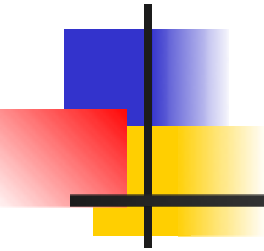


## Quadtrees and R-trees

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

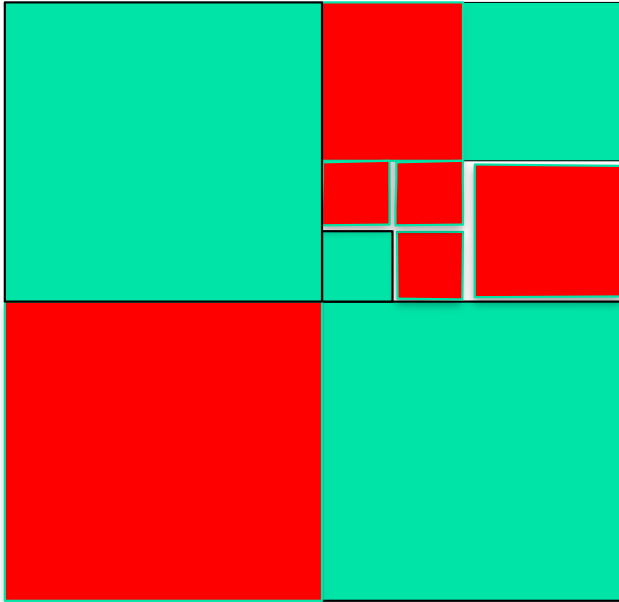
Support efficiently a very wide variety of queries.

Hierarchical Partition of the scene





# QuadTrees



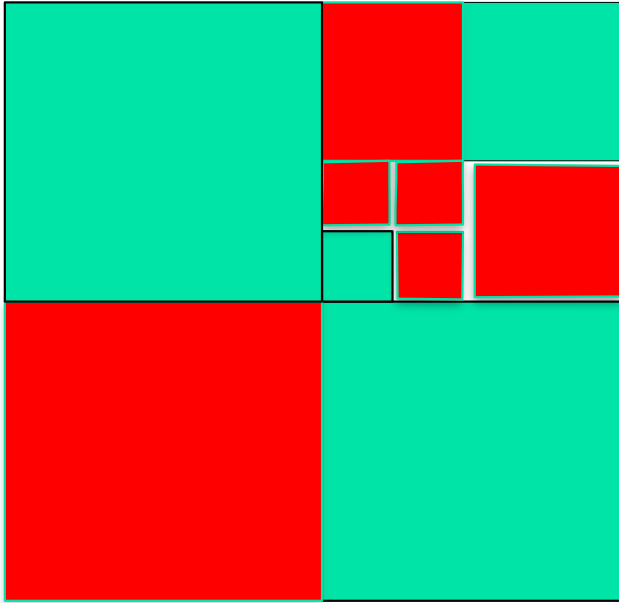
Assume we are given a red/green picture defined a  $2^h \times 2^h$  grid. E.g. pixels.

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

(more general and interesting examples – soon)

Need to represent the shape “compactly”

# QuadTrees



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

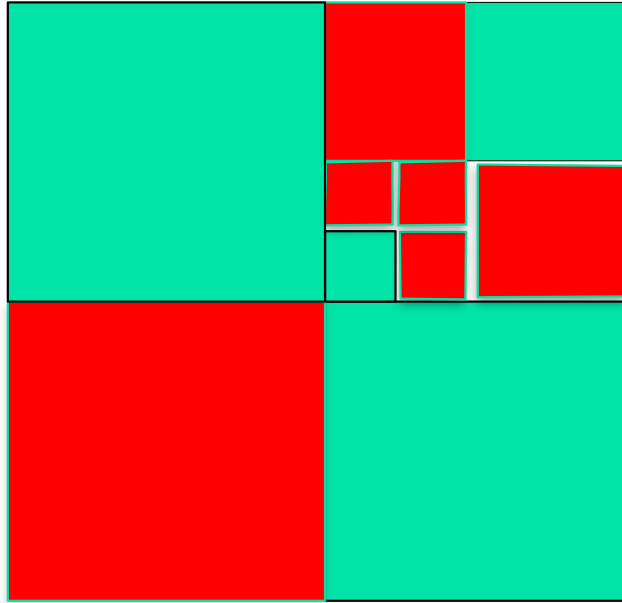
Each pixel is either **green** or **red**.

(more general and interesting examples – soon)

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

# QuadTrees



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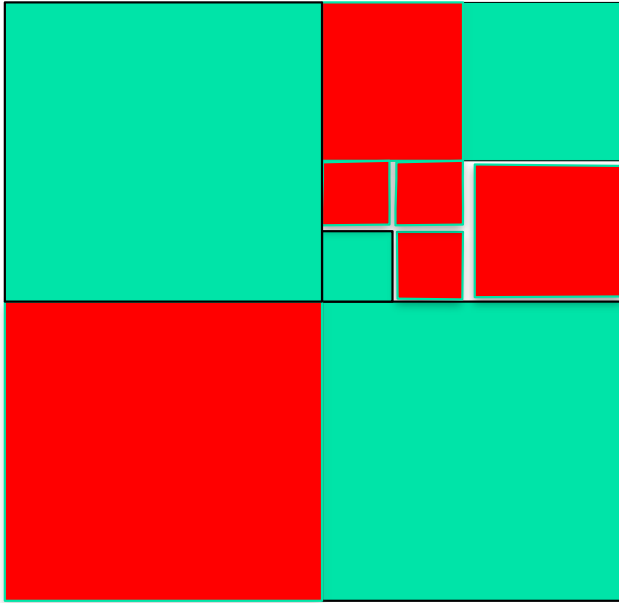
(more general and interesting examples – soon)

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1. For a given point  $q$ , is  $q$  **red** or **green** ?

# QuadTrees



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

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

(more general and interesting examples – soon)

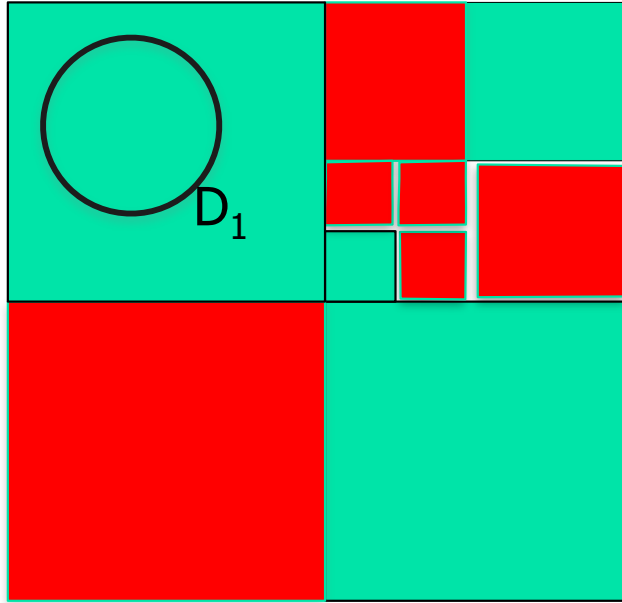
Need to represent the shape “compactly”

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  $D$  ?

# QuadTrees



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

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

(more general and interesting examples – soon)

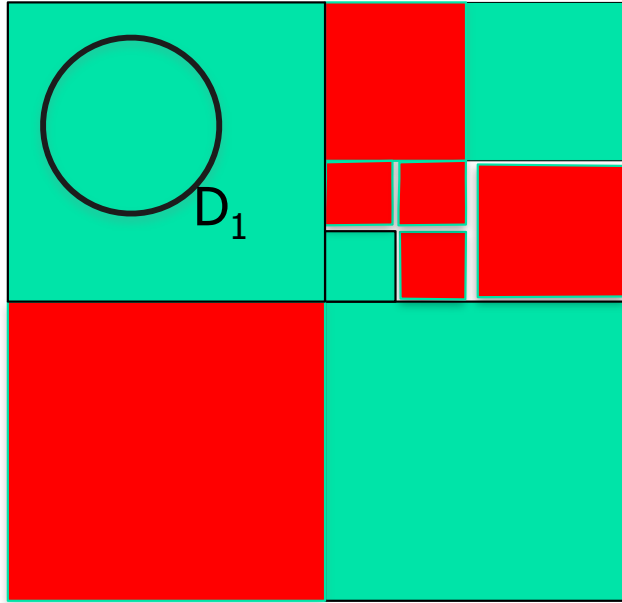
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(more general and interesting examples – soon)

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

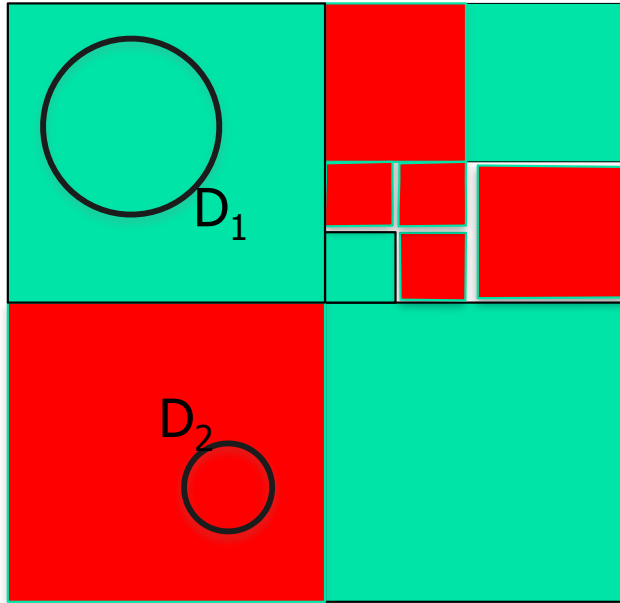
1. For a given point  $q$ , is  $q$  **red** or **green** ?

