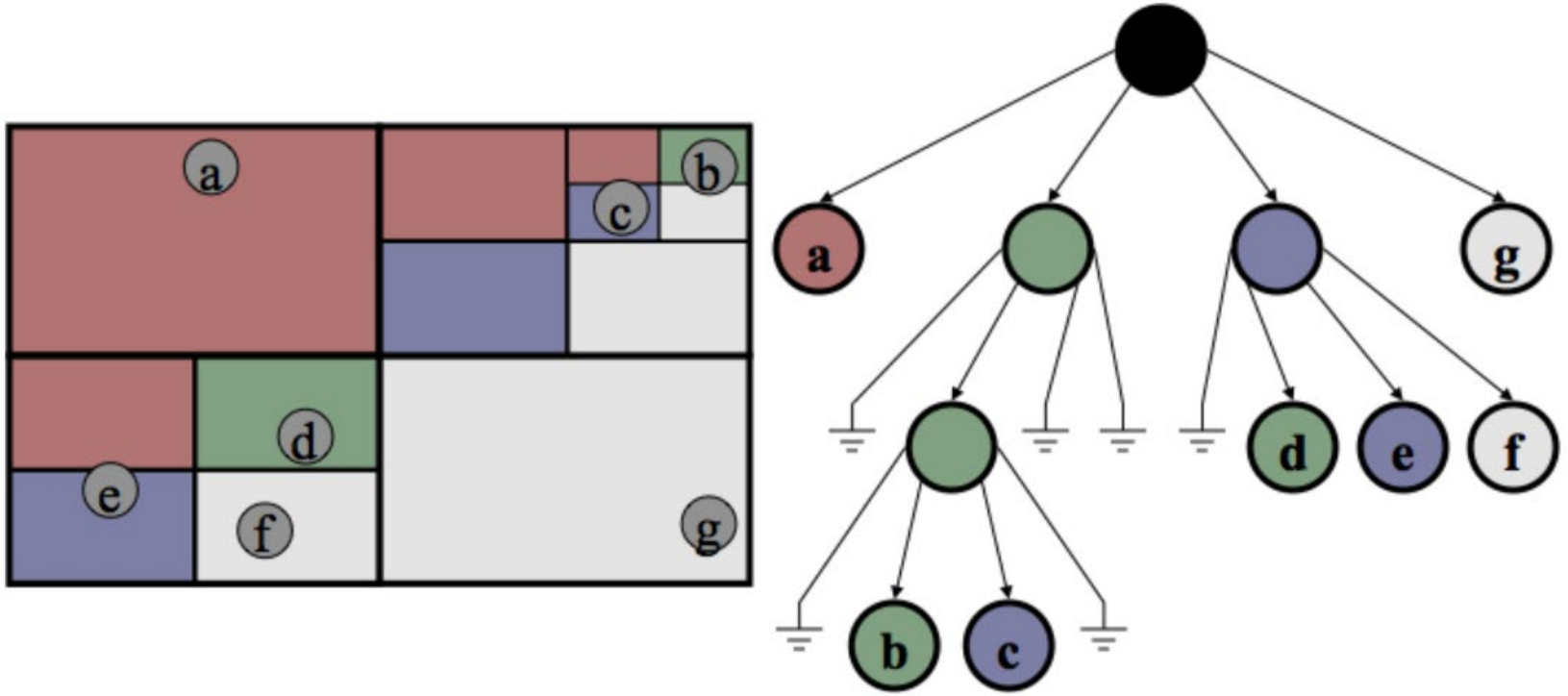


Quadtree: construction



Quadtree: construction

- Construction:
  - Input: set of objects P inside a square S $(x_1, y_1) \times (x_2, y_2)$, tree node v
  - If $|P| \leq 1$:
    - Quadtree consists of a single leaf with P.
  - Else:
    - P_{00}: set of points that fall in the bottom-left corner of S.
    - P_{01}: set of points that fall in the bottom-right corner of S.
    - …
    - v_{00}: node with points of P_{00}.
    - v_{01}: node with points of P_{01}.
    - …
    - Append v_{00}, v_{01}, v_{10}, v_{11} to v.

