Computer Graphics

- Spatial Index Structures -

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Motivation

- Tracing rays in O(n) is too expensive
 - Need hundreds of millions rays per second
 - Scenes consist of millions of triangles
- Reduce complexity through pre-sorting data
 - Spatial index structures
 - Dictionaries of objects in 3D space
 - Eliminate intersection candidates as early as possible
 - Can reduce complexity to O(log n) on average
 - Worst case complexity is still O(n)
 - Private exercise: Come up with a worst case example

Acceleration Strategies

Faster ray-primitive intersection algorithms

Does not reduce complexity, "only" a constant factor (but relevant!)

Less intersection candidates

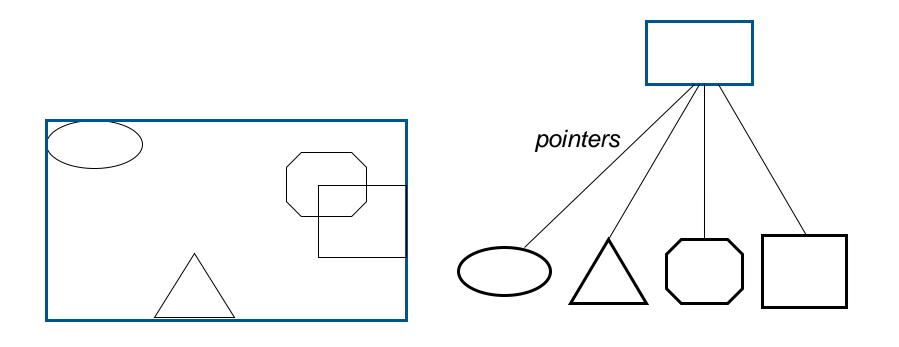
- Spatial indexing structures
- (Hierarchically) partition space or the set of objects
- Examples
 - · Grids, hierarchies of grids
 - Octrees
 - Binary space partitions (BSP) or kd-trees
 - Bounding volume hierarchies (BVH)
- Directional partitioning (not very useful)
- 5D partitioning (space and direction, once a big hype)
 - Close to pre-compute visibility for all points and all directions

Tracing of continuous bundles of rays

- Exploits coherence of neighboring rays, amortize cost among them
 - Frustum tracing, cone tracing, beam tracing, ...

Aggregate Objects

- Object that holds groups of objects
- Conceptually stores bounding box and list of children
- Useful for instancing (placing collection of objects repeatedly) and for Bounding Volume Hierarchies



Bounding Volumes

Observation

- BVs (tightly) bound geometry, ray must intersect BV first
- Only compute intersection if ray hits BV

Sphere

- Very fast intersection computation
- Often inefficient because too large

Axis-aligned bounding box (AABB)

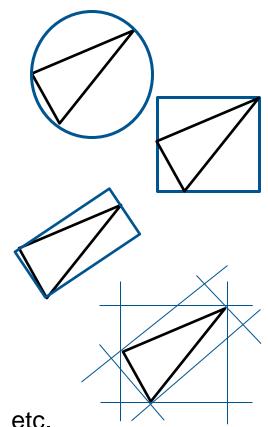
- Very simple intersection computation (min-max)
- Sometimes too large

Non-axis-aligned box

- A.k.a. "oriented bounding box (OBB)"
- Often better fit
- Fairly complex computation

Slabs

- Pairs of half spaces
- Fixed number of orientations/axes: e.g. x+y, x-y, etc.
 - Pretty fast computation



Bounding Volume Hierarchies (BVHs)

Definition

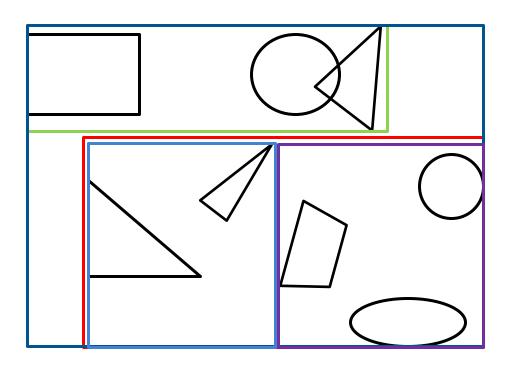
Hierarchical partitioning of a set of objects

BVHs form a tree structure

- Each inner node stores a volume enclosing all sub-trees
- Each leaf stores a volume and pointers to objects
- All nodes are aggregate objects
- Usually every object appears once in the tree
 - Except for instancing

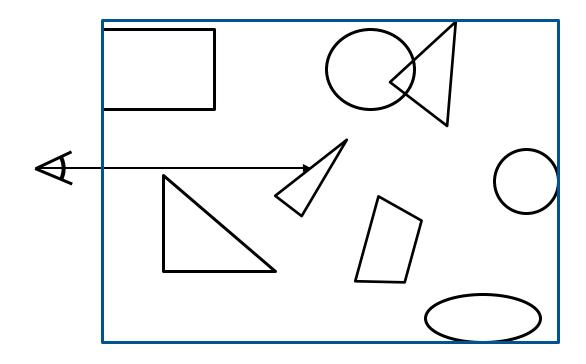
Bounding Volume Hierarchies (BVHs)

Hierarchy of groups of objects



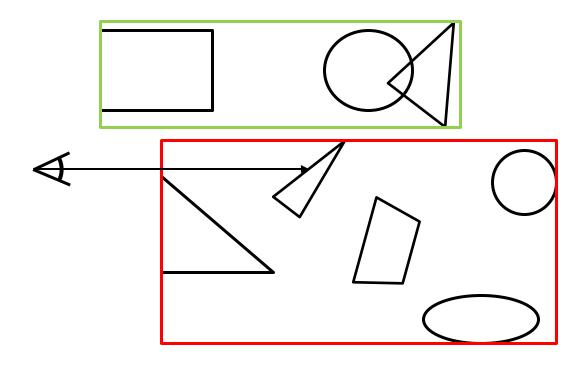
BVH traversal (1)

- Accelerate ray tracing
 - By eliminating intersection candidates
- Traverse the tree
 - Consider only objects in leaves intersected by the ray



BVH traversal (2)

- Accelerate ray tracing
 - By eliminating intersection candidates
- Traverse the tree
 - Consider only objects in leaves intersected by the ray



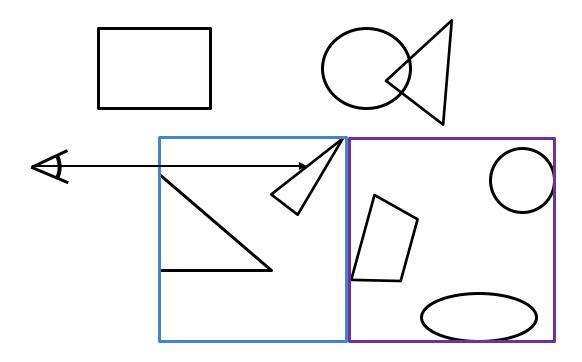
BVH traversal (3)

Accelerate ray tracing

By eliminating intersection candidates

Traverse the tree

- Consider only objects in leaves intersected by the ray
- Cheap traversal instead of costly intersection



Object vs. Space Partitioning

Object partitioning

- BVHs hierarchical partition objects into groups
- Create spatial index by spatially bounding each subgroup
- Subgroups may be overlapping!

