### **Computer Graphics**

- Spatial Index Structures -

**Philipp Slusallek** 

### 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

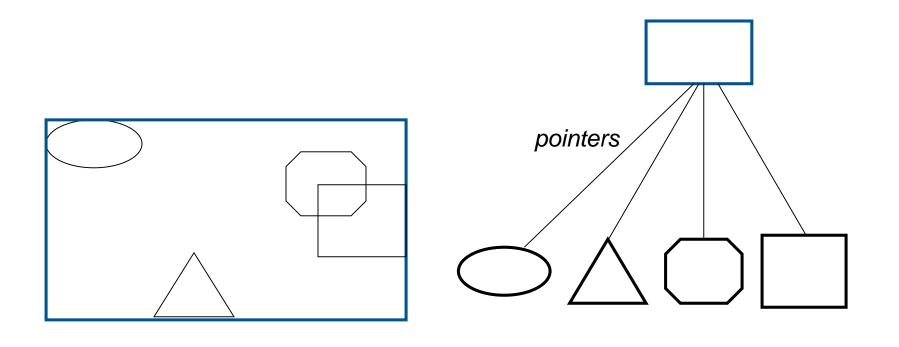
- Spatial indexing structures
- (Hierarchically) partition space or the set of objects
- Examples
  - Grids, hierarchies of grids
  - Octrees
  - Binary space partitions (BSP) or kd-trees
  - Bounding volume hierarchies (BVH)
- Directional partitioning (not very useful)
- 5D partitioning (space and direction, 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)

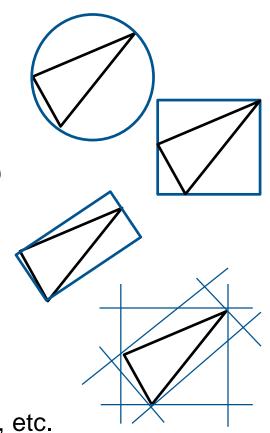
- Very simple intersection computation (min-max)
- Sometimes too large

### 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

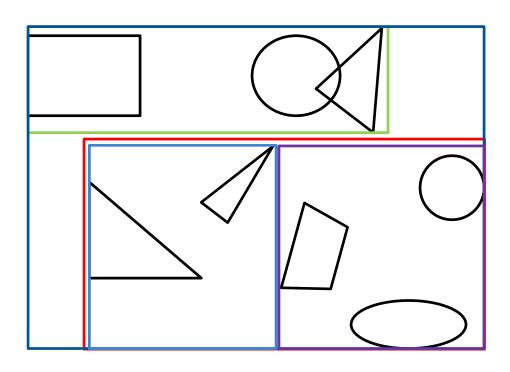
Hierarchical partitioning of a set of objects

#### 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

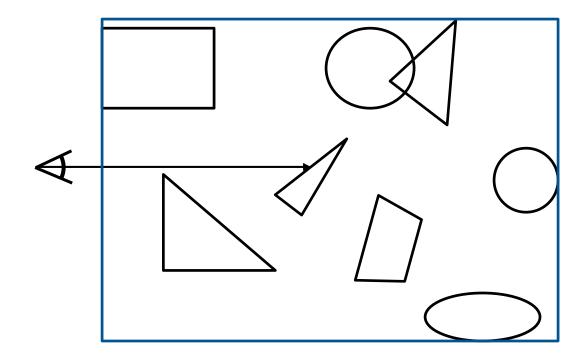
### Bounding Volume Hierarchies (BVHs)

Hierarchy of groups of objects



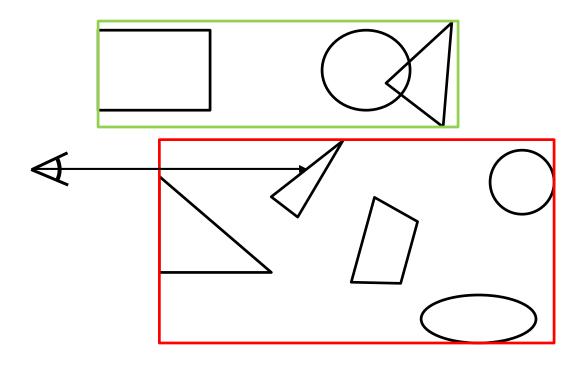
## BVH traversal (1)

- 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



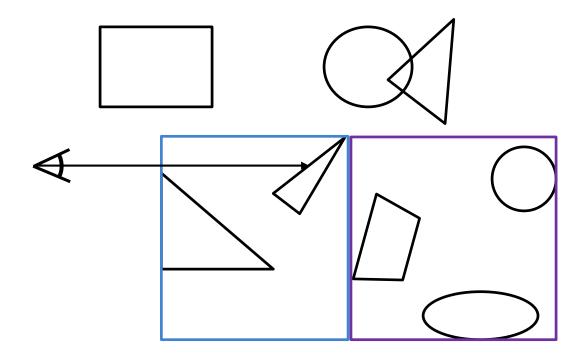
## BVH traversal (3)

### 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

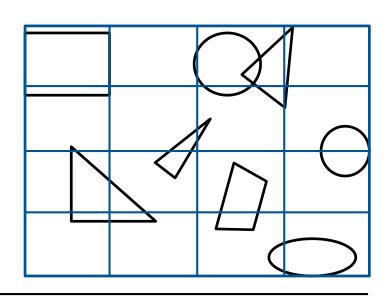
- Regular partitioning of space into equal-size cells
- Non-hierarchical structure

