

### Quadtrees and R-trees

A data simple data structure for geometric objects(e.g. points, houses, an image, 3D scene)

Support efficiently a very wide variety of queries.

Hierarchical Partition of the scene



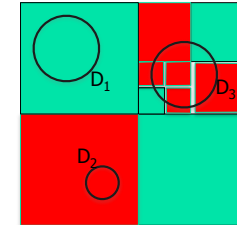
## Quadtrees

Assume we are given a red/green picture defined on a  $2^h \times 2^h$  grid. E.g. pixels.

Each pixel is either **green** or **red**.

(more general and interesting examples – soon)

Need to represent the shape “compactly”

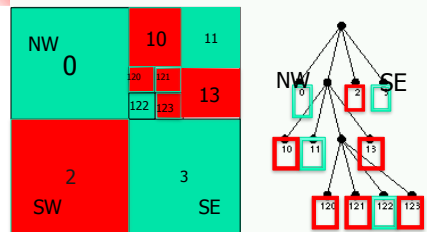


Need a data structure that could answers multiple types of queries. For example:

1. For a given point  $q$ , is  $q$  **red** or **green** ?
2. For a given query disk  $D$ , are there any green points in  $D$  ?
3. How many green points are there in  $D$  ?
4. Etc etc

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## Quadtrees



- Assume we are given a red/ green picture defined on a  $2^h \times 2^h$  grid of **pixels**.
- Each pixel has as a unique color (**Green** or **Red**)
- Every node  $v \in T$  is associated with a **geometric region**  $R(v)$

**Alg constructQT for a shape  $S$ .**

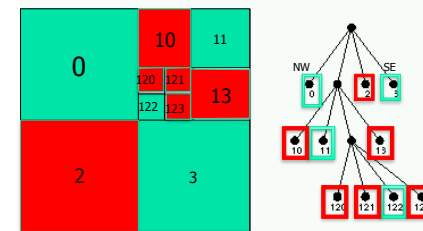
• **input** – a node  $v \in T$ , and a shape  $S$ .

• **Output** – a Quadtree  $T_v$  representing the shape of  $S$  within  $R(v)$ .

- If  $S$  is fully **green** in  $R(v)$ , or  $S$  is fully **red** in  $R(v)$  – then
  - $v$  is a leaf, labeled **Green** or **Red**. Return ;
- Otherwise, divide  $R(v)$  into 4 equal-sized quadrants, corresponding to nodes  $v.NW$ ,  $v.NE$ ,  $v.SW$ ,  $v.SE$ .
- Call **constructQT** recursively for each quadrant.

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## Quadtrees



Consider a picture stored on an  $2^h \times 2^h$  grid. Each pixel is either **red** or **green**.

We can represent the shape “compactly” using a QT.

Height – at most  $h$ .

Point location operation – given a point  $q$ , is it black or white

- takes time  $O(h)$
- could it be much smaller ?

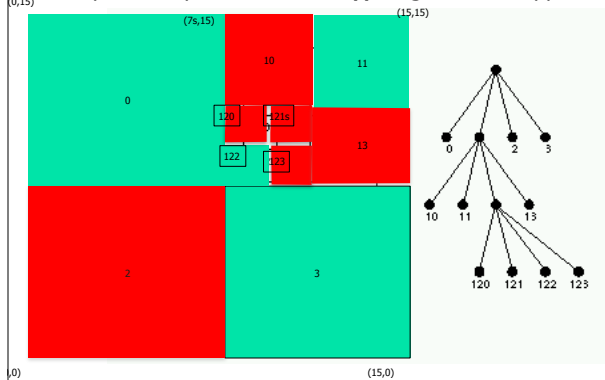
Many other operations are very simple to implement.