Space partitioning

- (Hierarchically) partitions space in subspaces
- Subspaces are non-overlapping and completely fill parent space
- Organize them in a structure (tree or table)

Next: Space partitioning

Uniform Grids

Definition

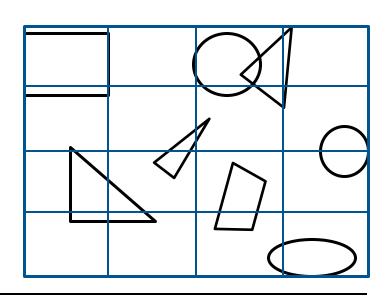
- Regular partitioning of space into equal-size cells
- Non-hierarchical structure

Resolution

- Want: number of cells in O(n)
- Resolution in each dimension proportional to $\sqrt[3]{n}$

- Usually
$$R_{x,y,z} = d_{x,y,z} \sqrt[3]{\frac{\lambda n}{V}}$$

- d: diagonal of box (a vector)
- n: #objects
- V: volume of Bbox
- λ: density (user-defined)



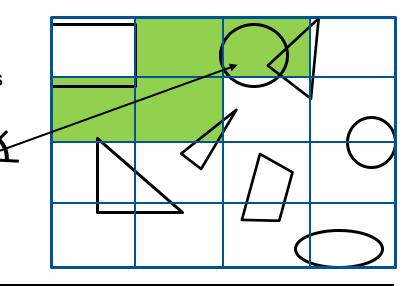
Uniform Grid Traversal

Grids are cheap to traverse

- 3D-DDA, modified Bresenham algorithm (see later)
- Step through the structure cell by cell
- Intersect with primitives inside non-empty cells

Mailboxing

- Single primitive can be referenced in many cells
- Avoid multiple intersections
- Keep track of intersection tests
 - Per-object cache of ray IDs
 - Problem with concurrent access
 - Per-ray cache of object IDs
 - Data local to a ray (better!)



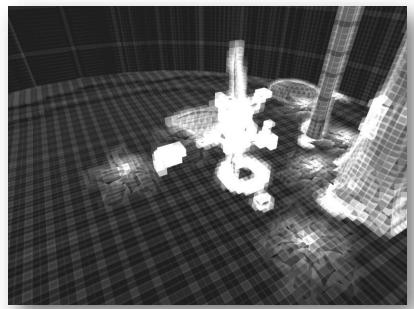
Nested Grids

Problem: "Teapot in a stadium"

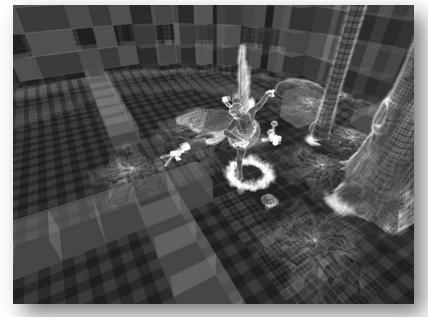
Uniform grids cannot adapt to local density of objects

Nested Grids

- Hierarchy of uniform grids: Each cell is itself a grid
- Fast algorithms for building & traversal (Kalojanov et al. '09, '11)



Cells of uniform grid (colored by # of intersection tests)



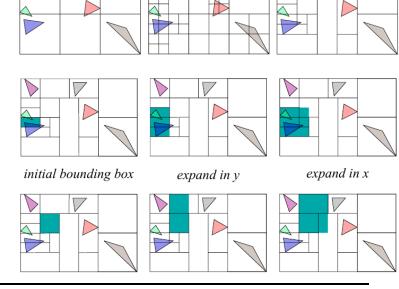
Same for two-level grid

Irregular Grids

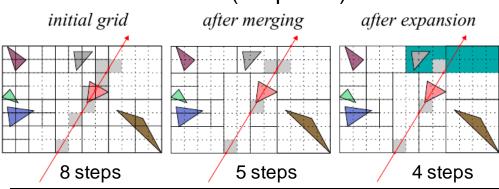
Irregular grids can accel traversal [Perard-Gayot'17]

- Build grid (hierarchical) base grid (power of 2, adapts to scene)
 - Base grid defines minimum resolution for computation
- Neighboring cells can be merged (eagerly)
 - As long as no change in set of primitives
- Can also expand cells (for exit operations)
 - As long as neighbors contain only subset of cells primitives
 - Allows for making larger steps
- Approach needs more memory

Construction (merge & expand)







Octrees and Quadtrees

Octree

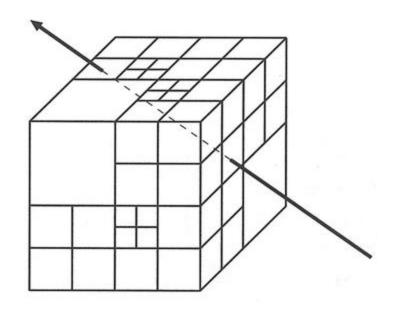
- Hierarchical space partitioning ("simplest hierarchical grid")
- Each inner node contains 8 (2x2x2 grid) equally sized voxels

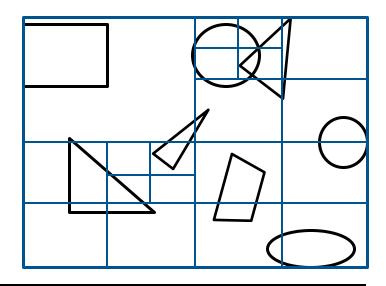
Quadtree

- 2D "octree"

Adaptive subdivision

Adjust depth to local scene complexity





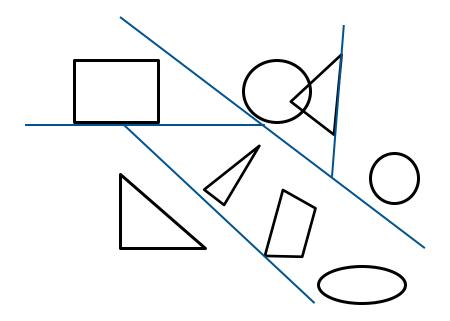
BSP Trees

Definition

- Binary Space Partition Tree (BSP)
- Recursively split space with planes
 - Arbitrary split positions
 - Arbitrary orientations

Used for visibility computation

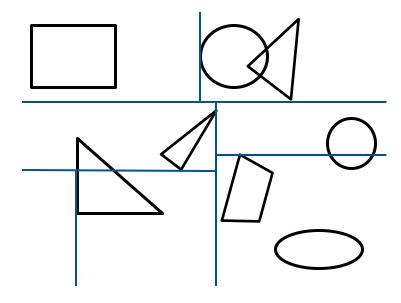
- E.g. in games (Doom)
- Enumerating objects in back to front order



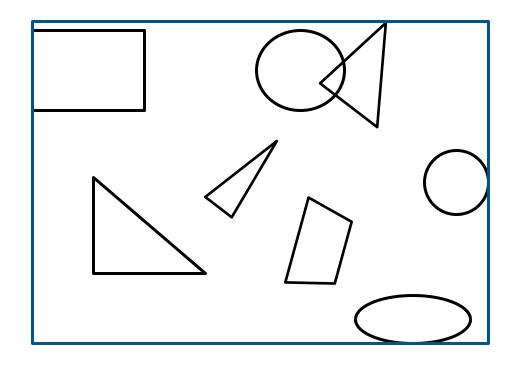
kD-Trees

Definition

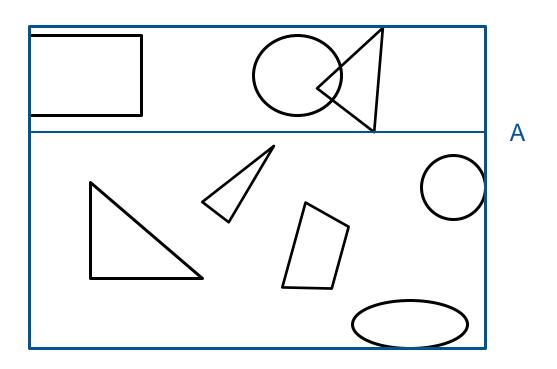
- Axis-Aligned Binary Space Partition Tree
- Recursively split space with axis-aligned planes
 - Arbitrary split positions
 - Greatly simplifies/accelerates computations



kD-Tree Example (1)