#### 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)



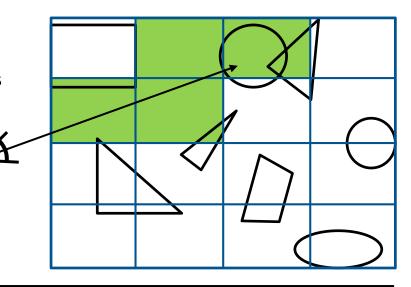
### **Uniform Grid Traversal**

### 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!)



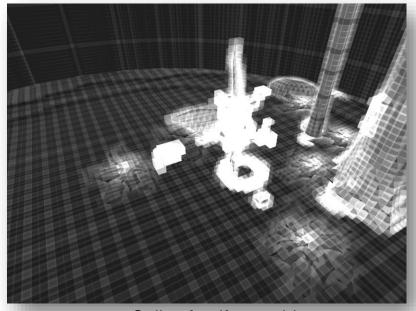
### **Nested Grids**

### Problem: "Teapot in a stadium"

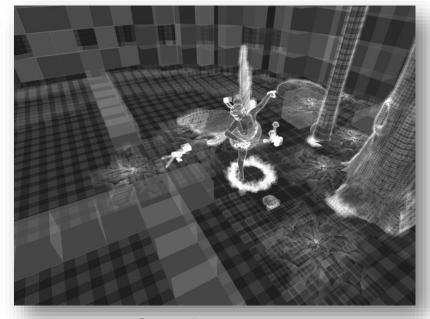
Uniform grids cannot adapt to local density of objects

#### 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)



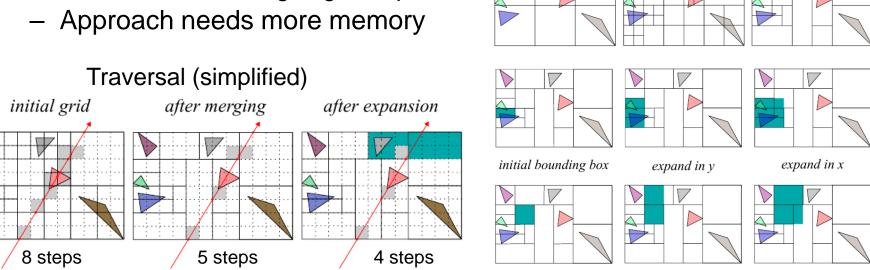
Same for two-level grid

## 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

#### Construction (merge & expand)



### Octrees and Quadtrees

#### Octree

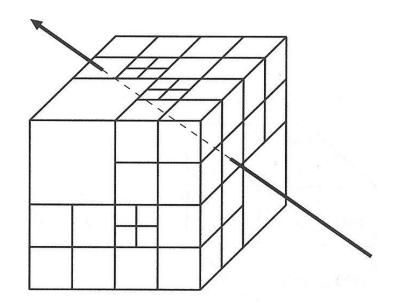
- Hierarchical space partitioning ("simplest hierarchical grid")
- Each inner node contains 8 (2x2x2 grid) equally sized voxels

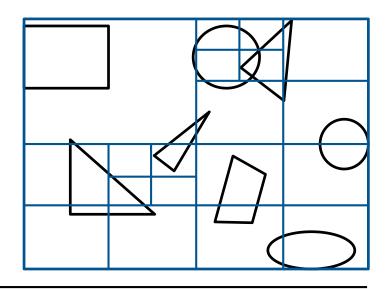
#### Quadtree

2D "octree"

### Adaptive subdivision

Adjust depth to local scene complexity





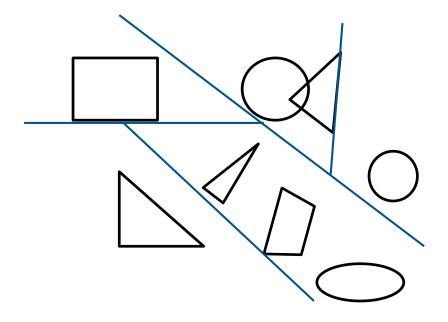
### **BSP Trees**

#### Definition

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

### Used for visibility computation

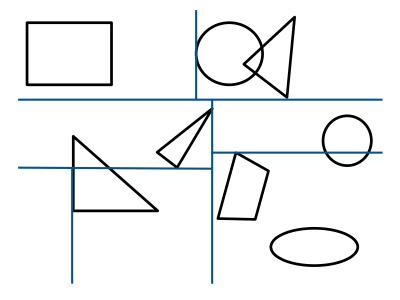
- E.g. in games (Doom)
- Enumerating objects in back to front order



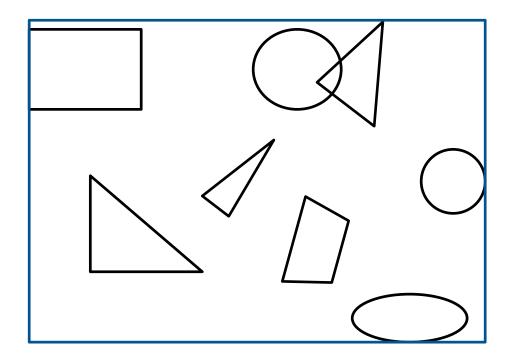
### kD-Trees

#### Definition

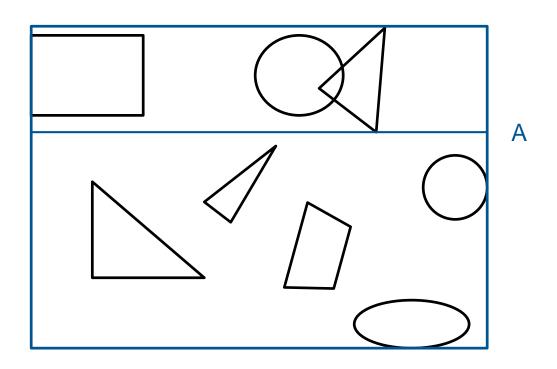
- Axis-Aligned Binary Space Partition Tree
- Recursively split space with axis-aligned planes
  - Arbitrary split positions
  - Greatly simplifies/accelerates computations



# kD-Tree Example (1)

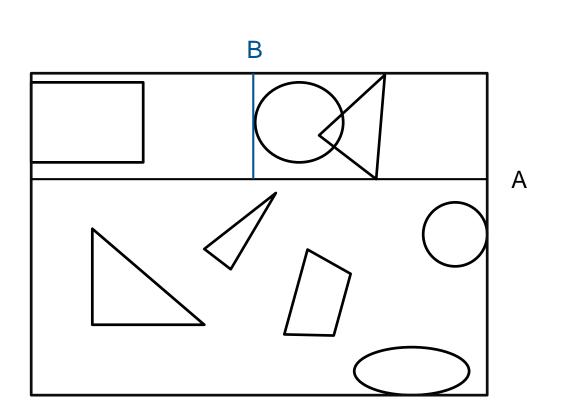


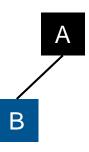
# kD-Tree Example (2)