Quadtree: query

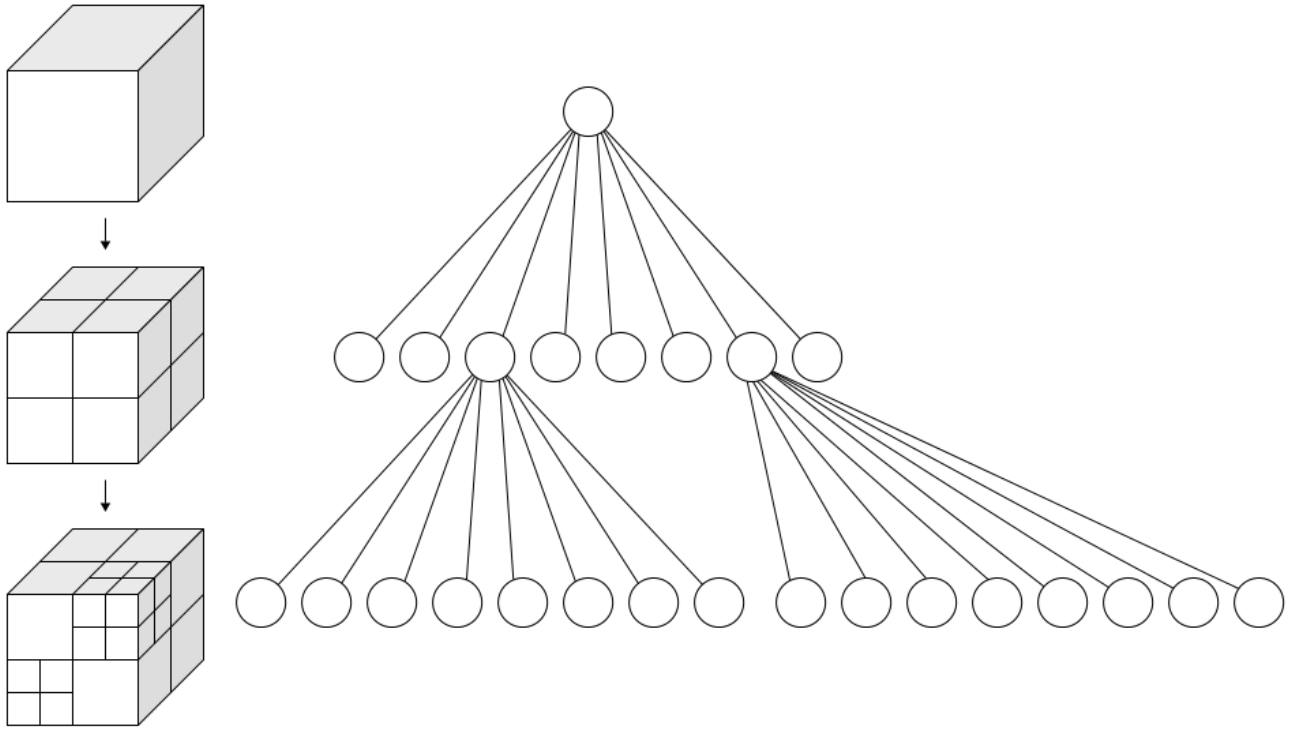
- Query:
  - Input: range query r $(x_1, y_1) \times (x_2, y_2)$, tree node v.
  - If v is a leaf:
    - Search points of v inside range r.
  - If v_{00} inside range r :
    - Query(v_{00}, r)
  - If v_{01} inside range r :
    - Query(v_{01}, r)
  - If v_{10} inside range r :
    - Query(v_{10}, r)
  - If v_{11} inside range r :
    - Query(v_{11}, r)

Quadtree: complexity

- Build time: $O(n)$
- Space: $O(n)$
- Range query: $O(\sqrt{n} + k)$
- Leaf traversal: $O(logn)$

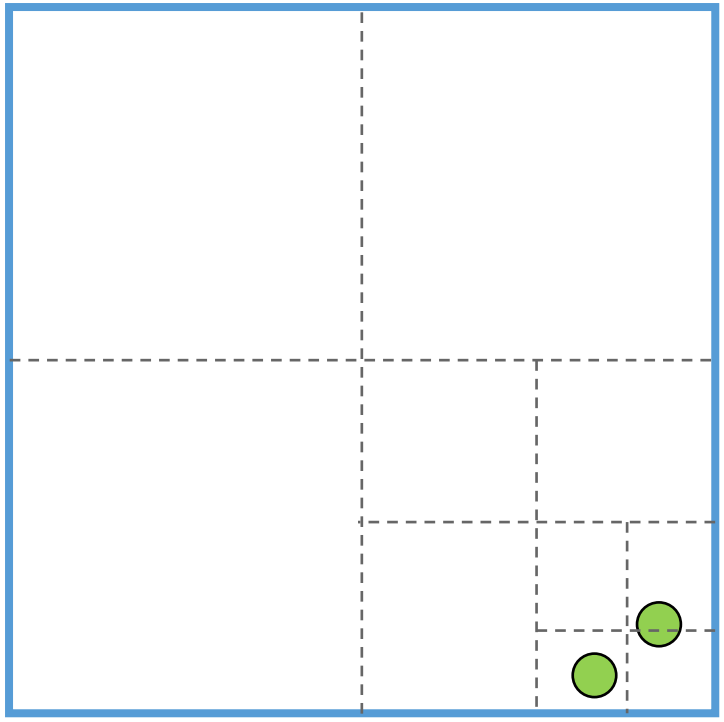
Octree

- Each inner node contains 8 equally sized voxels.
- A 3D quadtree.



Quadtree and octree: drawbacks

- Grater ability to adapt to location of scene geometry than uniform grid.
- But very long tree to store points that are concentrated in a small region.
- Many nodes will contain zero objects.