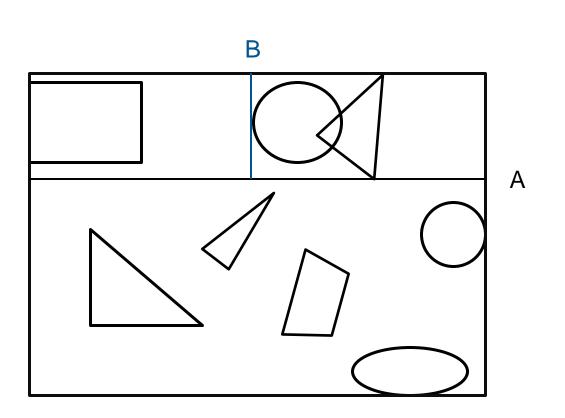


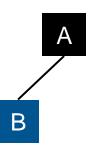
kD-Tree Example (2)



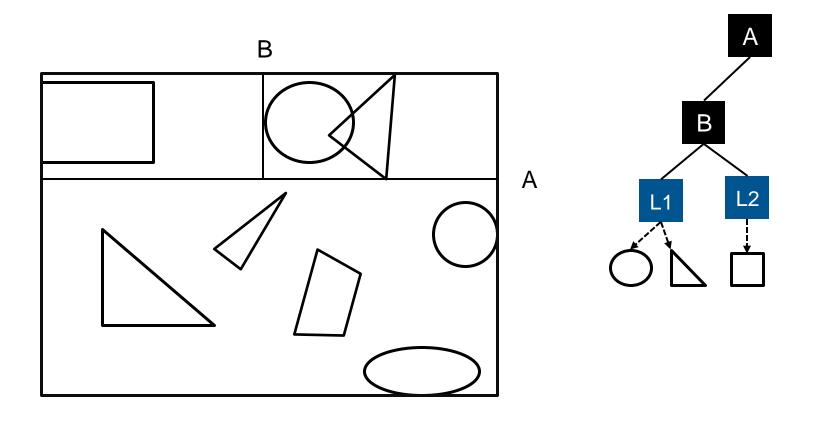


kD-Tree Example (3)

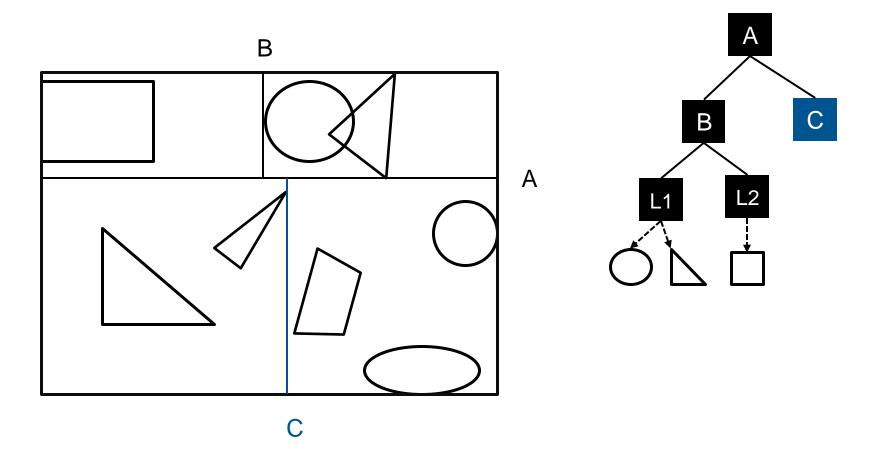




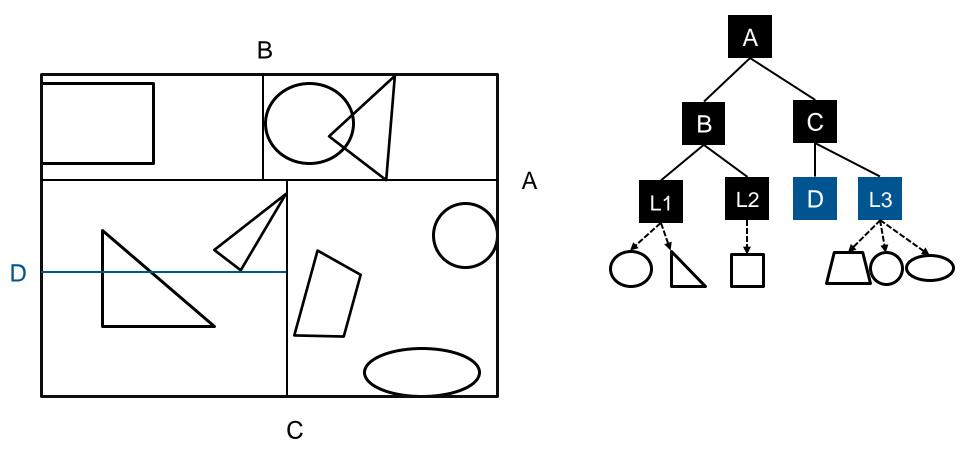
kD-Tree Example (4)



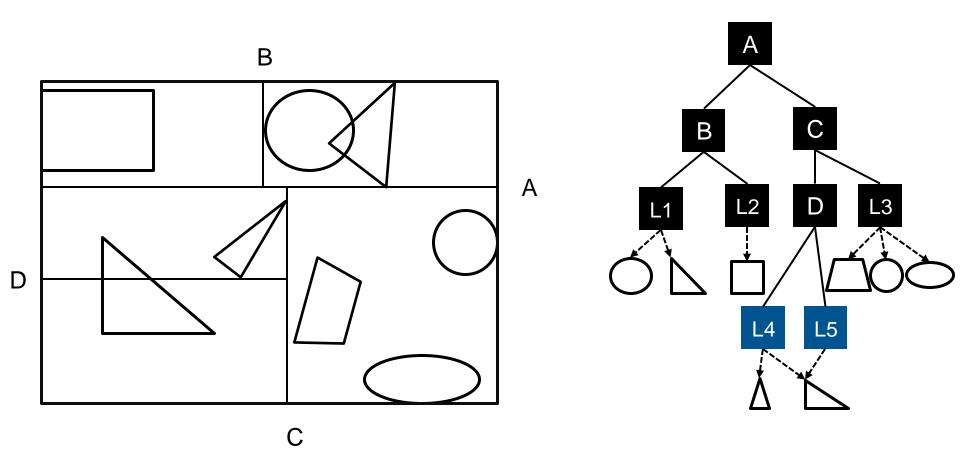
kD-Tree Example (5)



kD-Tree Example (6)



kD-Tree Example (7)



kD-Tree Traversal

"Front-to-back" traversal

Traverse child nodes in order along rays

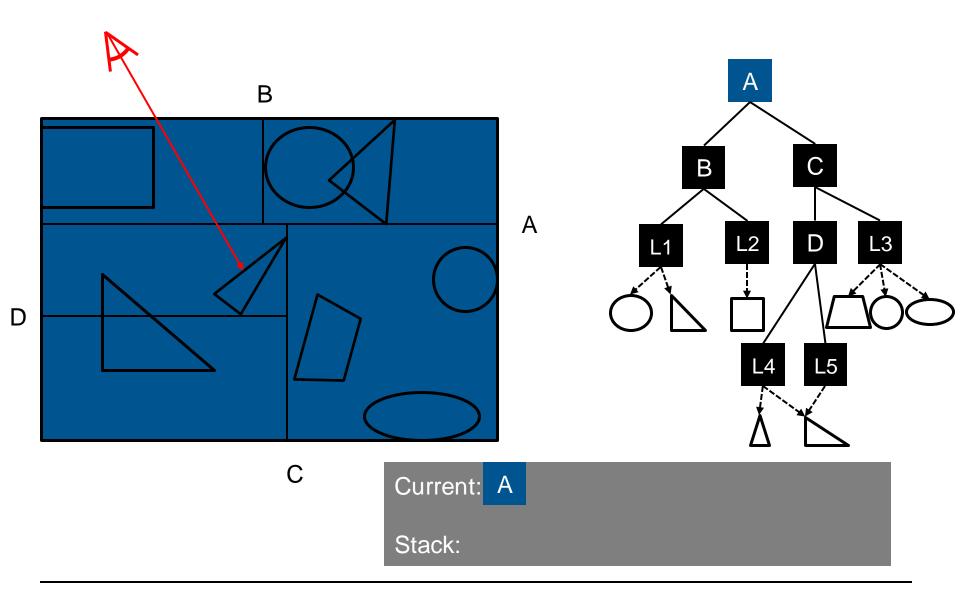
Termination criterion

 Traversal can be terminated as soon as surface intersection is found in the current node