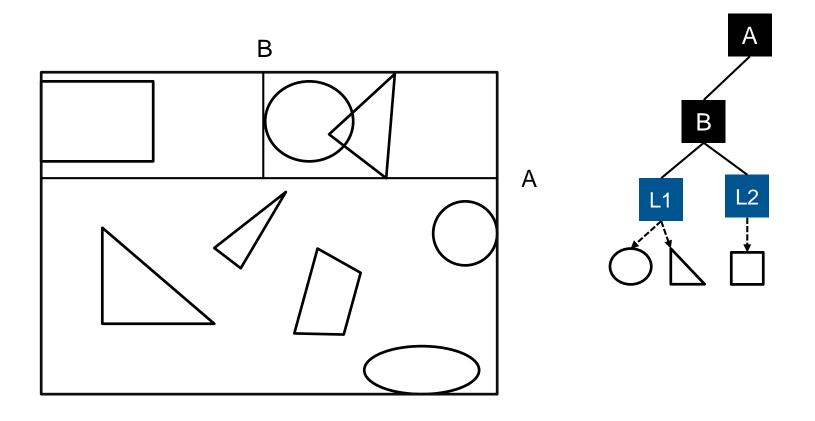


# kD-Tree Example (3)

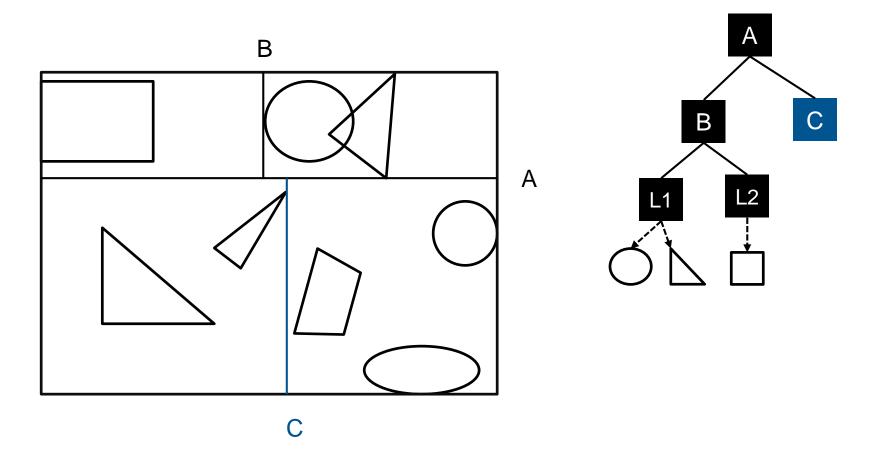




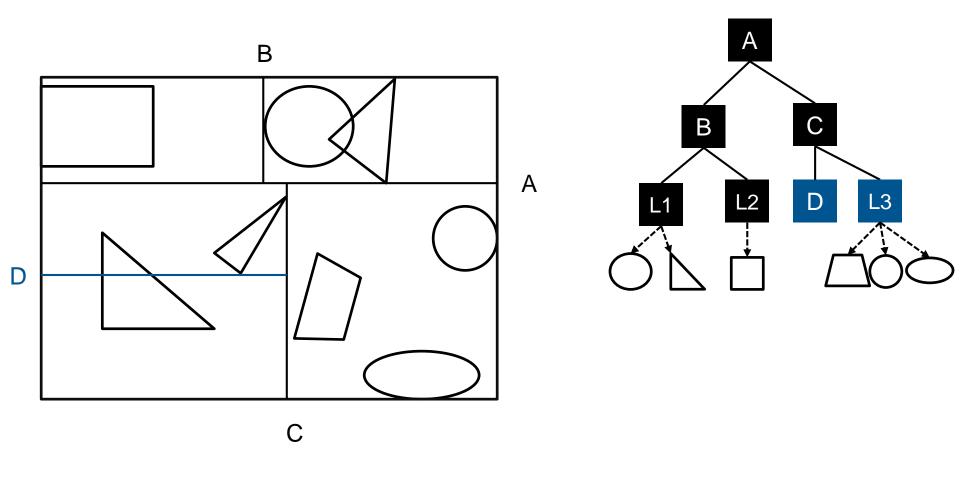
# kD-Tree Example (4)



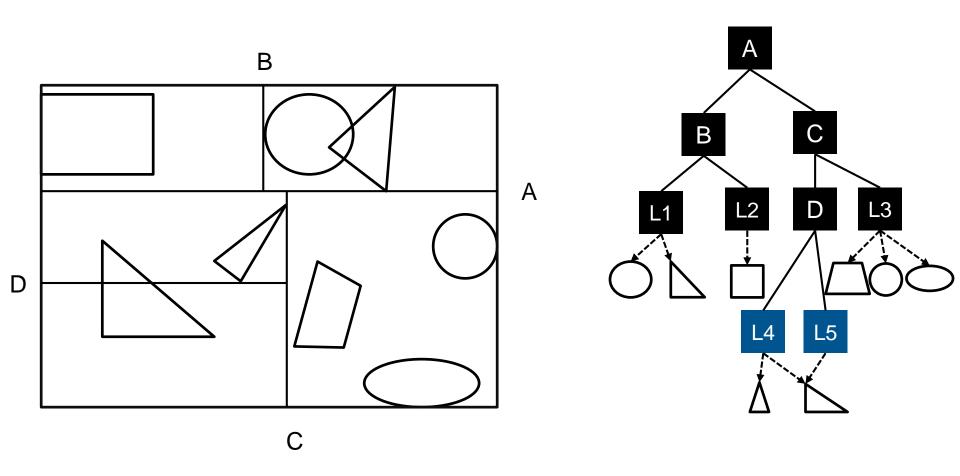
# kD-Tree Example (5)