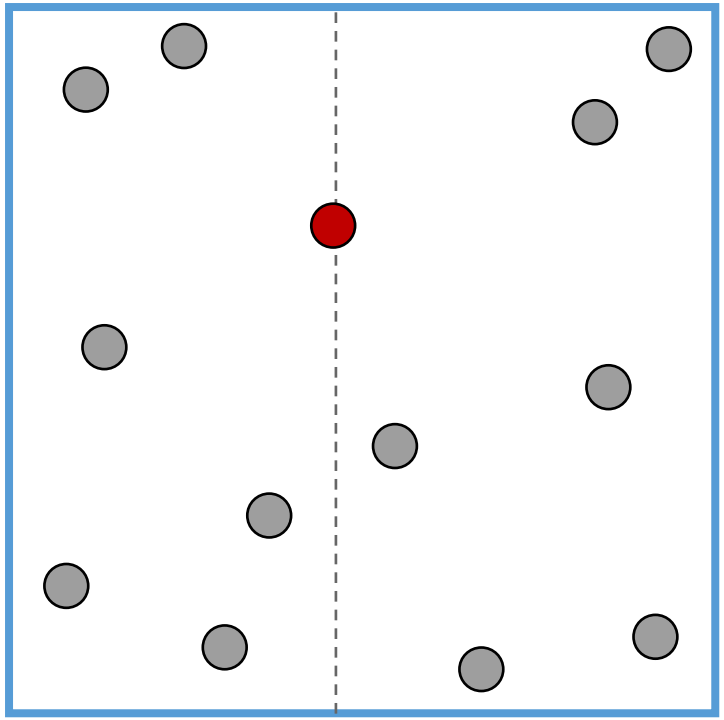


K-d tree

- Differently from quadtrees and octrees, k-d trees only split **one** dimension at each level.
- Where to split? Middle? Median? Proportional to surface area?
- At each level:
  - Quadtree creates 4 equal sized cells.
  - Octree creates 8 equal sized cells.
  - K-d tree creates 2 non-equal sized cells (2D case).

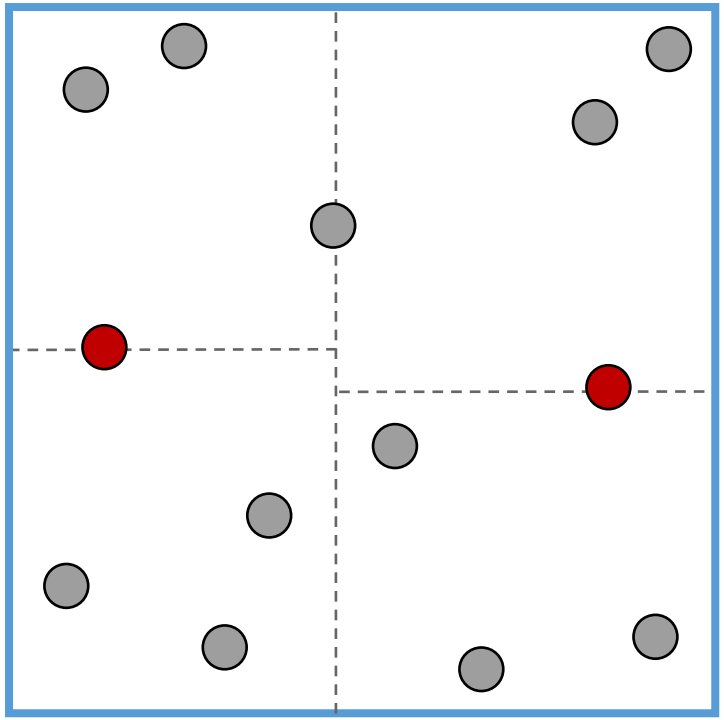
K-d tree: construction

- First split: x dimension (median point).



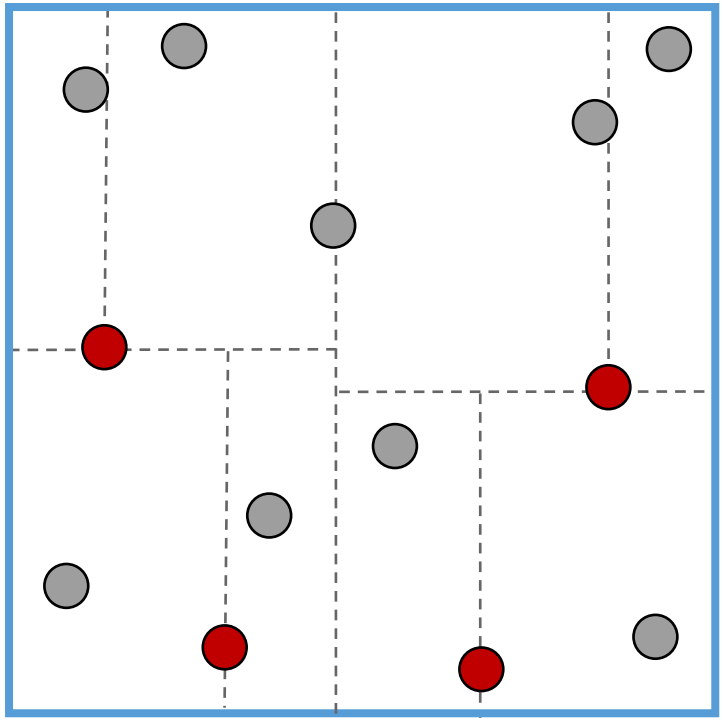
K-d tree: construction

- First split: x dimension (median point).
- Second split: y dimension.