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

# QuadTrees



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

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

(more general and interesting examples – soon)

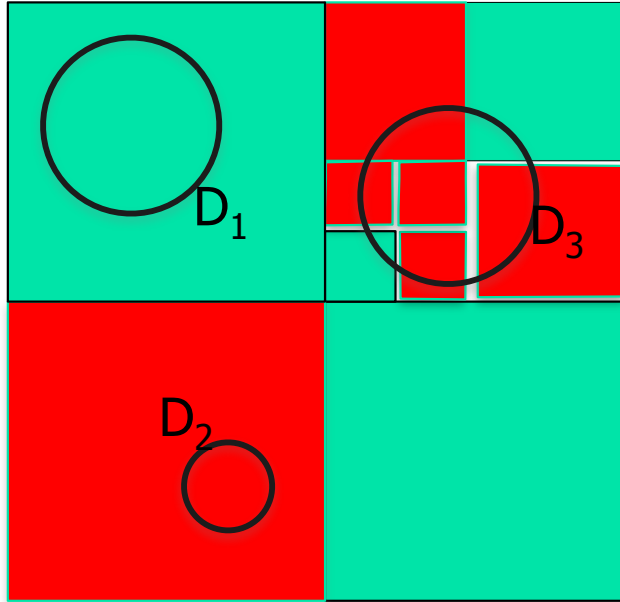
Need to represent the shape “compactly”

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



# QuadTrees



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

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

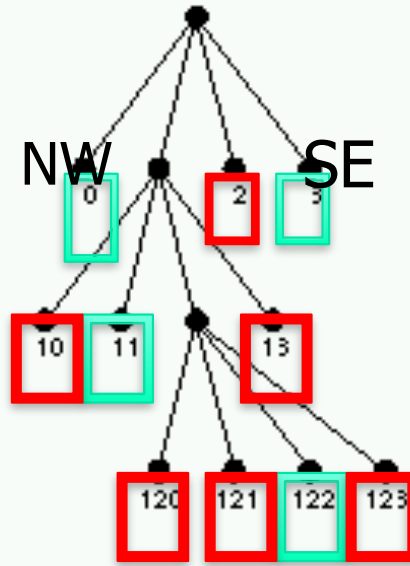
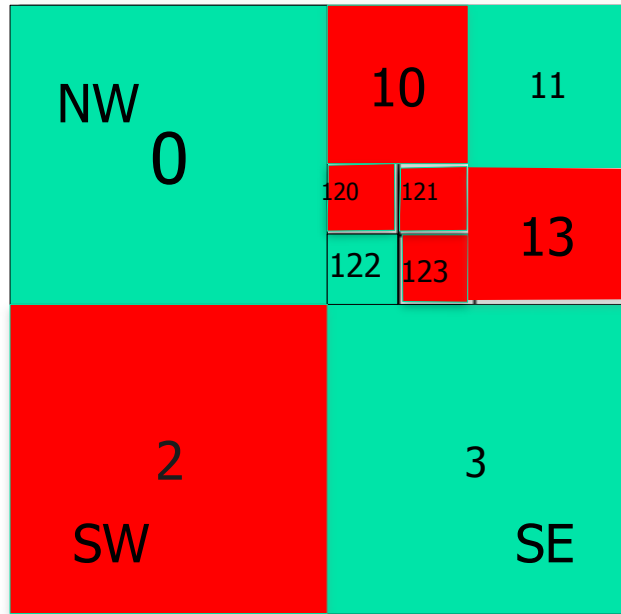
(more general and interesting examples – soon)

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** ?
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4. Etc etc

# QuadTrees

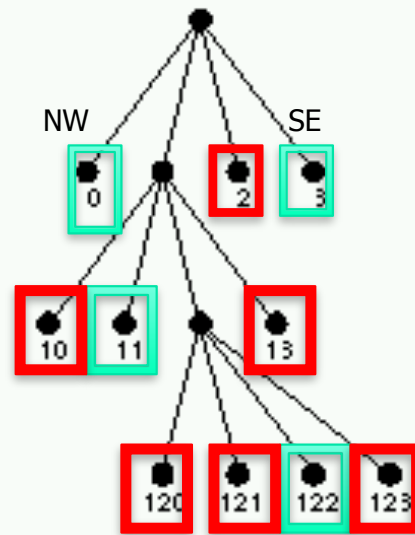
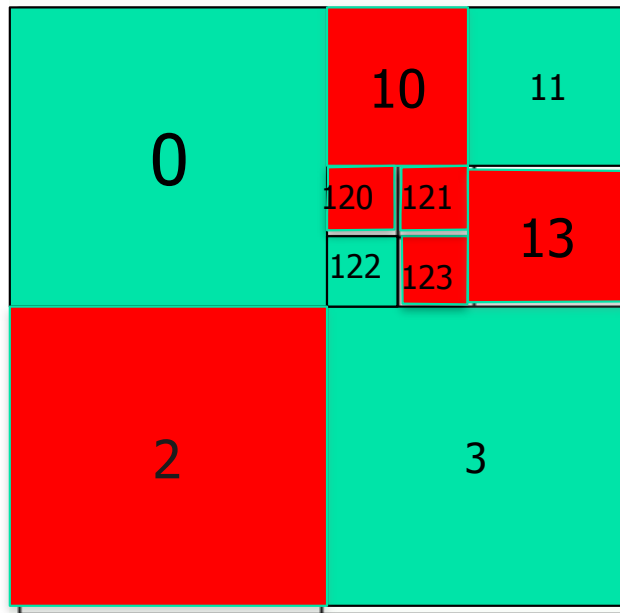


- Assume we are given a red/green picture defined on a  $2^h \times 2^h$  grid of **pixels**.
- Each pixel has as a unique color (**Green** or **Red**)
- Every node  $v \in T$  is associated with a geometric region  $R(v)$ .
- This is the region that  $v$  is “in charge of”.