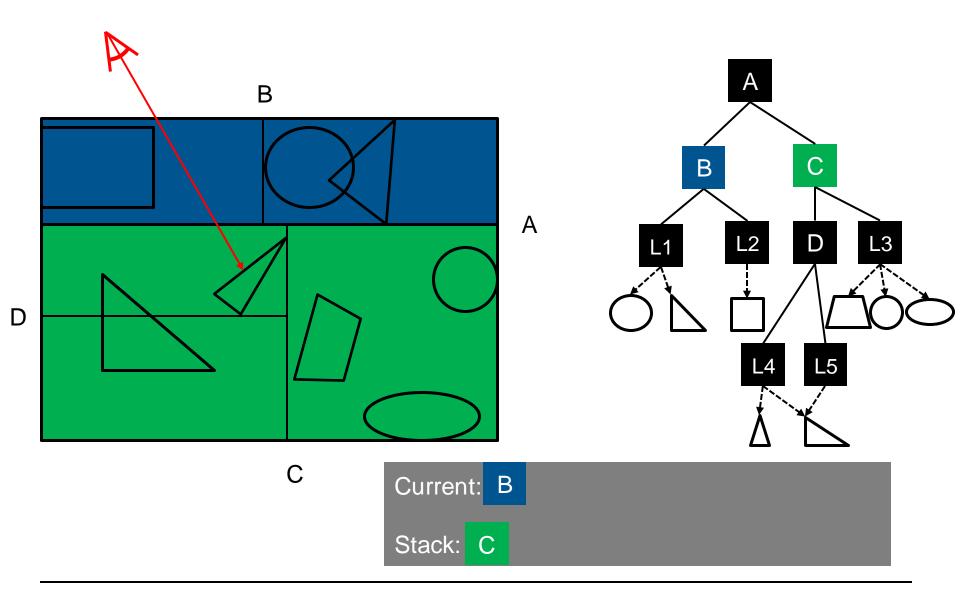
Maintain stack of sub-trees still to traverse

- More efficient than recursive function calls
- Algorithms with no or limited stacks are also available (for GPUs)

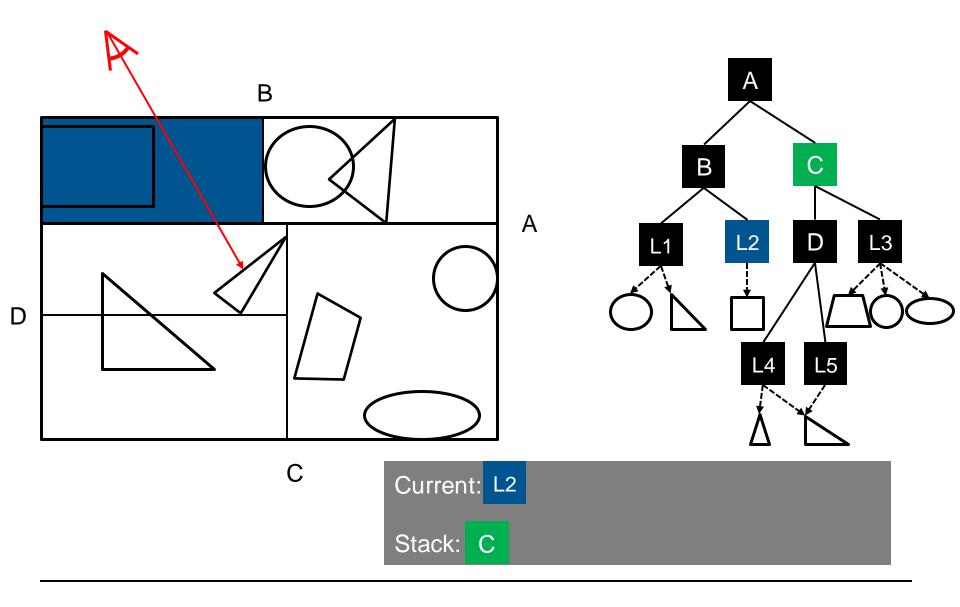
kD-Tree Traversal (1)



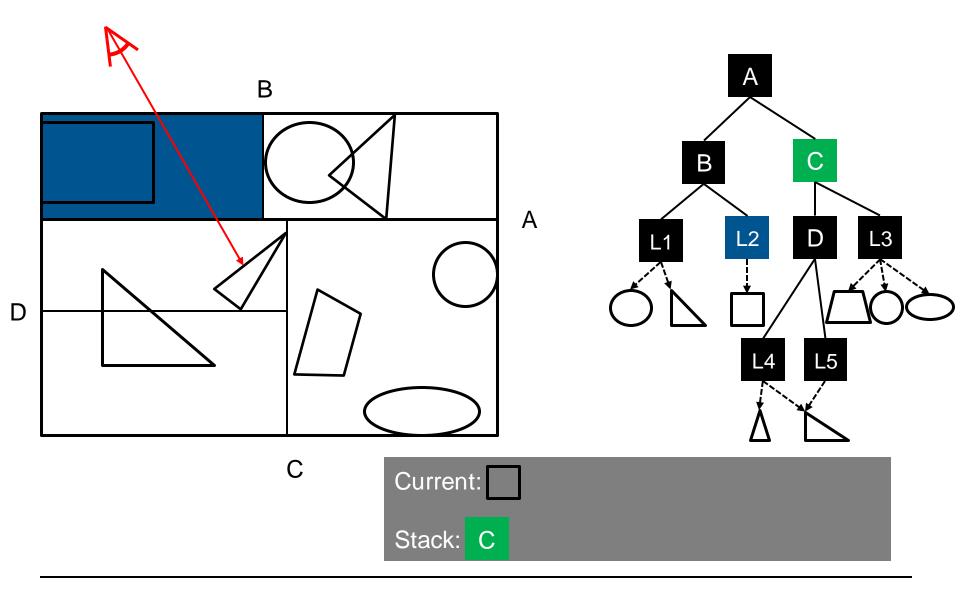
kD-Tree Traversal (2)