K-d tree: construction

- First split: x dimension (median point).
- Second split: y dimension.
- Repeat, alternating split dimensions



K-d tree: construction

- Construction:
  - Input: set of objects P inside a square S $(x_1, y_1) \times (x_2, y_2)$, tree node v
  - If $|P| \leq 1$:
    - K-d tree consists of a single leaf with P.
  - Else:
    - If depth is even:
      - Split P into P_0 and P_1, along a vertical line through the y axis.
    - Else:
      - Split P into P_0 and P_1, along a vertical line through the x axis.
    - v_0: $build(v, P_0, depth + 1)$.
    - v_1: $build(v, P_1, depth + 1)$.
    - …
    - Append v_0, v_1 to v.

K-d tree: query

- Query:
  - Input: range query r $(x_1, y_1) \times (x_2, y_2)$, tree node v.
  - If v is a leaf:
    - Search points of v inside range r.
  - If v_0 inside range r :
    - Query(v_0, r)
  - If v_1 inside range r :
    - Query(v_1, r)

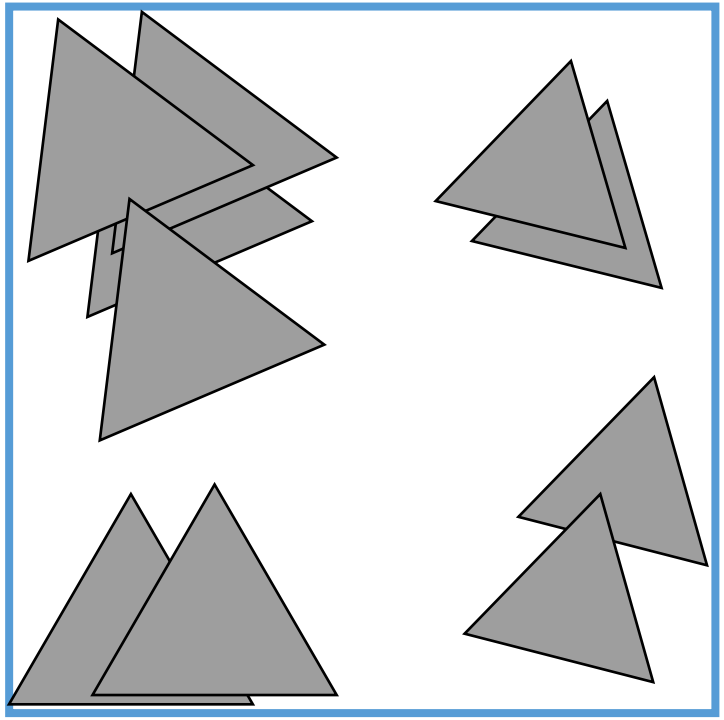
K-d tree: complexity

- Build time: $O(nlogn)$
- Space: $O(n)$
- Range query: $O(\sqrt{n}+k)$
- Leaf traversal: $O(logn)$

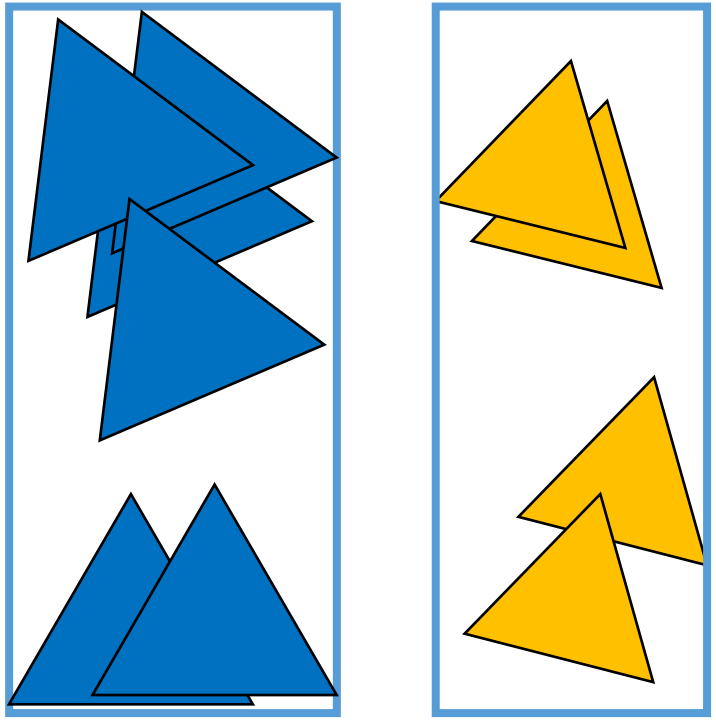
Bounding volume hierarchies (BVH)

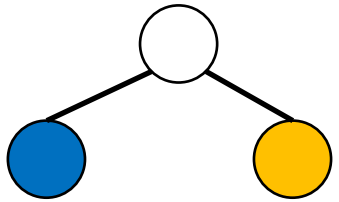
- Goal: use the scene hierarchy given by the scene graph (instead of a spatial derived one).
- Associate a bounding volume to each node.
  - A bounding volume of a node bounds **all** objects in the subtree.
  - Nodes are aggregated objects.
- Construction and update is fast.
  - Bottom-up: recursive.
- Querying it:
  - Top-down: visit.