**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 **ConstructOT** recursively for each quadrant.

# 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.

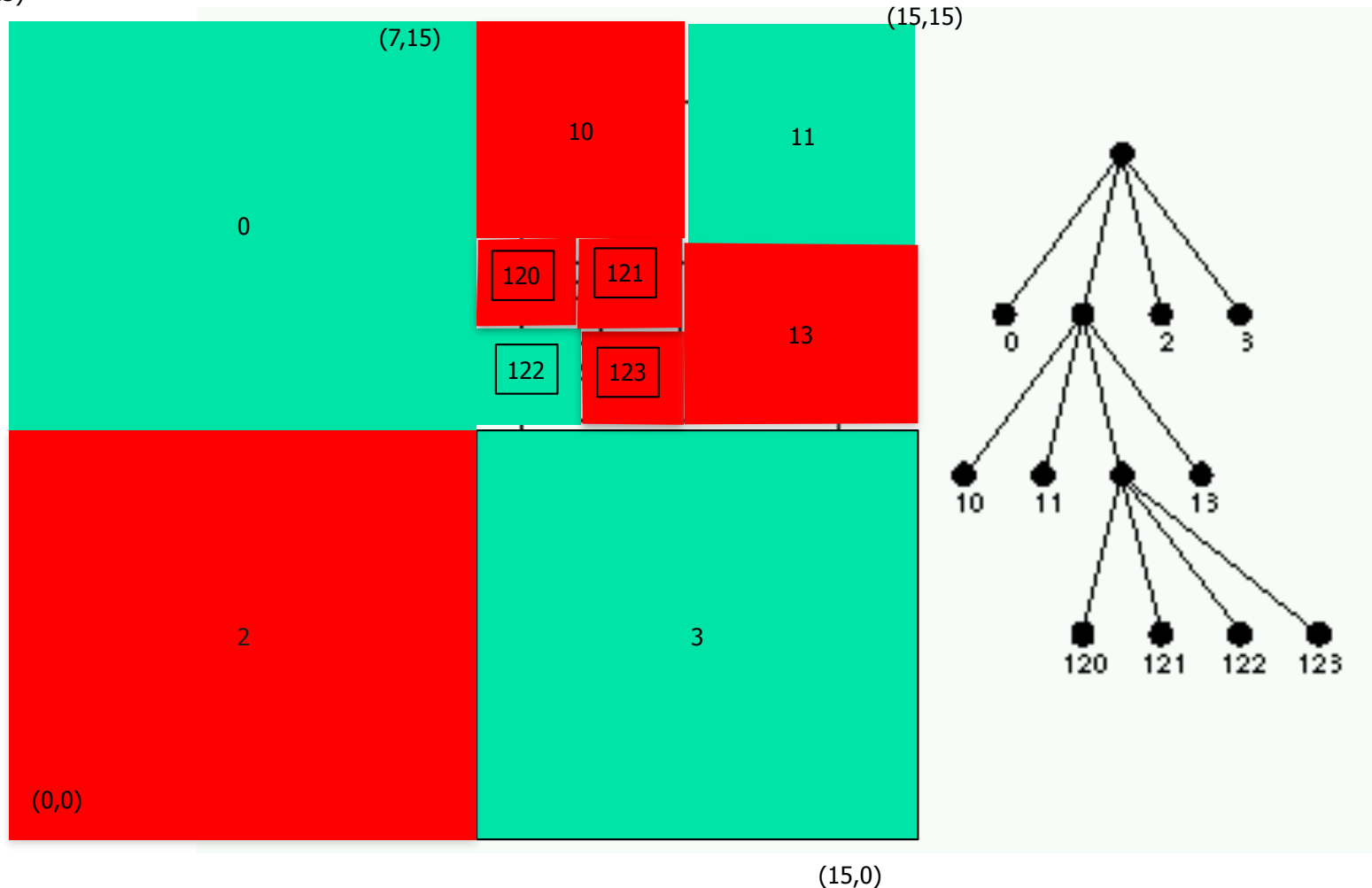
# Storing the **range** $R(v)$ of a node

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

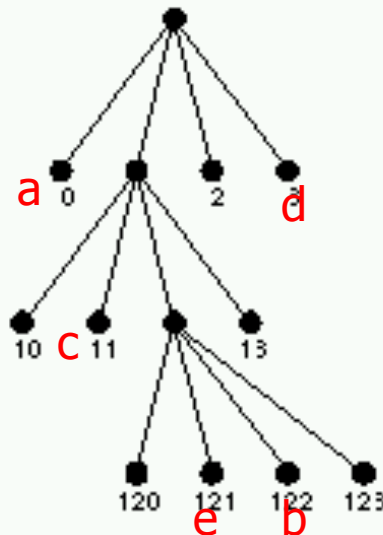
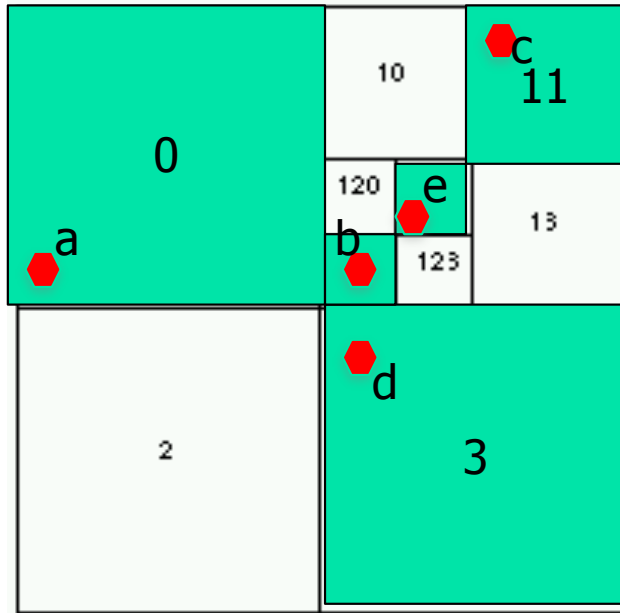
$(\text{MinX}, \text{MinY})$  – coordinates of the **lower left** corner of  $R(v)$


$(\text{MaxX}, \text{MaxY})$  coordinates of the **upper right** corner of  $R(v)$

$(0,15)$



# QuadTree for a set of points



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

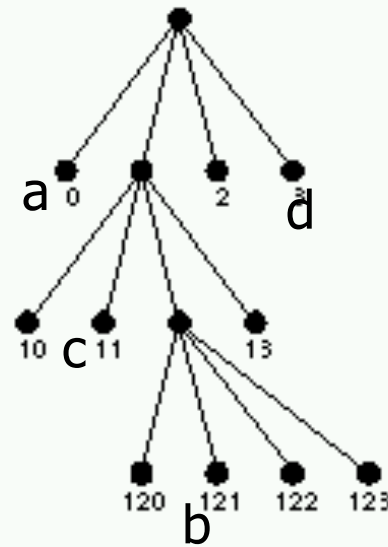
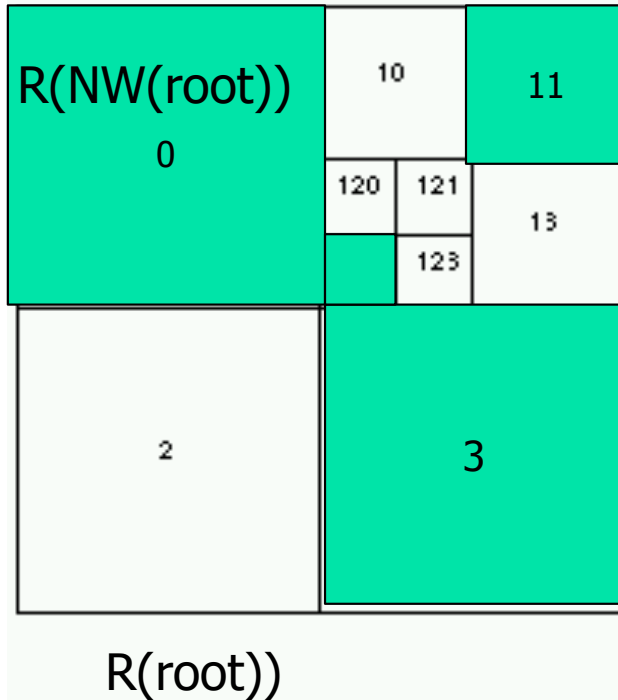
Splitting policy: Split until each quadrant contains  $\leq 1$  point.

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)

# Regions of nodes



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

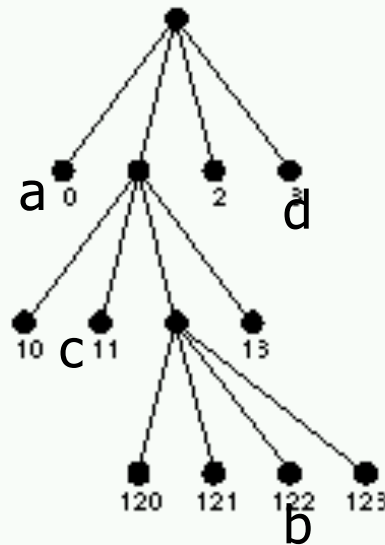
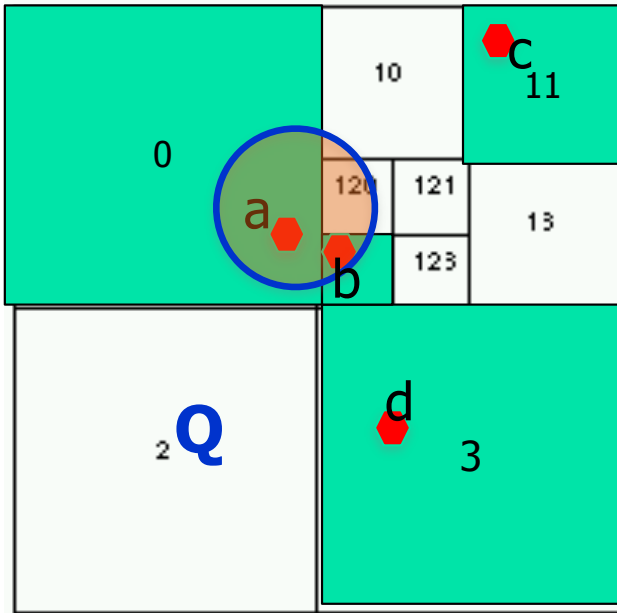
$R(\text{root})$  is the whole region

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  $R(NW(v))$ ,  $R(NE(v))$ ,  $R(SW(v))$ ,  $R(SE(v))$