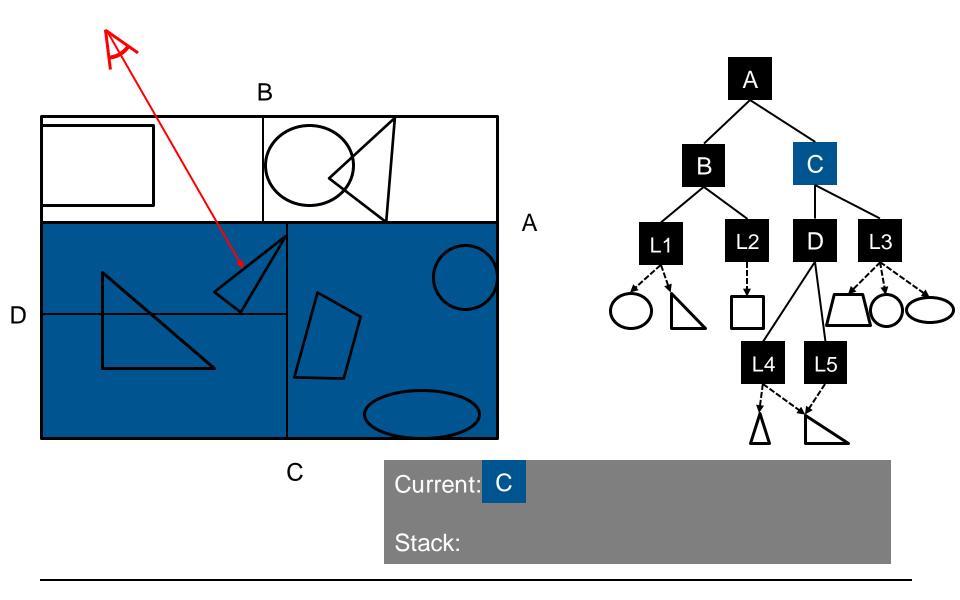
kD-Tree Traversal (3)



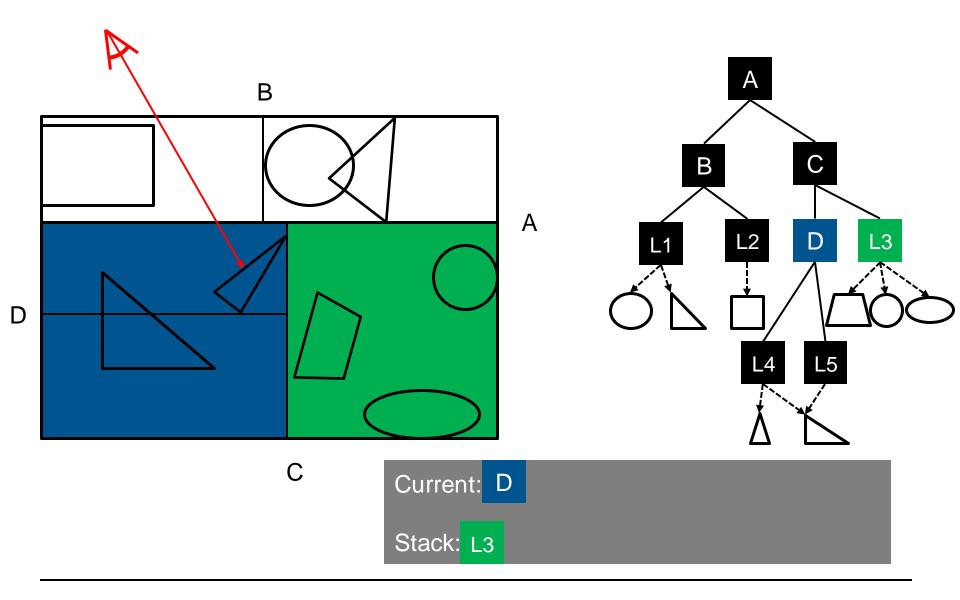
kD-Tree Traversal (4)



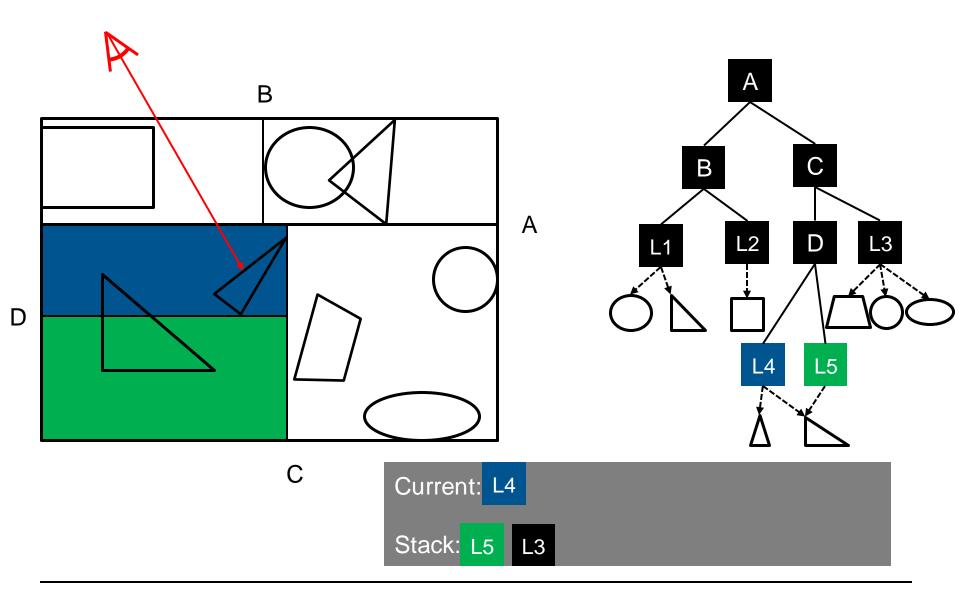
kD-Tree Traversal (5)



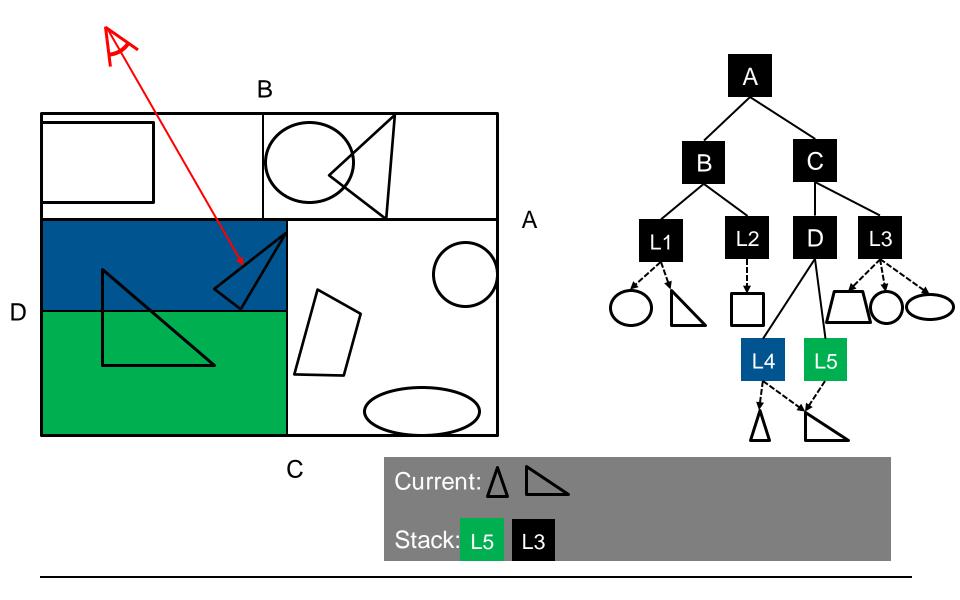
kD-Tree Traversal (6)