Bounding volume hierarchies (BVH)

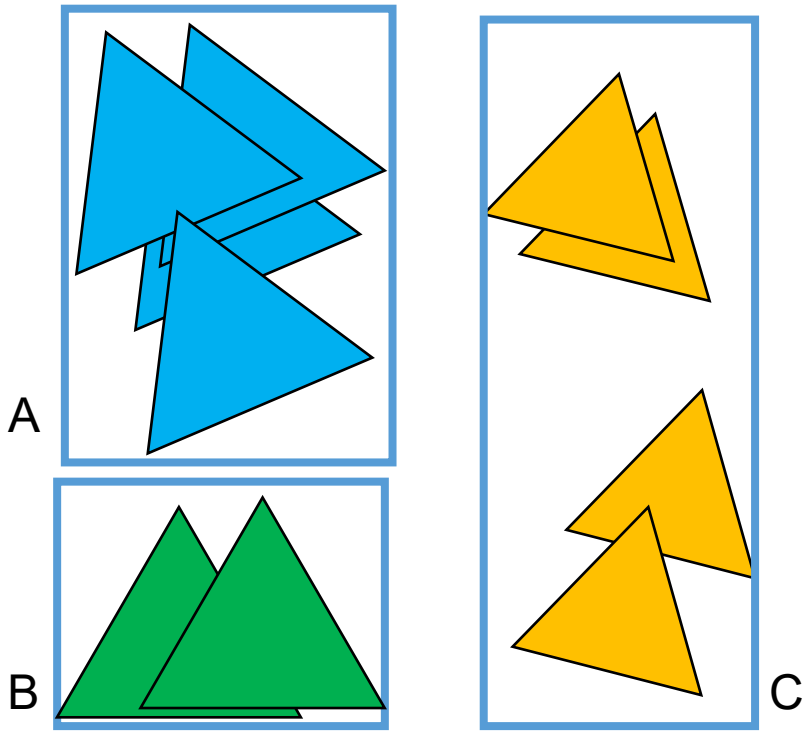


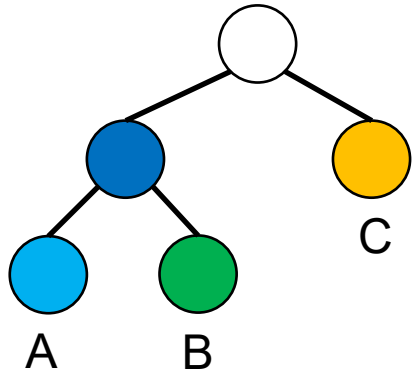
Bounding volume hierarchies (BVH)





Bounding volume hierarchies (BVH)





- Leaf nodes: contain *small* list of primitives.
- Interior nodes:
  - Proxy for a large subset of primitives.
  - Stores bounding box for all primitives in subtree.

Bounding volume hierarchies (BVH)

- How to split the scene?
- Primitive list of current node P is sorted based on the centroids of primitive bounding boxes. Ordered list is split into P_0 and P_1.
- Different heuristics:
  - Edge volume heuristic (EVH).
  - Split bounding volume hierarchy (SBVH).
  - Surface area heuristic (SAH):

$$C = C_t + \frac{SA(B_1)}{SA(B)}|P_1|C_i + \frac{SA(B_2)}{SA(B)}|P_2|C_i$$

Bounding volume hierarchies (BVH)

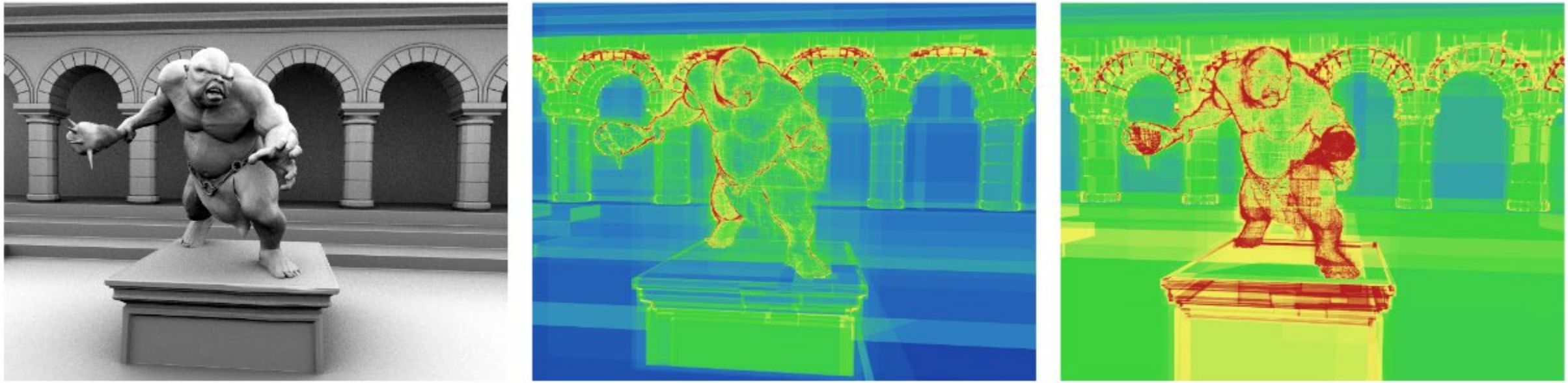


Figure 1: *Sample scene consisting of roughly 1.9 million triangles (left). Our method (middle) results in a significant reduction of ray shooting costs compared to a regular bounding volume hierarchy (right). The heat views visualize the summed number of traversal steps and primitive intersections for primary rays.*

[Stich et al., 2009]

Object vs. space partitioning

- Object partitioning:
  - Hierarchically partition **objects** into groups.
  - Spatial index is created by spatially bounding each subgroup.
- Space partitioning:
  - Hierarchically partition **space** into subspaces.
  - Subspaces are non-overlapping and completely fill parent space.
  - Tree or table structure.