# 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**

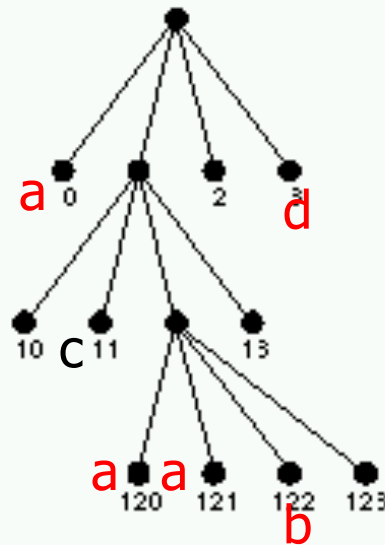
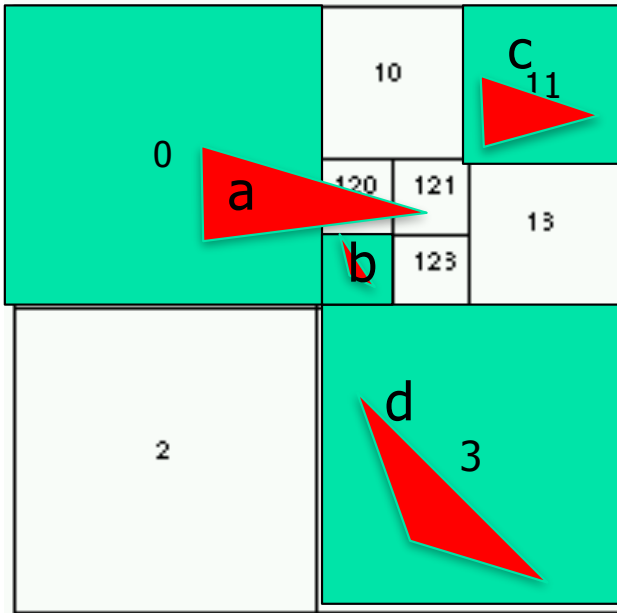
3.If  $R(v)$  is fully contained in  $Q$  – report all points in the subtree rooted at  $v$ .

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( $Q, SE(v)$ )

# QuadTrees for shape



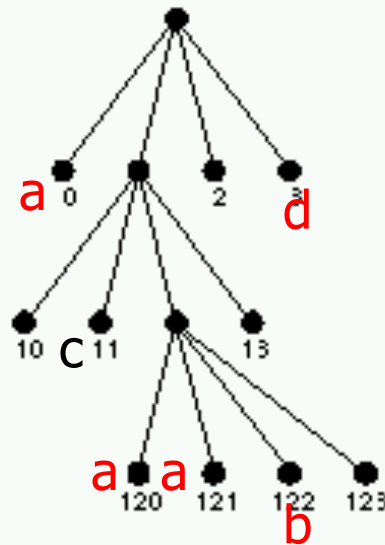
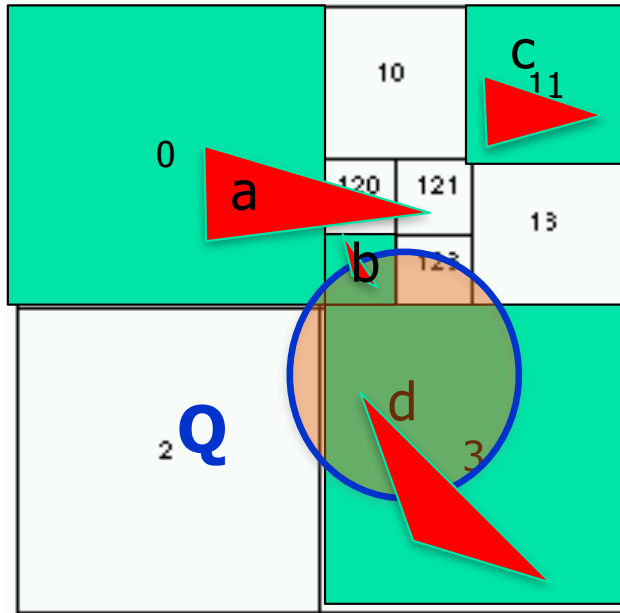
- Input: A set  $S$  of triangles  $S = \{t_1 \dots t_n\}$ .
- Each leaf  $v$  stores a list  $v.TriangleList$  of all triangles intersecting  $R(v)$ .
- Splitting policy: Split a quadrant if it intersects more than 5 (say) triangle of  $S$ .

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

Finding all triangles inside a query region  $Q$ . We essentially use the function  $Report(Q, v)$  from the previous slide (with minor modifications)



# QuadTrees for shape



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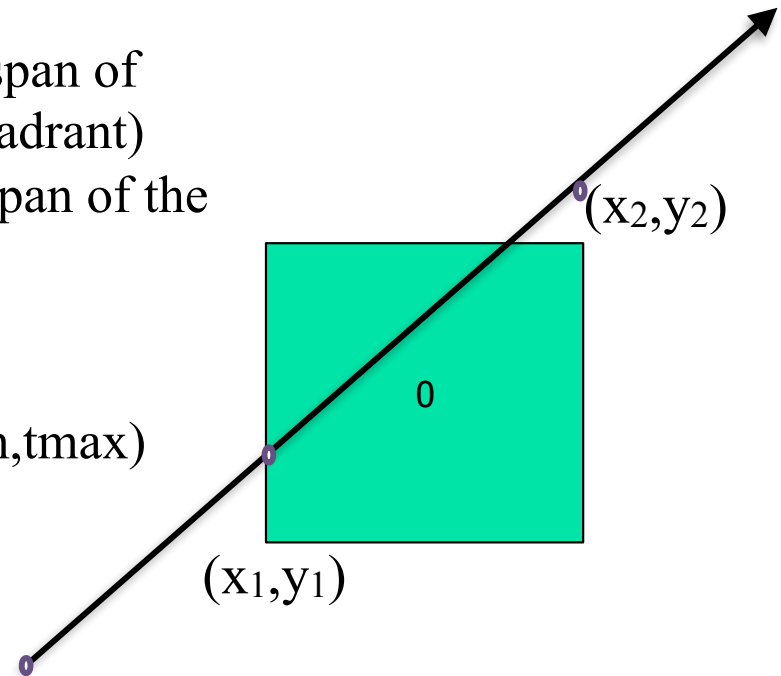
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# Ray tracing and QuadTrees

- Consider a quadrant with corners  $LL=(x_1,y_1)$  and  $UR=(x_2,y_2)$ .
- To find if a ray  $r=p+tv$  intersects this quadrant
  - Find  $tmin\_x, tmax\_x$ , where the ray is in the x-span of the quadrant (the vertical slab containing the quadrant)
  - Find  $tmin\_y, tmax\_y$ , where the ray is in the y-span of the quadrant
  - Set  $tmin=\max(tmin\_x, tmin\_y)$
  - Set  $tmax=\min(tmax\_x, tmax\_y)$
  - The ray is inside the quadrant only for  $t \in (tmin, tmax)$

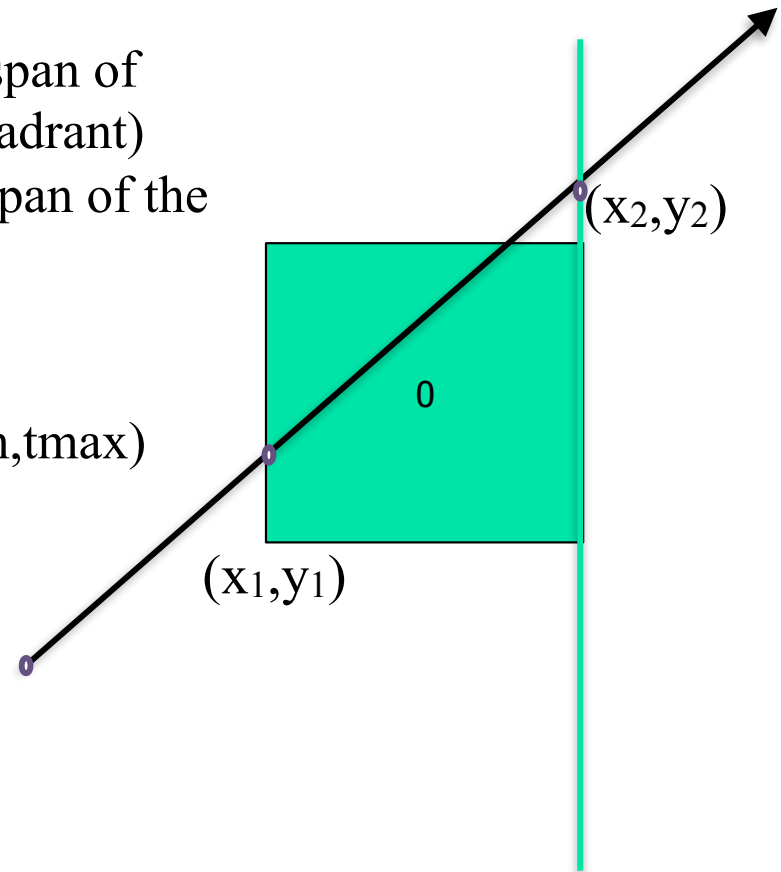
In 3D, we also check  $tmin\_z, tmax\_z$