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## Storing the range $R(v)$ of a node

Each node  $v$  is associated with a range  $R(v)$  – a square. The node  $v$  stores (in addition to other info) 4 values

(MinX,MinY) – coordinates of the **lower left** corner of  $R(v)$   
 (MaxX,MaxY) coordinates of the **upper right** corner of  $R(v)$

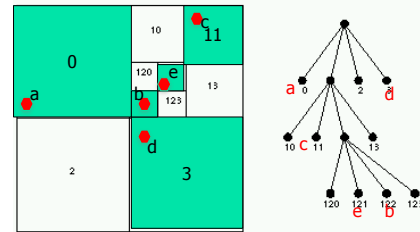


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## QuadTree for a set of points

Now consider a set of points (red) but on a  $2^h \times 2^h$  grid.

Splitting policy: Split until each quadrant contains  $\leq 1$  point.



Build a similar QT, but we stop splitting a quadrant when it contains  $\leq 1$  point (or some other small constant)

Point location operation – given a point  $q$ , is it black or white  
 – takes time  $O(h)$  (and less in practice)

Many other splitting policies are very simple to implement.  
 (eg. A leaf could contain **contains**  $\leq 17$  points)

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## Regions of nodes

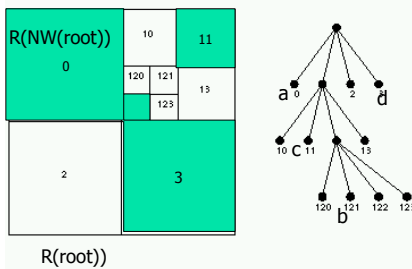
In general, every node  $v$  is associated with a region  $R(v)$  in the plane

$R(\text{root})$  is the whole region

The smallest area of  $R(v)$  is a single pixel.

Let  $NW(v)$  denote the North West child of  $v$ . (similarly  $NE$ ,  $SW$ ,  $SE$ )

$R(v)$  = is the union of  
 $R(NW(v))$ ,  $R(NE(v))$ ,  $R(SW(v))$ ,  $R(SE(v))$



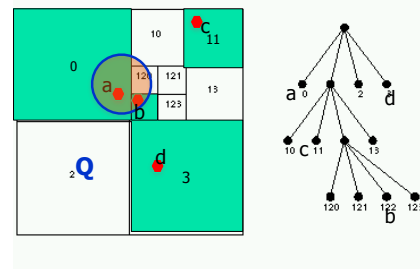
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## QuadTrees for a set of points

Report( $Q, v$ )

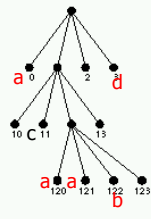
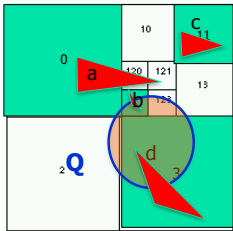
//  $Q$  – a query disk  
 /\*report all the points in stored at the subtree rooted at  $v$ , which are also inside  $Q$ . \*/

1. If  $v$  is NULL – **return**.
2. If  $R(v)$  is disjoint from  $Q$  – **return**
3. If  $R(v)$  is fully contained in  $Q$  – report all points in the subtree rooted at  $v$ .
4. If  $v$  is a leaf – check each point in  $R(v)$  if inside  $Q$
5. Else
  - ◆ Report( $Q$ ,  $NW(v)$ )
  - ◆ Report( $Q$ ,  $NE(v)$ )
  - ◆ Report( $Q$ ,  $SW(v)$ )
  - ◆ Report( $Q$ ,  $SE(v)$ )



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## QuadTrees for shape



Input: Set  $S$  of triangles  
 $S = \{t_1, \dots, t_n\}$

Splitting policy: Split  
 quadrant if it intersects  
 more than 1 triangle of  $S$ .