# kD-Tree Example (6)



# kD-Tree Example (7)



### kD-Tree Traversal

#### "Front-to-back" traversal

Traverse child nodes in order along rays

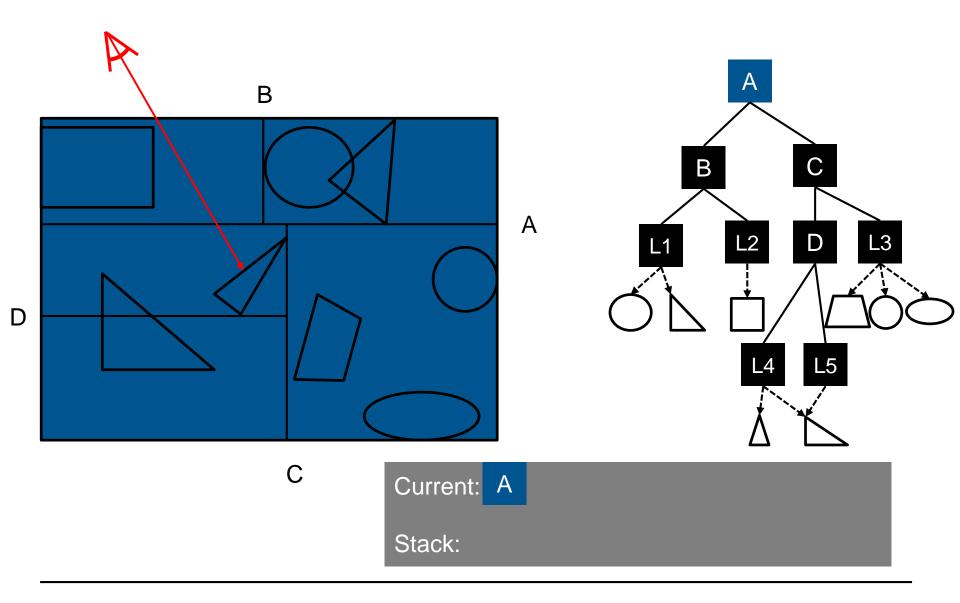
#### Termination criterion

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

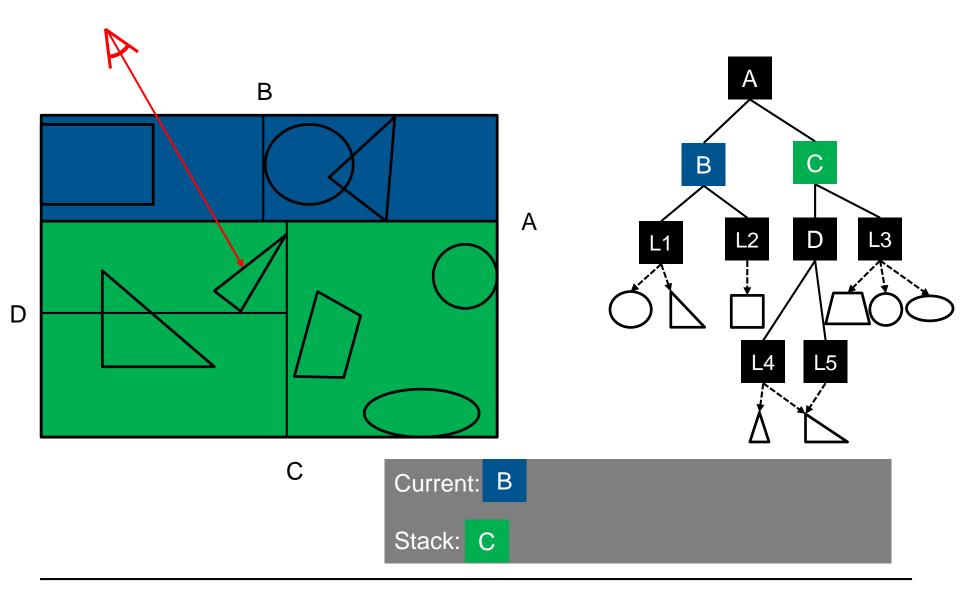
#### Maintain stack of sub-trees still to traverse

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

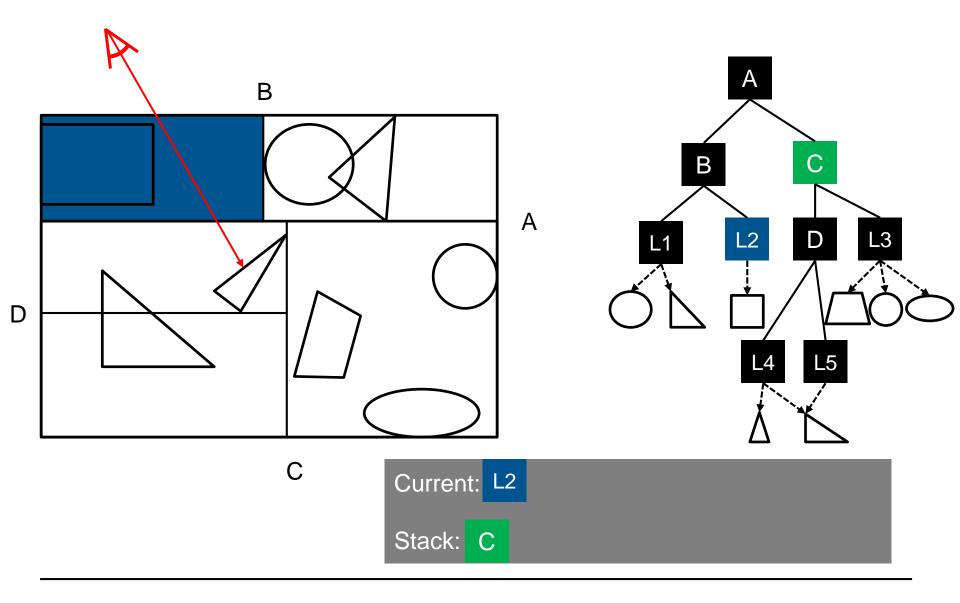
# kD-Tree Traversal (1)