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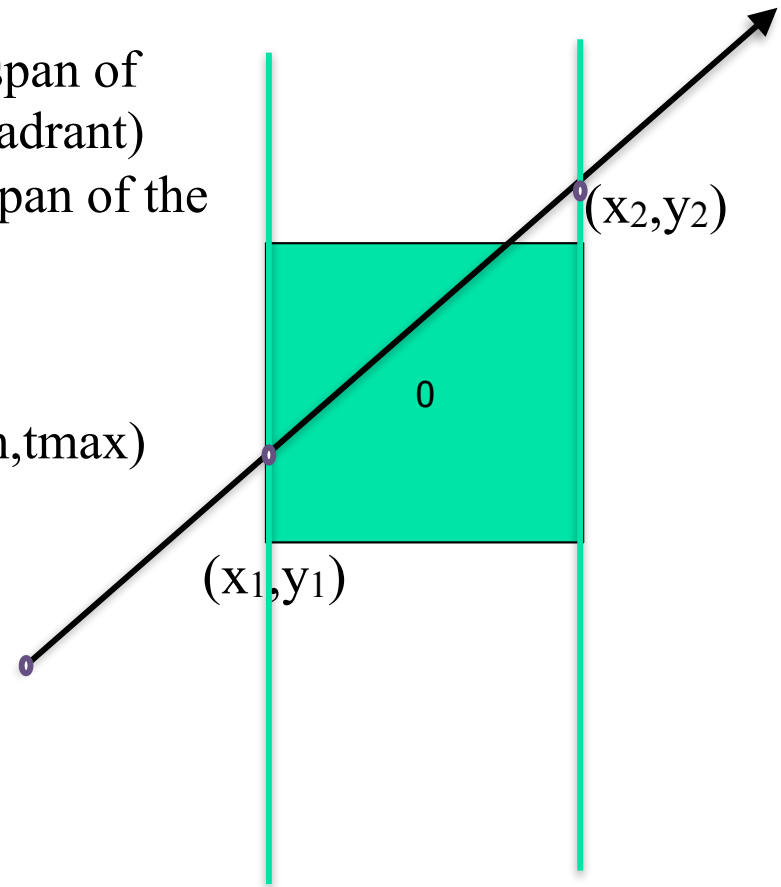
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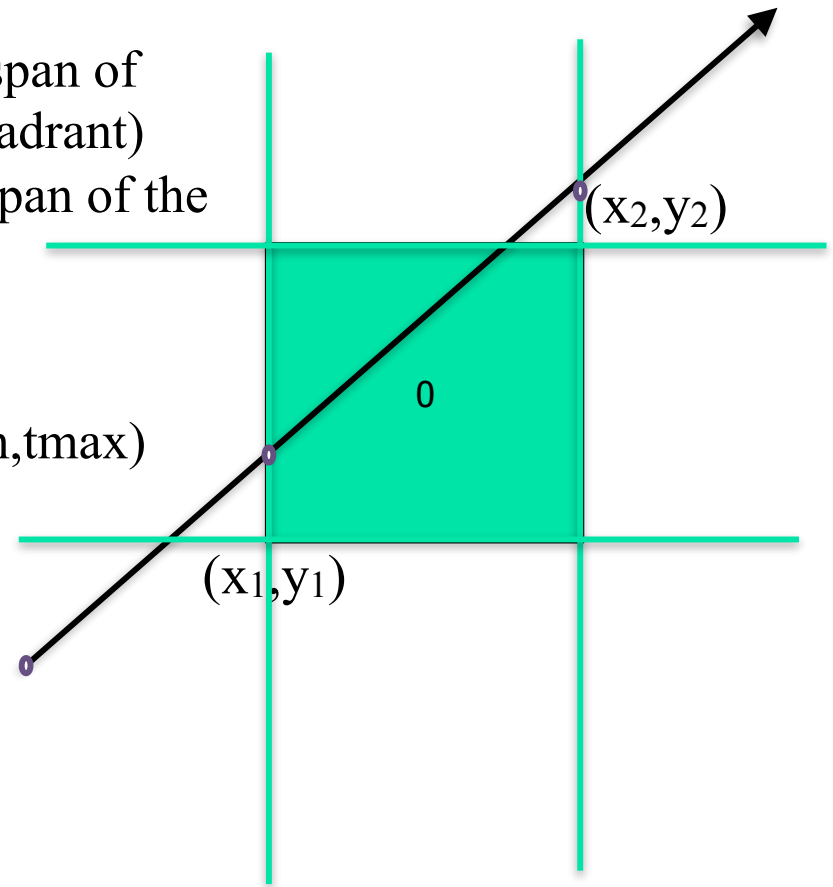
In 3D, we also check  $tmin\_z, tmax\_z$



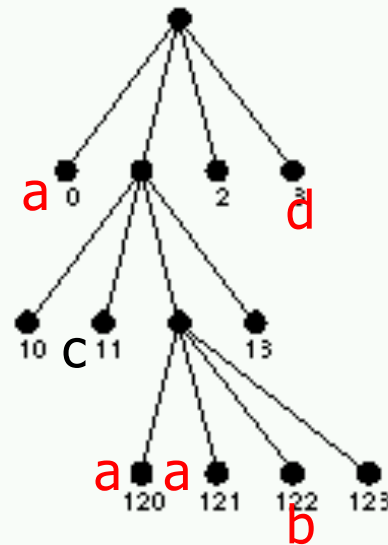
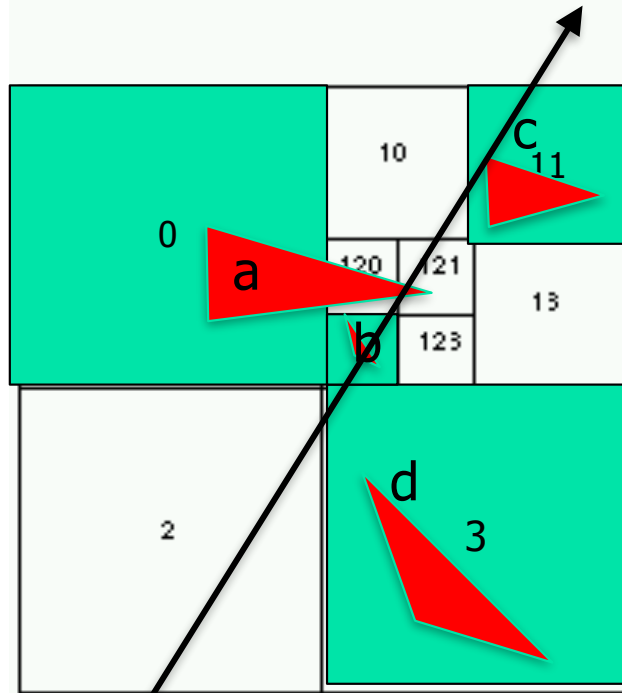
# Ray tracing and QuadTrees

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In 3D, we also check  $tmin\_z$ ,  $tmax\_z$

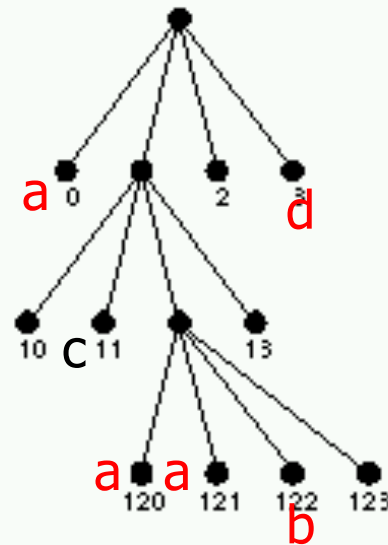
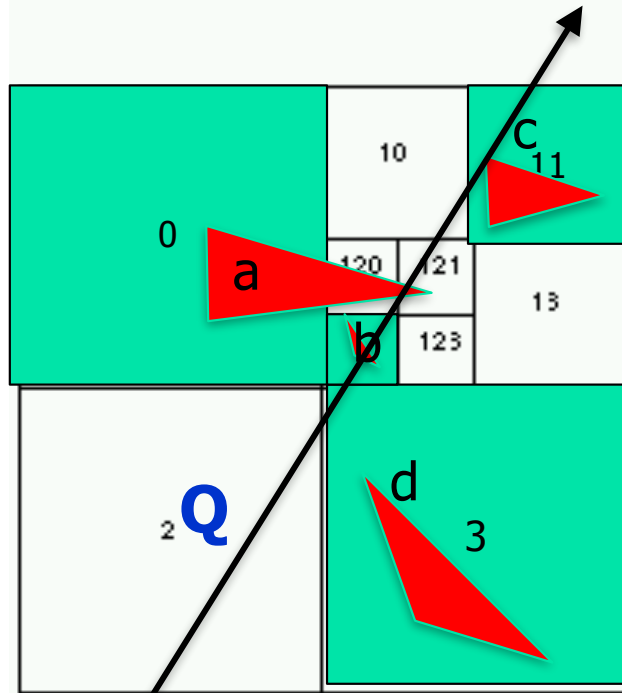