**Note** – a triangle might be stored in multiple leaves.  
 Some leaves might store no triangles.

Finding all triangles inside a query region  $Q$  –  
 essentially same Report Report( $Q, v$ ) as before  
 (minor modifications)

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## Inserting a new segment

```

insert(segment s, node v) {
    // Inserting a segment s into an existing node v of QT
    // v might or might not be a leaf
    If v is NULL - Error
    If R(v) is disjoint from s – Return. Else
    If v is not a leaf, then for each child u of v, call insert(s, u);
    Else // v is a leaf
        Add s to v.SegmentsList
    If number of segments in v.SegmentsList too long (e.g. >3) Call Split(v)
}
    
```

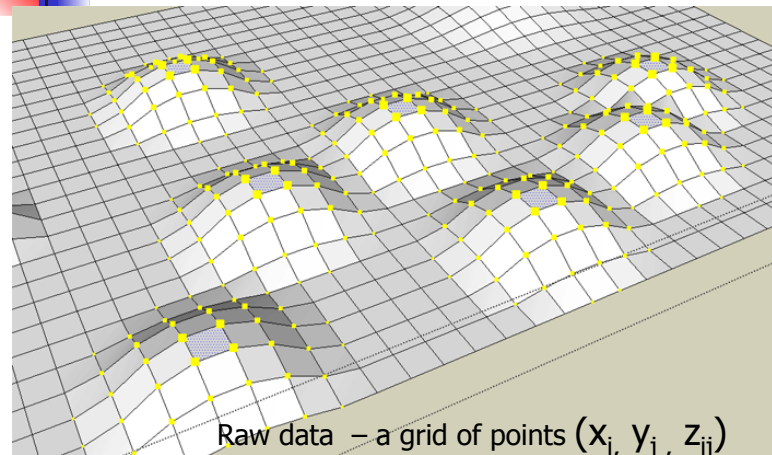
```

Split(v){
    // Assumption – v is a leaf, but has too many segments in its list
    // Create 4 children for v (make sure they know which region they cover.)
    For each child u of v
        For each segment s in v.SegmentsList Call insert(s, u)
    Empty v.SegmentsList
}
    
```

Material from this slide is optional for CSs345

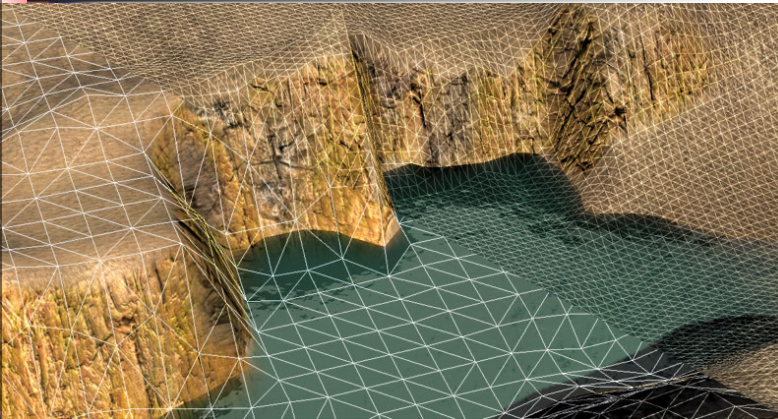
Raw data – a grid of points  $(x_i, y_j, z_{ij})$   
 For every grid point  $i, j$

## Terrain representations



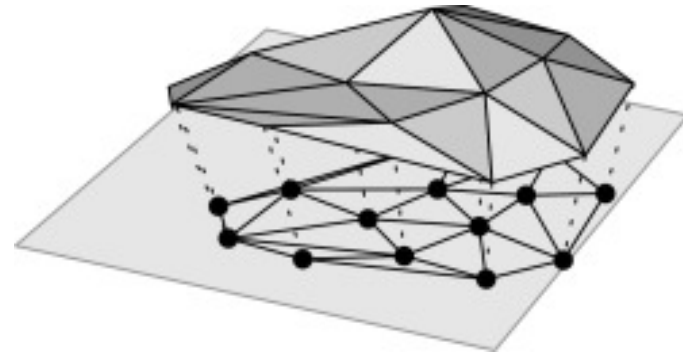
Raw data – a grid of points  $(x_i, y_j, z_{ij})$   
 For every grid point  $i, j$

## Triangulated terrain (TIN – Triangulated irregular network)



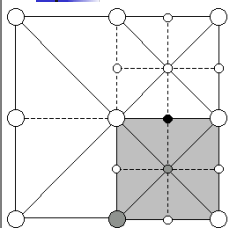
Each triangle approximately fits the surface below it

## How to find good triangulation ?



Each triangle approximately fits the surface below it  
(credit SCALGO)

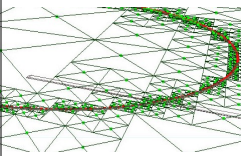
## How to find good triangulation ?