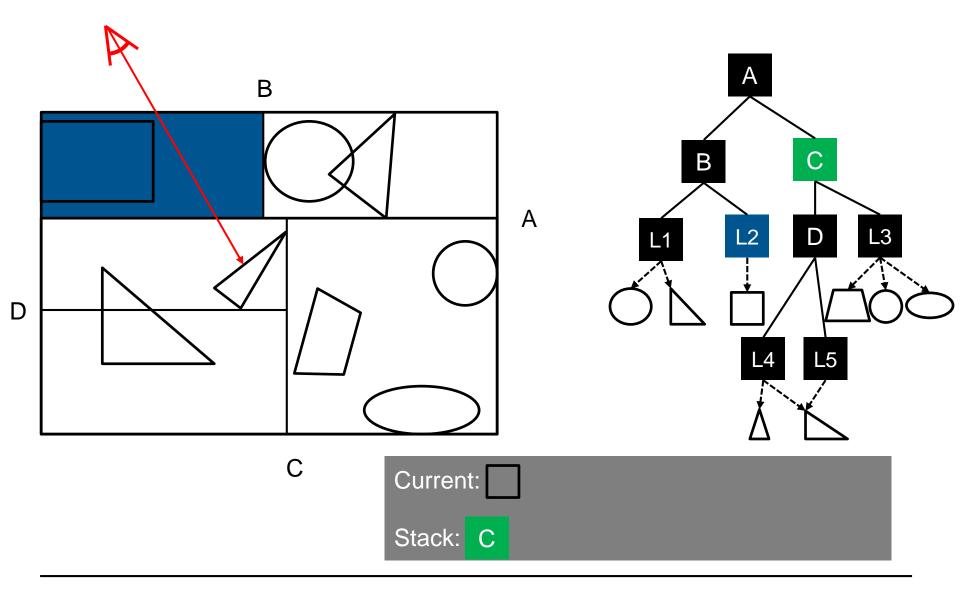
# kD-Tree Traversal (2)



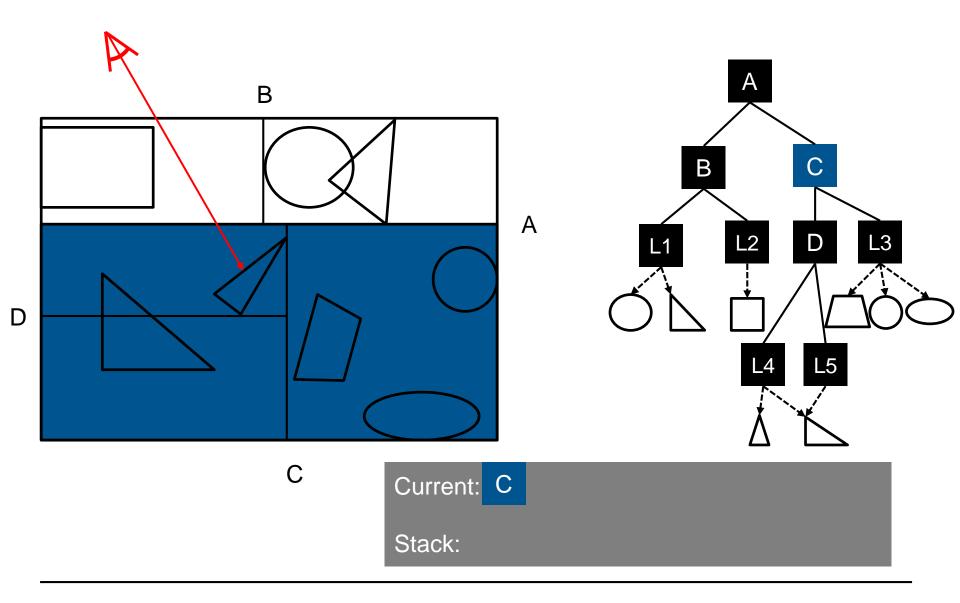
# kD-Tree Traversal (3)



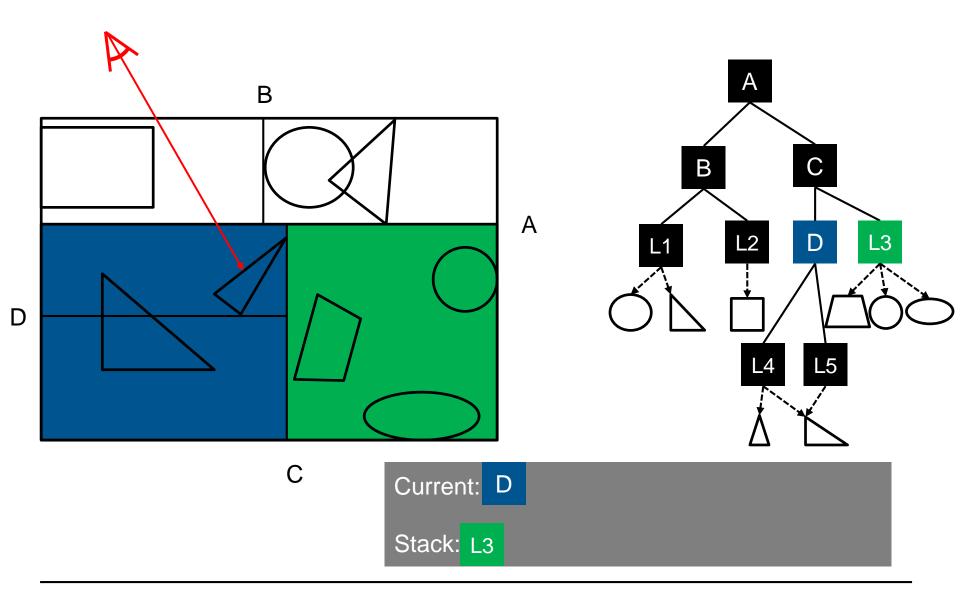
# kD-Tree Traversal (4)