Summary

- Choose the right structure considering the operations and data.
- Uniform grid:
  - The most parallelizable (to update, construct, use).
  - Constant time access (best!).
  - Quadratic / cubic space (2D, 3D).
  - Good performance under uniform distribution of objects.
- Quadtree, octree, k-d tree:
  - Compact.
  - Simple.
  - Non-constant accessing time.
  - Good performance under non-uniform distribution of objects.
- BVH:
  - Simple construction.
  - Ideal for dynamic parts of the scene.
  - Not necessarily very efficient to access.
    - May need to traverse multiple children.
    - If you do not have a scene-graph you need to create one.

Voxel-based rendering

- Outcast, 1999.
- *“At this point Yves started wondering if dedicated hardware would ever catch-up the exponential acceleration of CPU power and flexibility”.*
- Flexibility to program effects not available on GPUs at the time:
  - Ripple effects
  - Real-time shadows
  - Curved surfaces

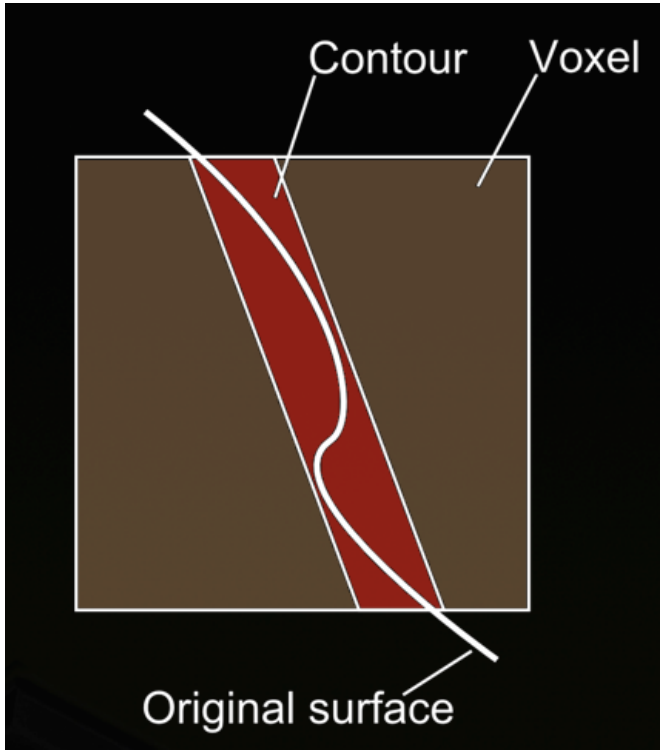
http://francksauer.com/index.php/games/test/15-games/published-games/47-outcast-pc

Sparse voxel octrees

- Using voxel representations as a generic way for expressing complex and feature-rich geometry.
- Voxels as a representation of opaque surfaces.
  - Fine-grained resolution control
  - Compact representation
  - Easy to downsample
- Efficient sparse voxel octrees avoid building the octree with maximum depth by providing each voxel with a contour.
  - If contours are good approximation to the original geometry, subdivision is stopped.

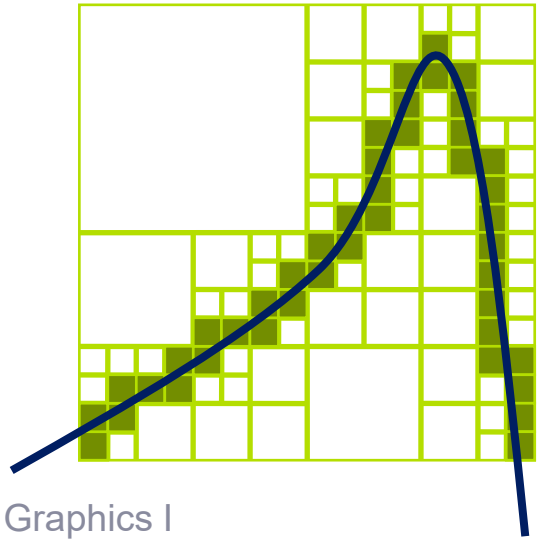
Sparse voxel octrees: contours

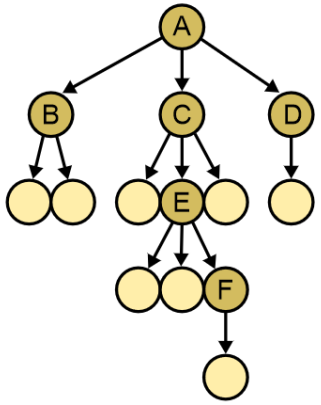
- Contours are slabs that modify voxel's shape.
- Improve accuracy for representing surfaces.
- Simple for rendering.