# Ray tracing and QuadTrees



- Now, it is easy to find the first triangle hit by a ray r:
- Start from  $v = \text{root}$ . If empty, then continue tracing the ray from the point it leaves the quadrant.
- If  $v$  is internal node, check which of its quadrants is first hit by  $r$ , and continue recursively.
- If  $v = \text{leaf}$ , check each triangle in  $v$

# Ray tracing and QuadTrees



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- Start from  $v = \text{root}$ . If empty, then continue tracing the ray from the point it leaves the quadrant.
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# Inserting a new triangle

---

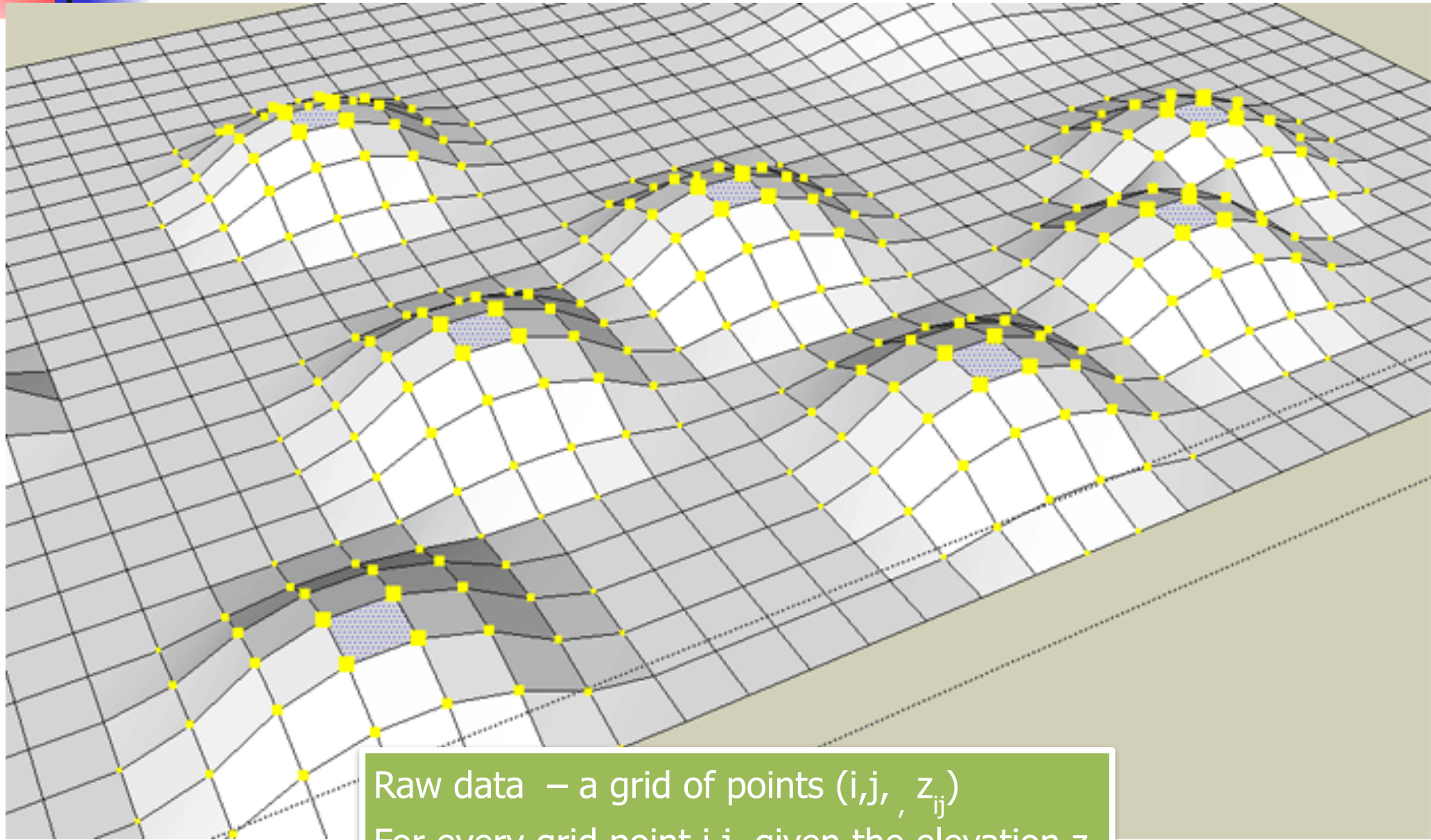
```
insert(triangle  $t_i$ , node  $v$ ) {  
    // Inserting a new triangle  $t_i$  into an existing node  $v$  of the Quadtree.  
    //  $v$  is not necessarily a leaf.  
    If  $v$  is NULL - Error  
    If  $R(v)$  is disjoint from  $t_i$  (share no points)– Return. Nothing to do.  
    If  $v$  is not a leaf, then for each child  $u$  of  $v$ , call insert( $t_i, u$ );  
    Else //  $v$  is a leaf  
        Add  $t_i$  to  $v$ .TrianglesList  
        If number of triangles in  $v$ .SegmentsList is too long (e.g.  $>5$ ) Call Split( $v$ )  
}
```

-----

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



# Terrain representations



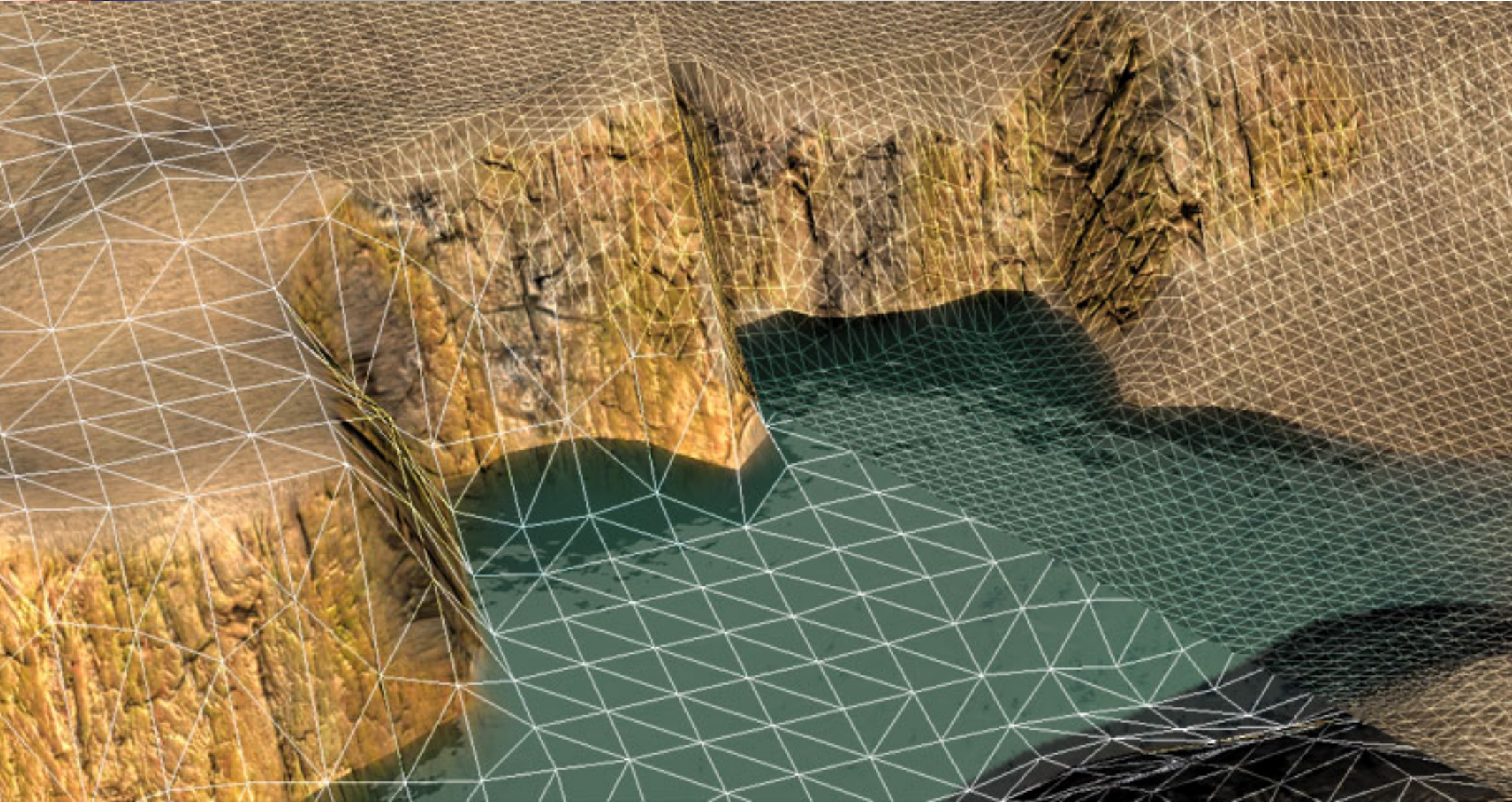
Raw data – a grid of points  $(i, j, z_{ij})$