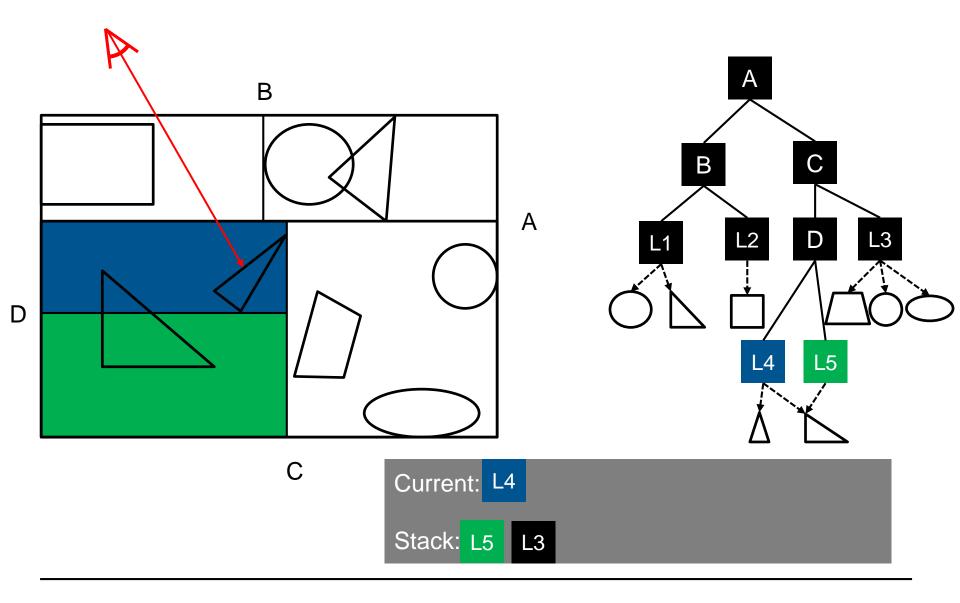
# kD-Tree Traversal (5)



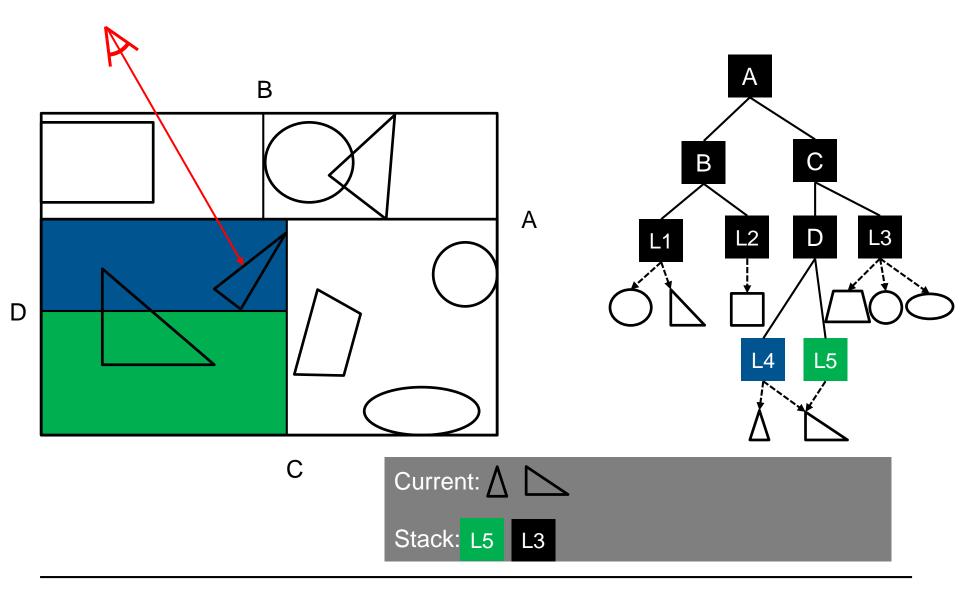
# kD-Tree Traversal (6)



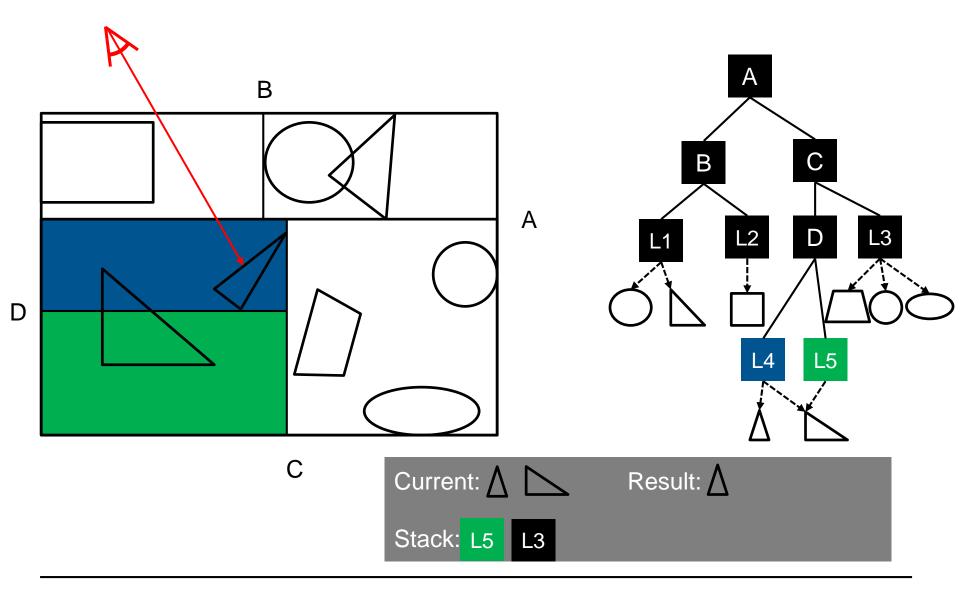
# kD-Tree Traversal (7)