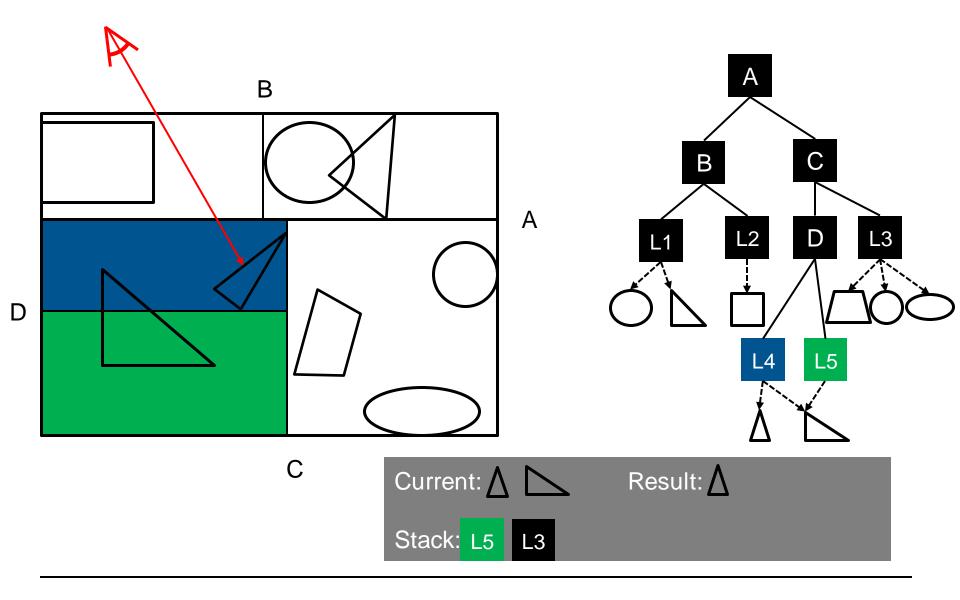
kD-Tree Traversal (7)



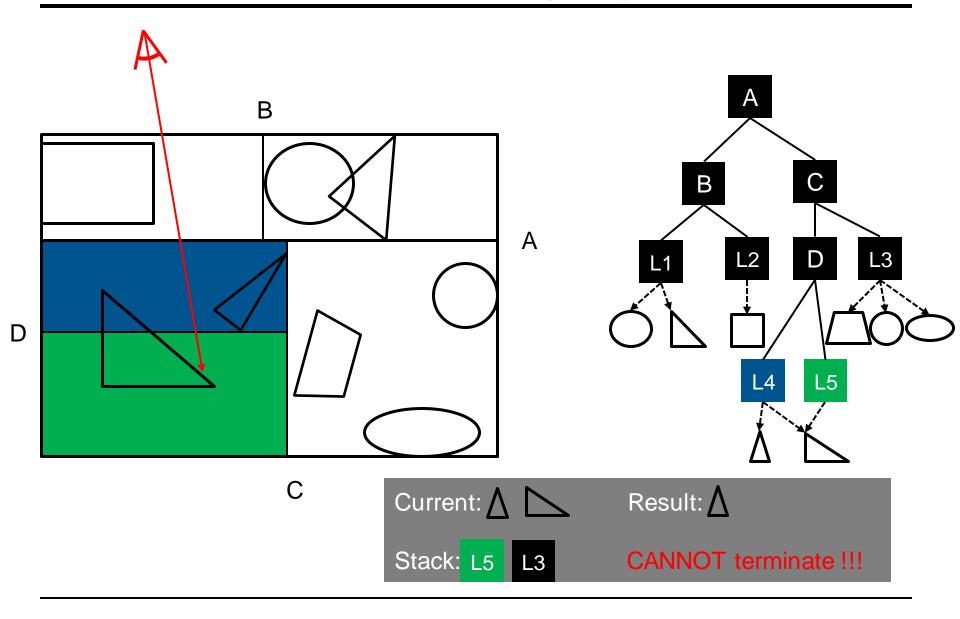
kD-Tree Traversal (8)



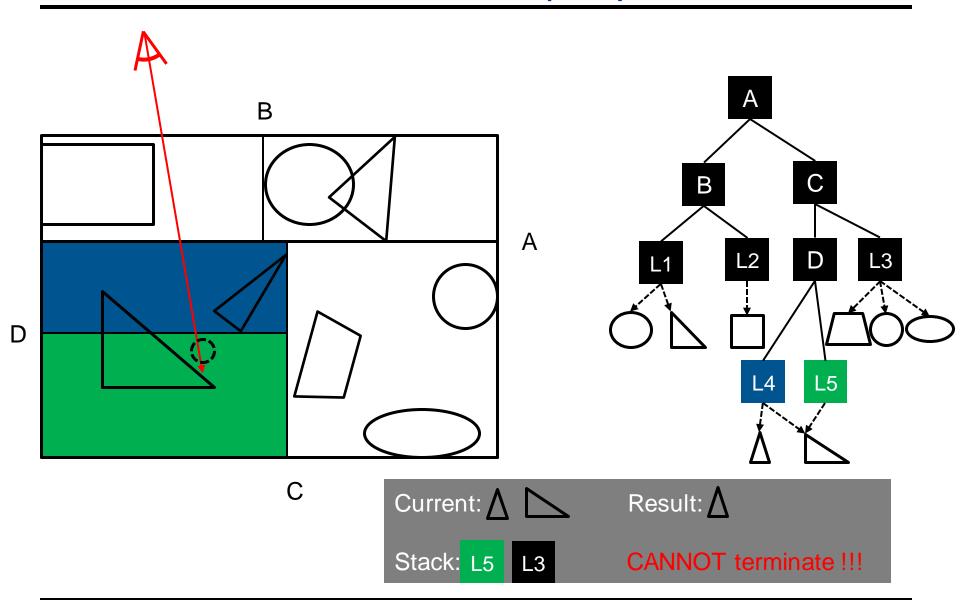
kD-Tree Traversal (9)



kD-Tree Traversal (10)



kD-Tree Traversal (11)



kD-Tree Properties

kD-Trees

- Split space instead of sets of objects
- Split into disjoint, fully covering regions

Adaptive

Can handle the "Teapot in a Stadium" well

Compact representation

- Relatively little memory overhead per node
- Node stores:
 - Split location (1D), child pointer (to both children), Axis-flag (often merged into pointer)
 - Can be compactly stored in 8 bytes
- But replication of objects in (possibly) many nodes
 - Can greatly increase memory usage

Cheap Traversal

- One subtraction, multiplication, decision, and fetch
- But many more cycles due to instruction dependencies

Overview: kD-Trees Construction

- Adaptive
- Compact
- Cheap traversal

Exploit Advantages

Adaptive

You have to build a good tree

Compact

- At least use the compact node representation (8-byte)
- You can't be fetching whole cache lines every time

Cheap traversal

No sloppy inner loops! (one subtract, one multiply!)

Building kD-trees

Given:

- Axis-aligned bounding box ("cell")
- List of geometric primitives (triangles?) touching cell

Core operation:

- Pick an axis-aligned plane to split the cell into two parts
- Sift geometry into two batches (some redundancy)
- Recurse

Building kD-trees

Given:

- Axis-aligned bounding box ("cell")
- List of geometric primitives (triangles?) touching cell

Core operation:

- Pick an axis-aligned plane to split the cell into two parts
- Sift geometry into two batches (some redundancy)
- Recurse
- Termination criteria!