- ◆ Input – a very large set of points  $S = \{ (x_i, y_i, z_{ij}) \}$ .
- ◆  $z_{ij}$  is the elevation at point  $(x_i, y_i)$
- ◆ Want to create a surface, consists of triangles, where each triangle interpolates the data points underneath it.
- ◆ Idea: Build a QT  $T$  for the 2D points.
- ◆ (if want triangles: Each quadrant is split into 2 triangles)
- ◆ Assign to each vertex the height of the terrain above it.
- ◆ The approximated elevation of the terrain at any point is the linear interpolation of its elevated vertices.

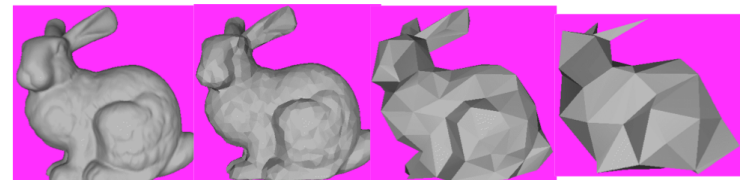
**QT Split Policy:** Splitting a quadrant into 4 sub-quadrants:

- ◆ split a node  $v$  if for some data point  $(x_i, y_i) \in R(v)$ , the elevation of  $z_{ij}$  is too far from the corresponding triangle. If not, leave  $v$  as a leaf.
- ◆ That is,  $(x_i, y_i, z_{ij})$  it is too far from the interpolated elevation.
- ◆ **Note:** A quadrant might contain a huge number of points, but they behave smoothly. E.g. all a the slope of a mountain, but this slope is more or less linear.



## Level Of Details

- Idea – the same object is stored several times, but with a different level of details
- Coarser representations for distant objects
- Decision which level to use is accepted 'on the fly'  
(eg in graphics applications, if we are far away from a terrain, we could tolerate usually large error)



69,451 polys

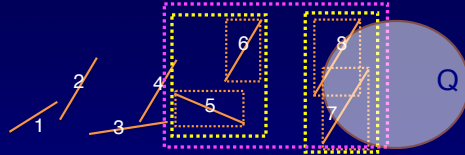
2,502 polys

251 polys

76 polys

## R-trees

- Input: A set  $S$  of shapes (segments in this example)
- Prepare a tree that could assist finding the segments intersecting a query region.
- Fewer theoretical guarantees, but extremely useful.

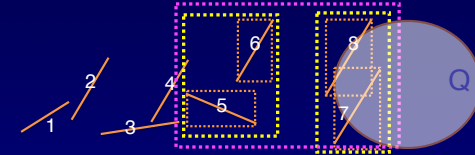


- We compute for each segment its bounding bounding box (rectangle).
  - These are the leaves of  $T$
  - Match pairs of bounding boxes. For example 1-2, 3-4, 5-6, 7-8. For each such pair, compute their bounding boxes. Each node in level 2 is such a box.
  - Match these bounding boxes. These are the nodes of level 3.
  - In general for every node  $v$ ,  $BB(v) = BB(BB(v.right) \cup BB(v.left))$
- Once a query region  $Q$  is given we determine whether it intersects  $BB(\text{root})$   
 If not, we are done. If yes, check recursively if  $Q$  intersects  $BB(v.left)$  and  $BB(v.right)$



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- Answer: Mostly **Simplicity** in computation of intersection.
- R-trees are very useful also in higher dimensions.
- Other big question" Which pairs to match. Obviously closer is better. But many variants Multiple heuristics

- Problem in reporting: Many "false alarms" : BBs that intersect  $Q$  while their segments don't.
- Should we use axis parallel BB instead of something that could "snag" better? For example, rotated rectangles ?

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- Other big question" Which pairs to match. Obviously closer is better. But many variants Multiple heuristics