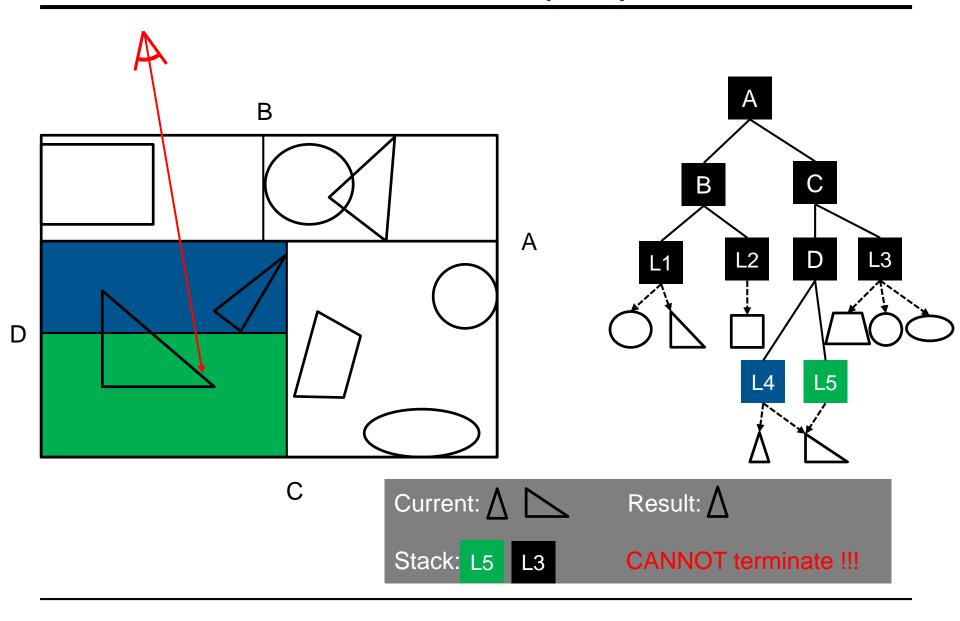
# kD-Tree Traversal (8)



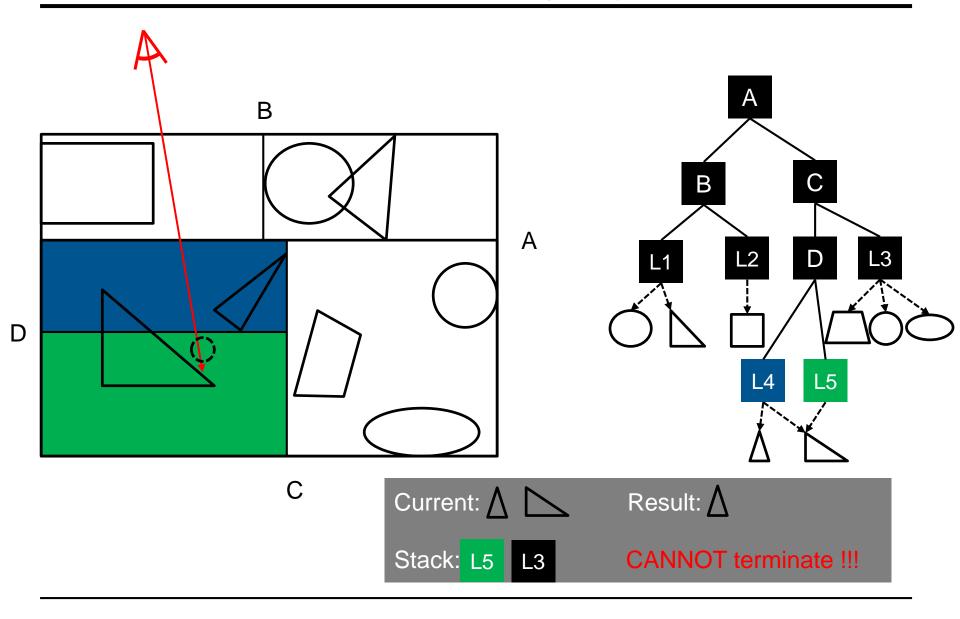
## kD-Tree Traversal (9)



## 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

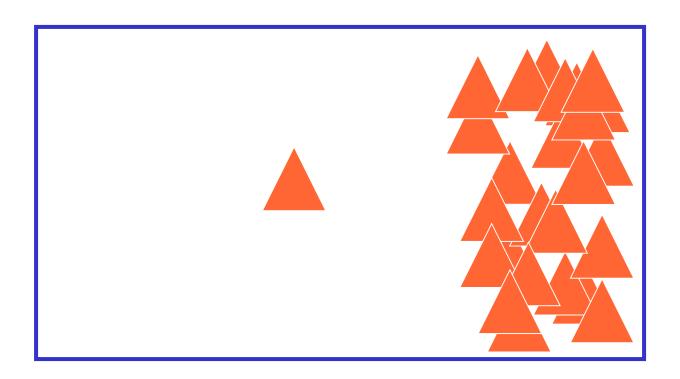
# Building good kD-trees

- 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)

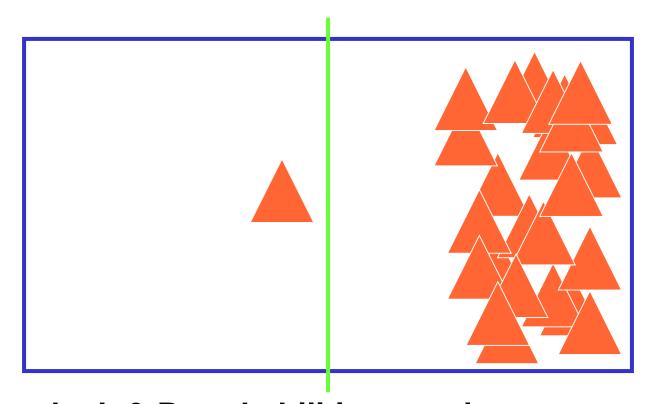
```
Cost(cell) = C_trav + Prob(hit L) * Cost(L) + Prob(hit R) * Cost(R)
```