For every grid point  $i, j$ , given the elevation  $z_{ij}$





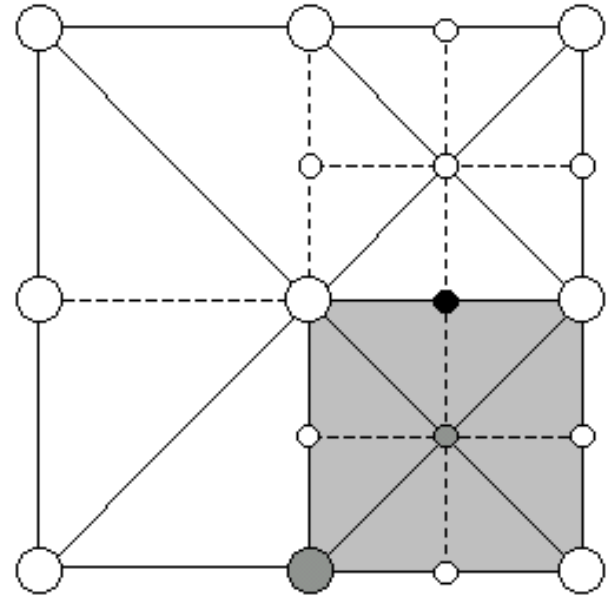
(TIN – Triangulated Irregular Network)



Each triangle approximately fits the surface below it



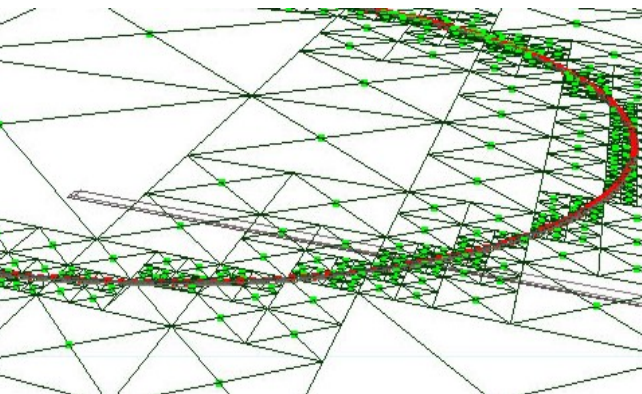
# How to find good triangulation ?



- ◆ Input – a very large set of points  $S = \{ (i,j, z_{ij}) \}$ .
- ◆  $z_{ij}$  is the elevation at point  $(i,j)$  (*latitude and longitude*)
- ◆ 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 right-hand triangles)
- ◆ Assign to each vertex the height of the terrain above it.
- ◆ The approximated elevation of the terrain at any point  $(x,y)$  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 the corresponding triangle. If not, leave  $v$  as a leaf.
- ◆ That is, for any point  $(i,j)$  on the plane, the elevation  $(i,j, z_{ij})$  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 sloop 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. E.g., sub pixels error are not noticeable.)



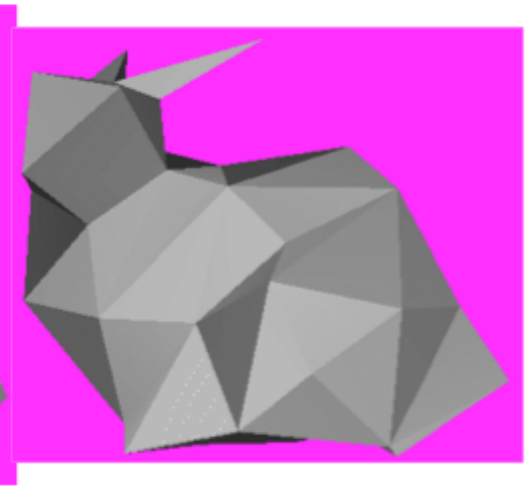
69,451 polys



2,502 polys



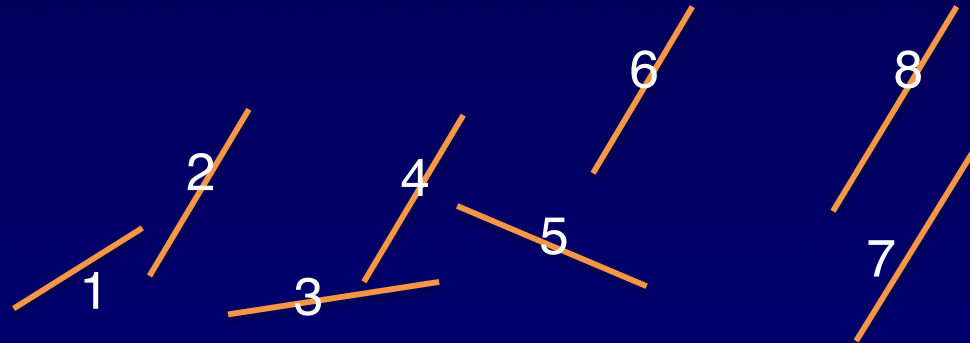
251 polys



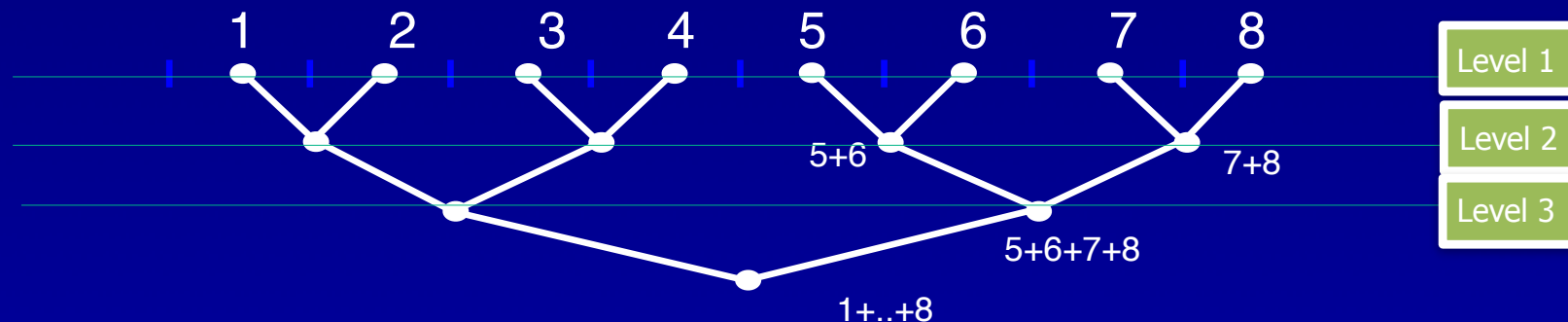
76 polys

# R-trees

- Input: A set  $S$  of shapes (segments in this example. Triangles in graphics apps)
- Build a tree that could expedite
  - (i) finding the segments intersecting a query region,
  - (ii) answering ray tracing
  - (iii) Emptiness queries. etc