"Intuitive" kD-Tree Building

Split Axis

Round-robin; largest extent

Split Location

Middle of extent; median of geometry (balanced tree)

Termination

Target # of primitives, limited tree depth

"Intuitive" kD-Tree Building

- Split Axis
 - Round-robin; largest extent
- Split Location
 - Middle of extent; median of geometry (balanced tree)
- Termination
 - Target # of primitives, limited tree depth
- All of these techniques are NOT very clever

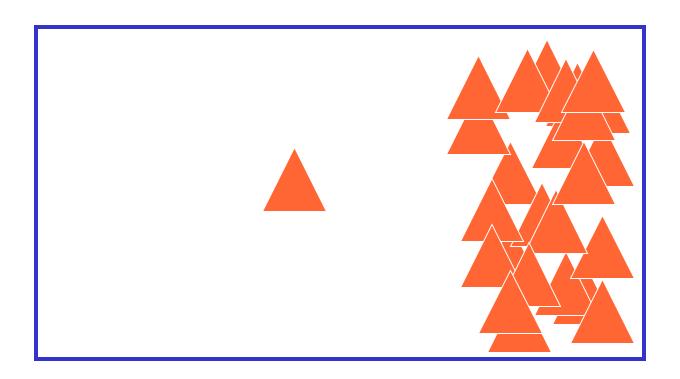
Building good kD-trees

- What split do we really want?
 - Clever Idea: The one that makes ray tracing cheap
 - Write down an expression of cost and minimize it
 - → Cost Optimization
- What is the cost of tracing a ray through a cell?
 - Surface Area Heuristic (SAH)

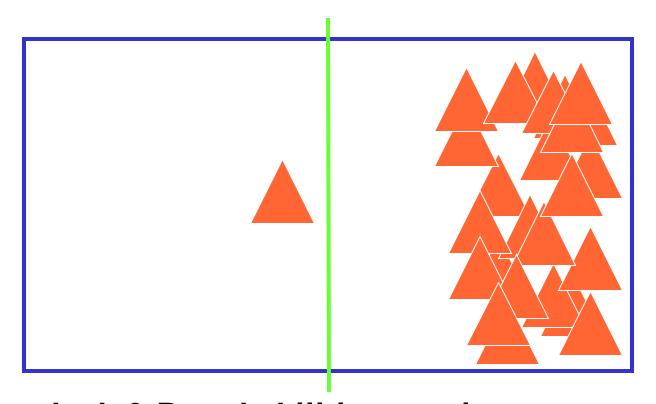
```
Cost(cell) = C_trav + Prob(hit L) * Cost(L) + Prob(hit R) * Cost(R)
```

- Cost of traversal of the inner node itself, plus
- Relative probability of hitting one child, times
- Cost of hitting that child
- Same for other child

Splitting with Cost in Mind

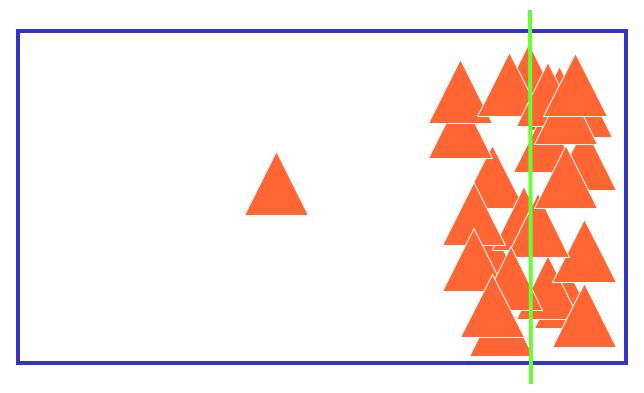


Split in the middle



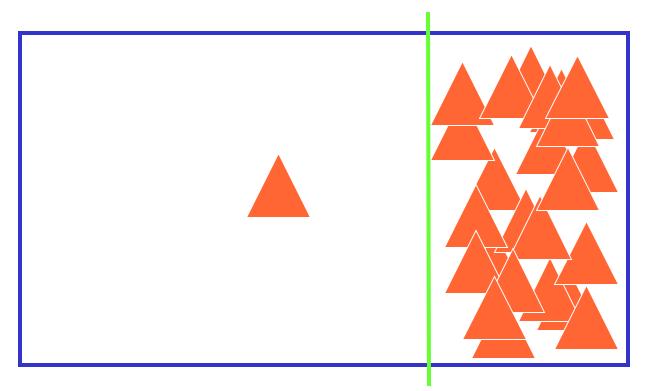
- Makes the L & R probabilities equal
- Pays no attention to the L & R costs

Split at the Median



- Makes the L & R costs equal
- Pays no attention to the L & R probabilities

Cost-Optimized Split



- Automatically and rapidly isolates complexity
- Produces large chunks of empty space

Building good kD-trees

Need the probabilities

- Turns out to be proportional to surface area (SA)
- Not the volume

Need the child cell costs

- Simple triangle count works great (very rough approx.)
- Many attempts to improve this did not work out

```
Cost(c) = C_{trav} + Prob(hit L) * Cost(L) + Prob(hit R) * Cost(R)= C_{trav} + SA(L)/SA(c) * TriCount(L) + SA(R)/SA(c) * TriCount(R)
```

Termination Criteria

When should we stop splitting?

- Another clever idea: When splitting does not help any more.
- Use the cost estimates in your termination criteria

Threshold of cost improvement

But stretch decision over multiple levels, to avoid local minima

Threshold of cell size

Absolute (!) probability so small there is no point in going on

Building good kD-trees

Basic build algorithm

- Pick an axis, or optimize across all three
- Build a set of candidate split locations
 - Based on BBox of triangles (in/out events) or
 - Predefined locations (fixed number of bins across bbox axis)
- Sort the triangle events or bin them
- Walk through candidates to find minimum cost split

Characteristics of the tree you're looking for

- Deep and thin
- Typical depth of 50-100,
- About 2 triangles per leaf,
- Big empty cells

Building kD-trees quickly

- Very important to build good trees first
 - Otherwise you have no basis for comparison
- Don't give up cost optimization!
 - Use the math, Luke…
- Luckily, lots of flexibility...
 - Axis picking ("hack" pick vs. full optimization)
 - Candidate picking (bboxes, exact; binning, sorting)
 - Termination criteria ("knob" controlling tradeoff)

Building kD-trees quickly

- Remember, profile first! Where's the time going?
 - Split personality
 - Memory traffic all at the top (NO cache misses at bottom)
 - Sifting through bajillion triangles to pick one split (!)
 - Hierarchical building?
 - Computation mostly at the bottom
 - Lots of leaves, need more exact candidate info
 - Lazy building?
 - Change criteria during the build?

Fast Ray Tracing w/kD-Trees

- Adaptive
 - Build a cost-optimized kD-tree w/ the surface area heuristic
- Compact
- Cheap traversal

What's in a node?

A kD-tree internal node needs:

- Am I a leaf?
- Split axis
- Split location
- Pointers to children

Compact (8-byte) Nodes

- kD-Tree node can be packed into 8 bytes
 - Split location
 - 32 bit float
 - Always two children, put them side-by-side
 - Only one 32-bit pointer
 - Leaf flag + Split axis
 - 2 bits

Compact (8-byte) Nodes

kD-Tree node can be packed into 8 bytes

- Split location
 - 32 bit float
- Always two children, put them side-by-side
 - Only one 32-bit pointer
- Leaf flag + Split axis
 - 2 bits

So close! Sweep those 2 bits under the rug...

- Encode bits in lowest 2 bits of pointer
- Bits are not used as structure is multiple of 8, anyway

No Bounding Box!

- kD-Tree node corresponds to an AABB
- Does not mean it has to *contain* one
 - Would be 24 bytes: 4X explosion (!)

Memory Layout

- Cache lines are much bigger than 8 bytes!
 - Advantage of compactness lost with poor layout
- Pretty easy to do something reasonable
 - Building depth first, watching memory allocator

Other Data

- Memory should be separated by rate of access
 - Frames
 - << Pixels</p>
 - << Samples [Ray Trees]</p>
 - << Rays [Shading (not quite)]</p>
 - << Triangle intersections</p>
 - << Tree traversal steps</p>
- Example: pre-processed triangle, shading info...