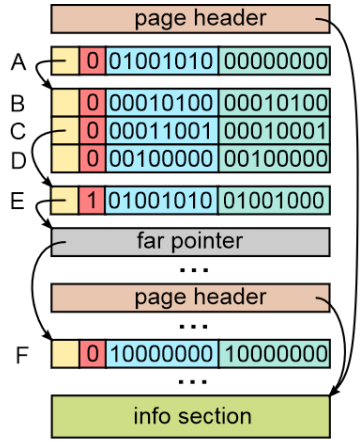


Efficient sparse voxel octrees

- How to efficiently store octrees?
  - One child pointer per voxel.
  - Efficient encoding of empty regions.
  - Culling away nodes that correspond to empty space.

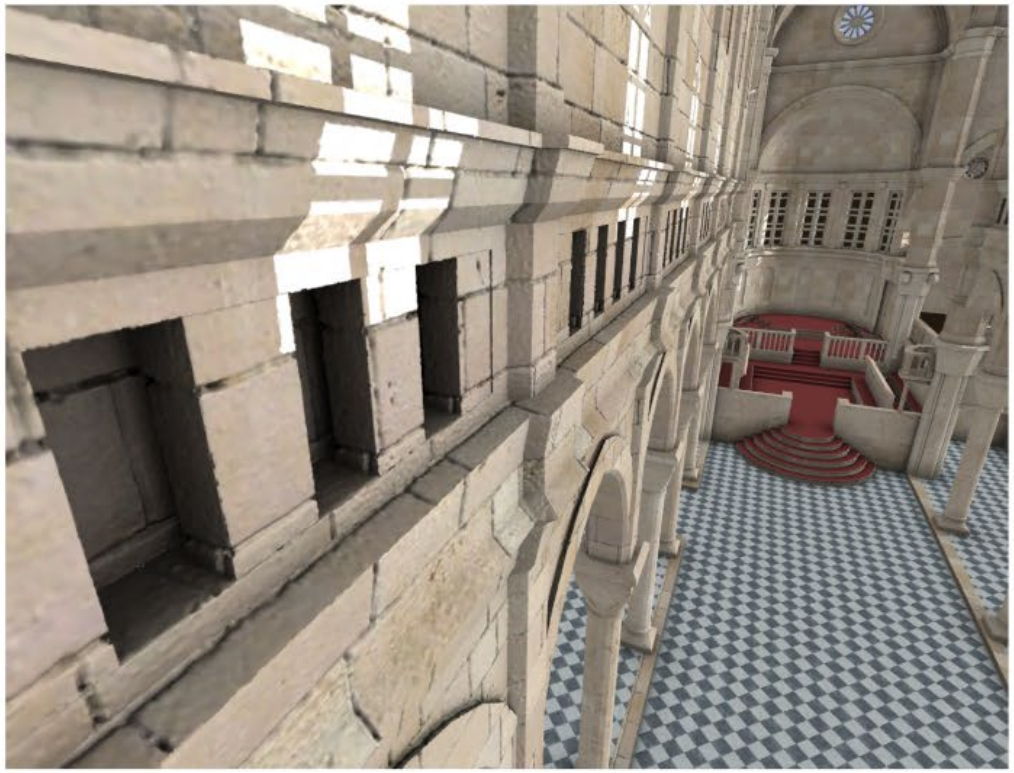


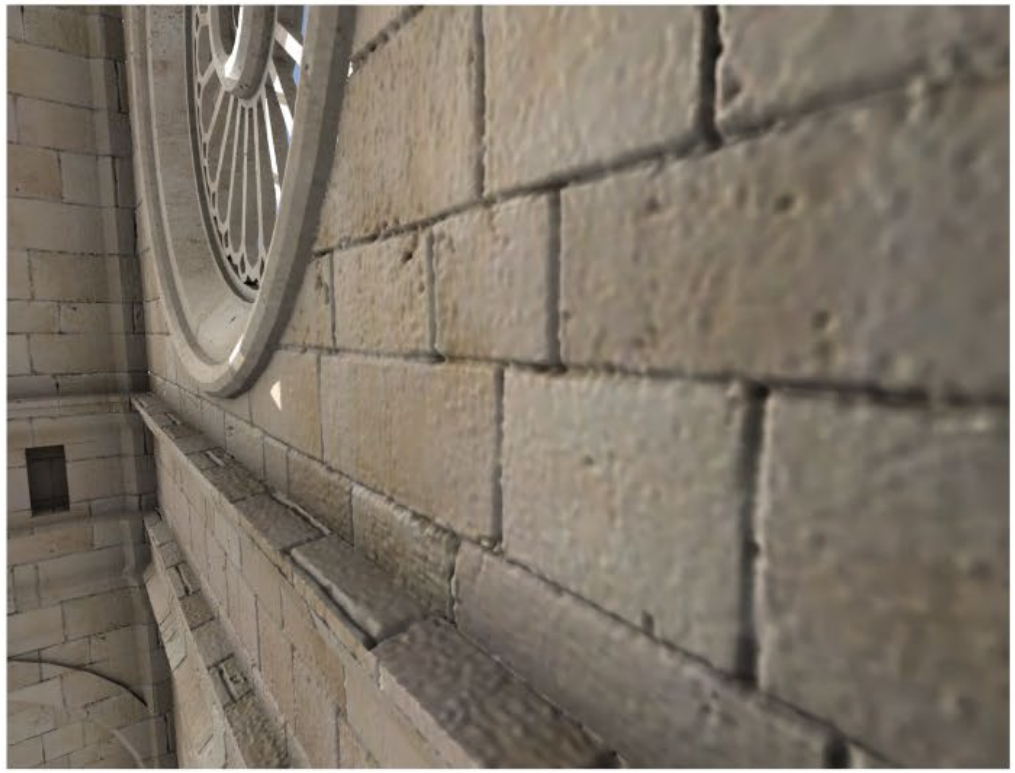




Laine and Karras, 2010

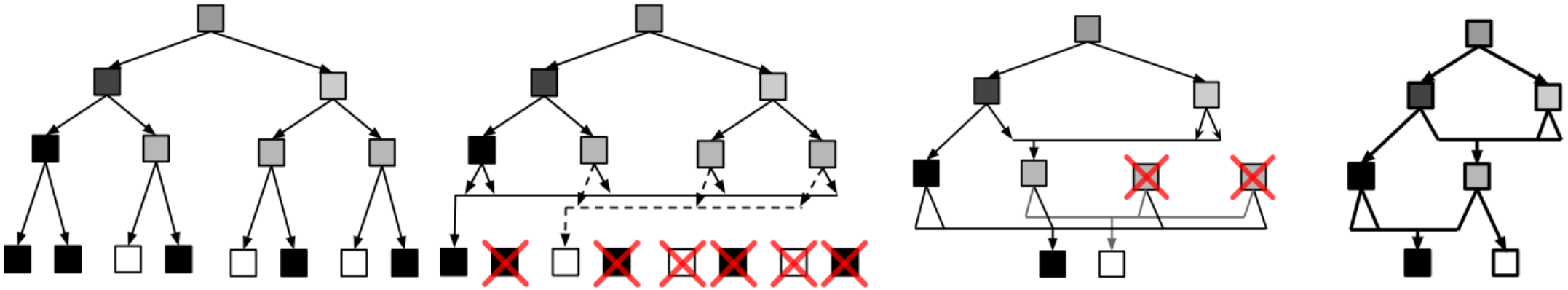
Efficient sparse voxel octrees





High resolution sparse voxel DAGs

- Binary voxel grid – generalization of octree to a directed acyclic graph.
- Assumption that SVO contains redundant subtress.
  - Searches the tree for common subtrees and only reference unique instances.
- DAG representation decreases memory consumption by up to 38x.



High resolution sparse voxel DAGs

- SVO to DAG:
  - Top down approach:
    - Start at the root node and test whether its children correspond to identical subtrees, and proceed down recursively.
    - Expensive.
  - Bottom up approach:
    - Leaf nodes are uniquely defined by their childmasks (8-bit mask where bit i tells us if child i contains geometry). Then at most $2^8 = 256$ unique leaf nodes in an SVO.
    - First step: merge identical leaves.
    - Proceed to level above, identifying nodes that have identical childmasks and identical pointers (roots of identical subtrees that can be merged).
    - Iteratively perform previous step.

High resolution sparse voxel DAGs

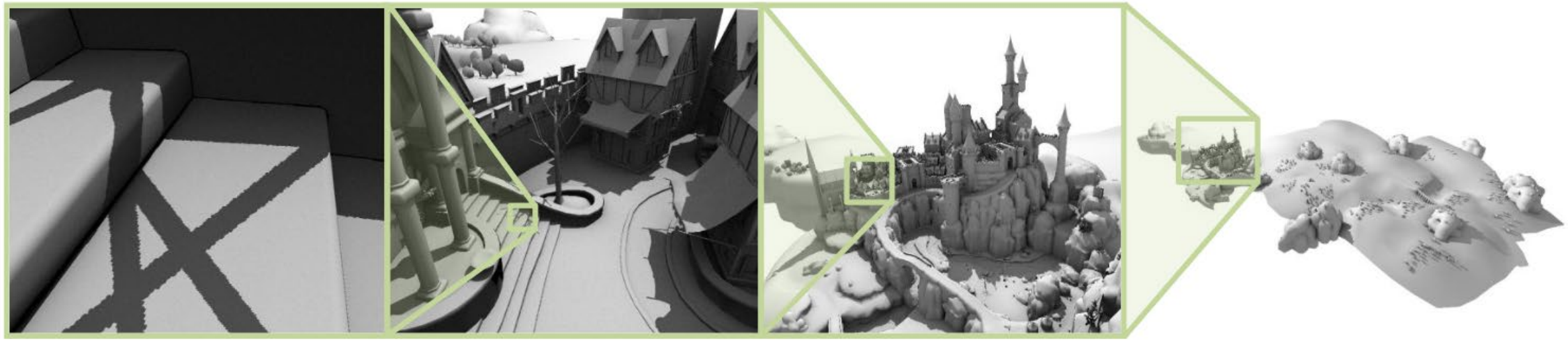


Figure 1: *The* EPICCITADEL *scene voxelized to a* $128K^3$ *(*$131\,072^3$*) resolution and stored as a Sparse Voxel DAG. Total voxel count is 19 billion, which requires 945MB of GPU memory. A sparse voxel octree would require 5.1GB without counting pointers. Primary shading is from triangle rasterization, while ambient occlusion and shadows are raytraced in the sparse voxel DAG at 170 MRays/sec and 240 MRays/sec respectively, on an NVIDIA GTX680.*

Kampe et al., 2013