- 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

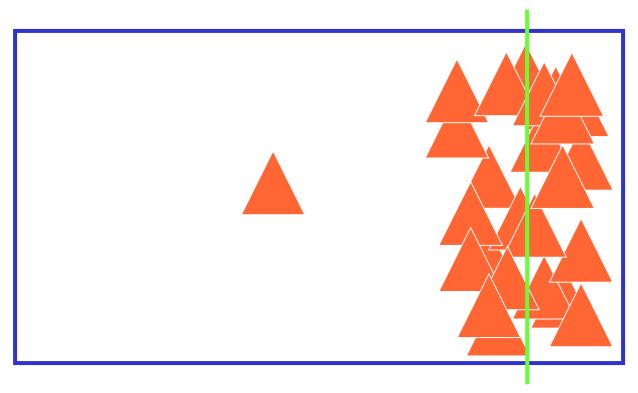


## Split in the middle



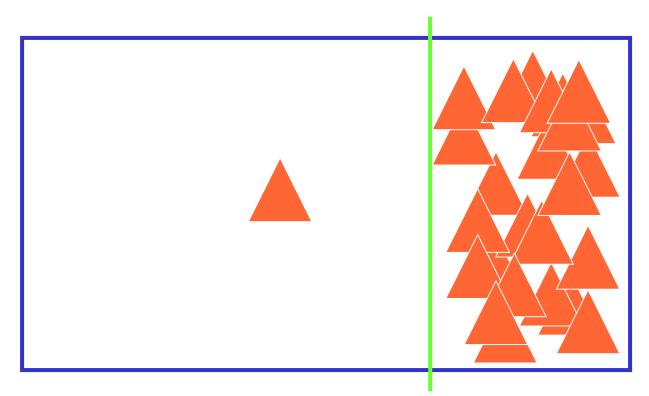
- Makes the L & R probabilities equal
- Pays no attention to the L & R costs

## 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

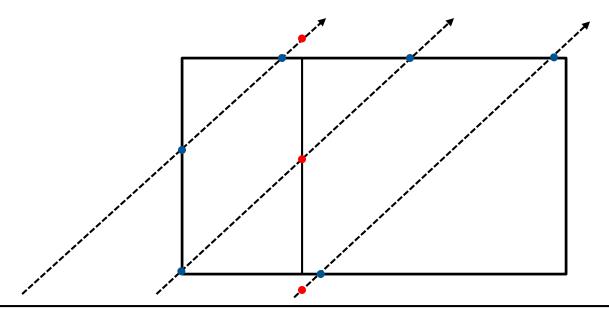
- Maintain on a stack
  - Entry and exit distance to node (t\_near and t\_far)
- Three cases

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

- t\_near < t\_split < t\_far Go to both (use stack)</p>

t\_split < t\_near</li>Go 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)</pre>
         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

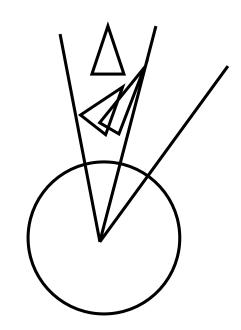
# **Directional Partitioning**

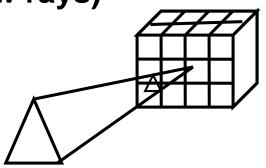
### Applications

- 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

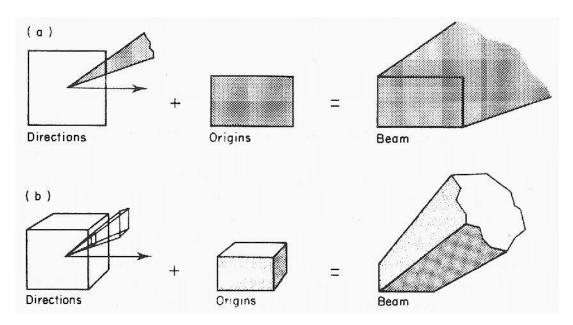




## Ray Classification

### 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

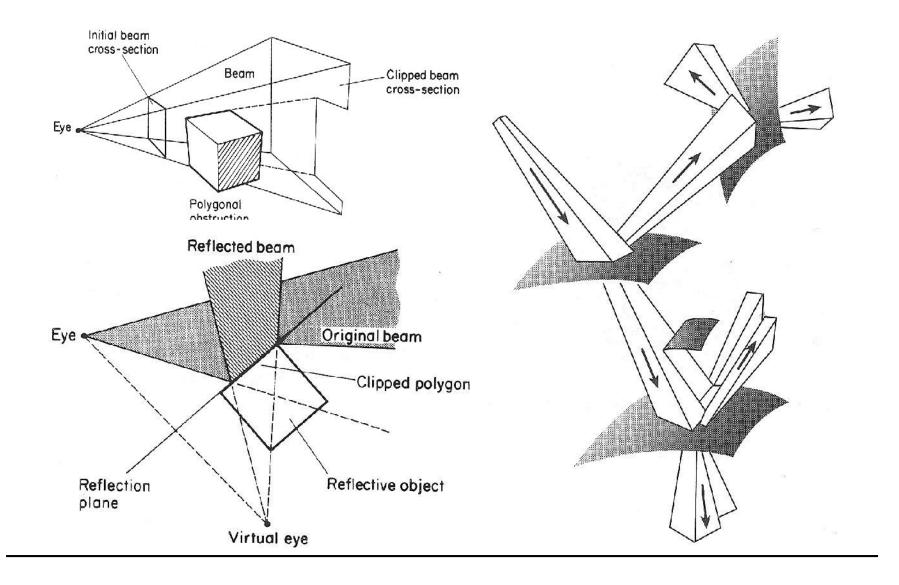


# **Packet Tracing**

### 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

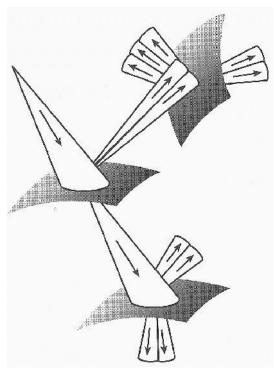
### Beam Tracing:

- 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 !!

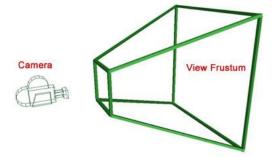


# Frustum Tracing

- 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