Fast Ray Tracing w/kD-Trees

Adaptive

Build a cost-optimized kD-tree w/ the surface area heuristic

Compact

- Use an 8-byte node
- Lay out your memory in a cache-friendly way

Cheap traversal

kD-Tree Traversal Operation

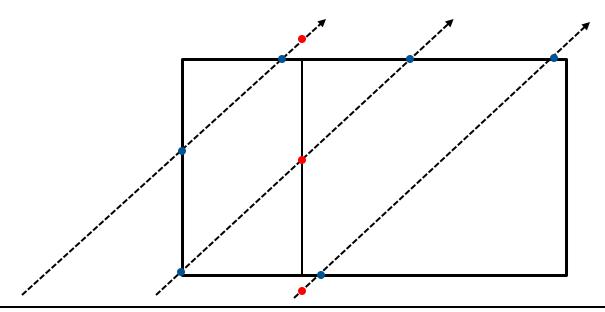
- Maintain on a stack
 - Entry and exit distance to node (t_near and t_far)
- Three cases

```
t_split > t_far:Go only to near node
```

- t_near < t_split < t_far Go to both (use stack)</p>

t_split < t_nearGo only to far node

Near and far depend on direction of ray!



kD-Tree Traversal: Inner Loop

```
Given (node, t_near, t_far)
while (! node.isLeaf())
   t_at_split = ( split_location - ray->origin[split_axis] ) * ray->inv_dir[split_axis]
   if (t_split <= t_min)
         continue with (far child, t_split, t_far) // hit either far child or none
   if (t_split >= t_max)
                                                       // hit near child only
         continue with (near child, t_min, t_split)
   // hit both children
   push (far child, t_split, t_max) onto stack
   continue with (near child, t_min, t_split)
```

Optimize Your Inner Loop

- kD-Tree traversal is the most critical kernel
 - It happens about a zillion times
 - It's tiny
 - Sloppy coding will show up
- Optimize, Optimize, Optimize
 - Remove recursion and minimize stack operations
 - Other standard tuning & tweaking

Can it go faster?

- How do you make fast code go faster?
- Parallelize it!
 - Not covered here

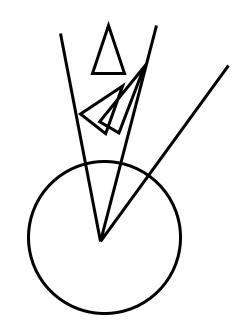
Directional Partitioning

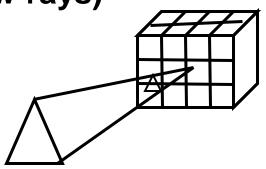
Applications

- Useful only for rays that start from a single point
 - Camera
 - Point light sources
- Preprocessing of visibility
- Requires scan conversion of geometry
 - For each object locate where it is visible
 - Expensive and linear in # of objects
- Generally not used for primary rays



- Lazy and conservative evaluation
- Store last found occluder in directional structure
- Test entry first for next shadow test

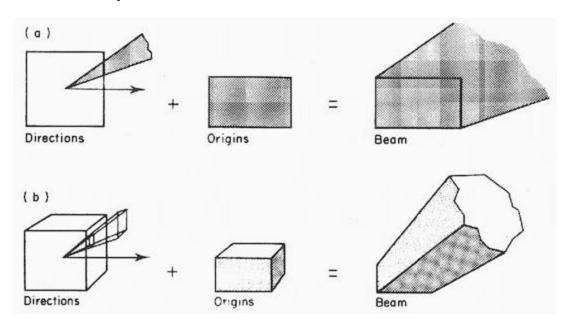




Ray Classification

Partitioning of space and direction [Arvo & Kirk'87]

- Roughly pre-computes visibility for the entire scene
 - What is visible from each point in each direction?
- Very costly preprocessing, cheap traversal
 - Improper trade-off between preprocessing and run-time
- Memory hungry, even with lazy evaluation
- Seldom used in practice

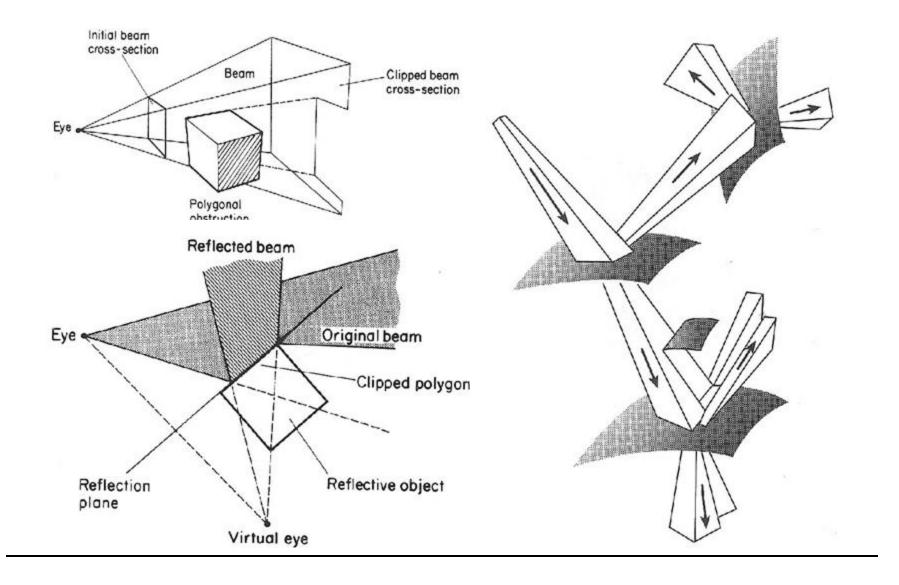


Packet Tracing

Approach

- Combine many similar rays (e.g. primary or shadow rays)
- Trace them together in SIMD fashion
 - All rays perform the same traversal operations
 - All rays intersect the same geometry
 - Can use SIMD instructions in modern processors
- Exposes coherence between rays
 - All rays touch similar spatial indices
 - Loaded data can be reused (in registers & cache)
 - More computation per recursion step → better optimization
- Overhead
 - Rays will perform unnecessary operations
 - Overhead low for coherent and small set of rays (e.g. up to 4x4 rays)
- Needs an API that provides coherent sets of rays

Beam Tracing



Beam and Cone Tracing

General idea:

Trace continuous bundles of rays

Cone Tracing:

- Approximate collection of ray with cone(s)
- Subdivide into smaller cones if necessary

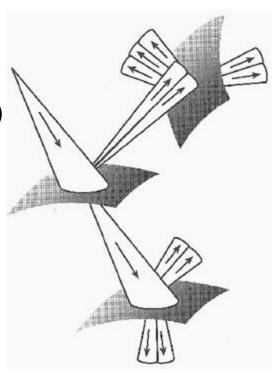
Beam Tracing:

- Exactly represent a ray bundle with pyramid
- Create new beams at intersections (polygons)

Problems:

- Clipping of beams?
- Good approximations?
- How to compute intersections?

Not really practical !!

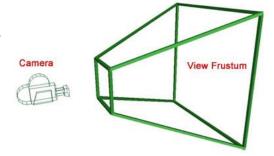


Frustum Tracing

- Bound set of rays with frustum (NOT frustrum!!)
 - Only during traversal
 - API needs to provide coherent groups of rays
 - Possibly hierarchically



- Small overhead (largely avoided by SIMD)
 - Compute with 4 corner rays
- Avoid traversing many rays individually
 - Particularly beneficial in the upper levels of index
- Switch to (packets of) rays when needed (intersection)
 - Might be able to only use subset (e.g. based on extend of triangle)
- Split frustum hierarchically and traverse separately in lower levels
 - Avoids overhead of carrying to many rays into small nodes
- E.g. fast primary ray traversal by W. Hunt (Oculus)



Distribution Ray Tracing

- Formerly called Distributed Ray Tracing [Cook`84]
- Stochastic Sampling of

Pixel: Antialiasing

– Lens: Depth-of-field

BRDF: Glossy reflections

Lights: Smooth shadows from

area light sources

– Time: Motion blur

Covered in detail in RIS course