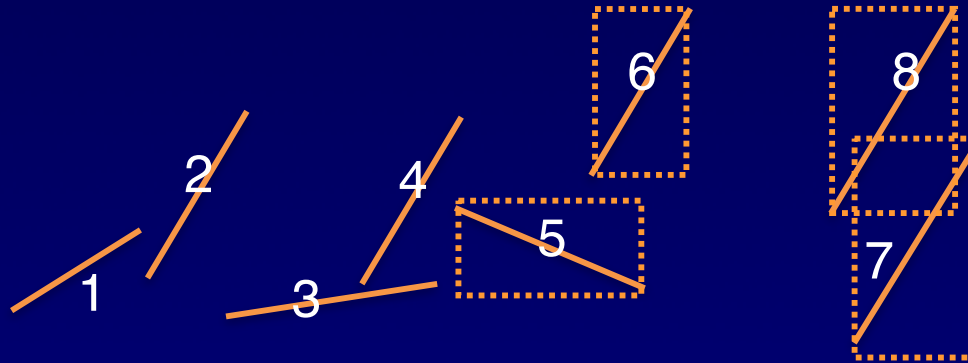


- We compute for each segment its bounding box (rectangle).
- These are the leaves of  $T$ . Call them "Level 1".
- Find the nearest pair of segments (say 7,8). Remove them from level 1, and replace them by a single BB encapsulate both. It corresponds to a node of level 2.
- Repeat until no vertex is left in level 1.
- Next, pick the nearest two BBs from level 2, and replace them by a vertex at level 3.
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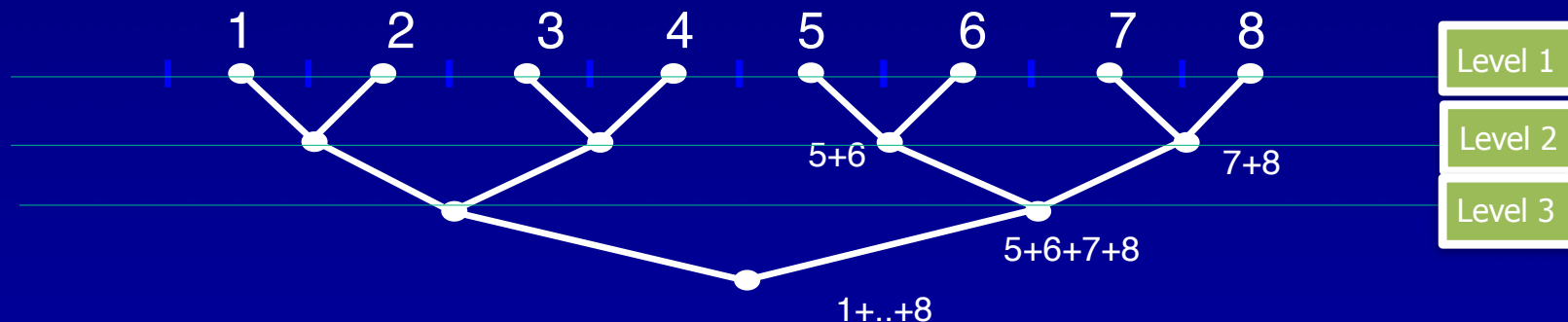


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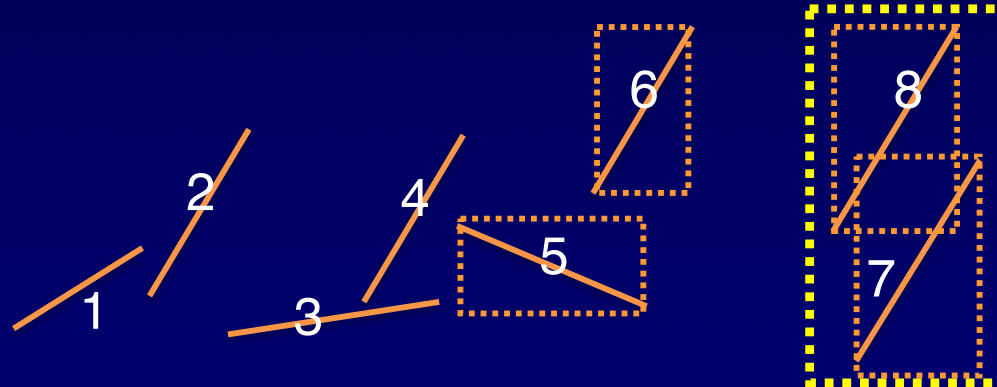


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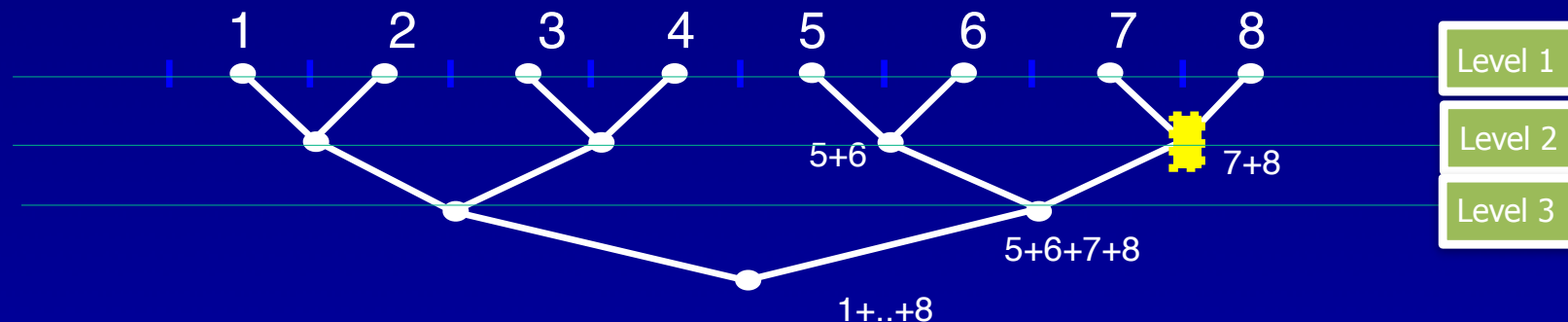


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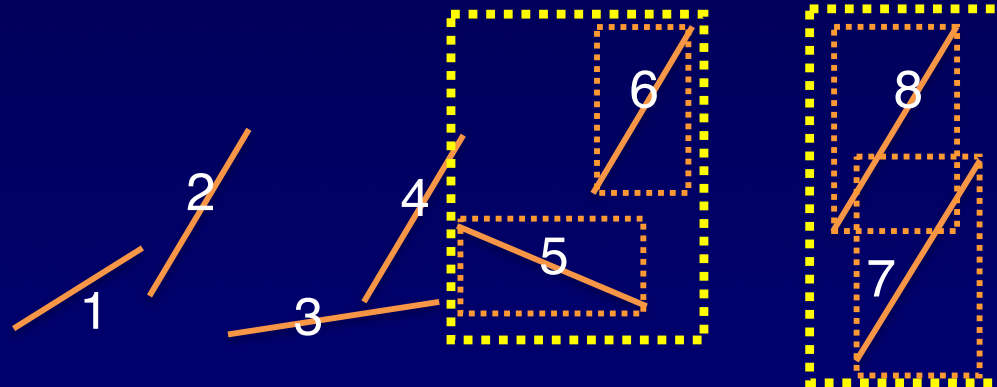


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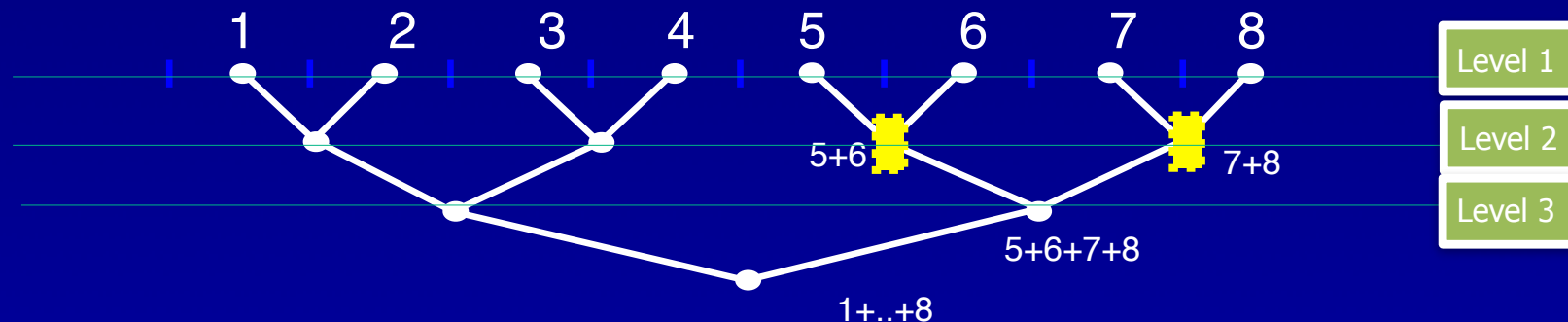


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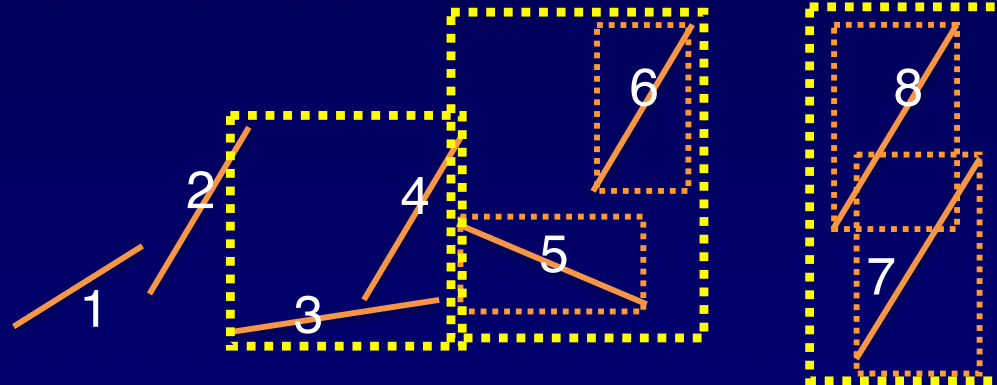
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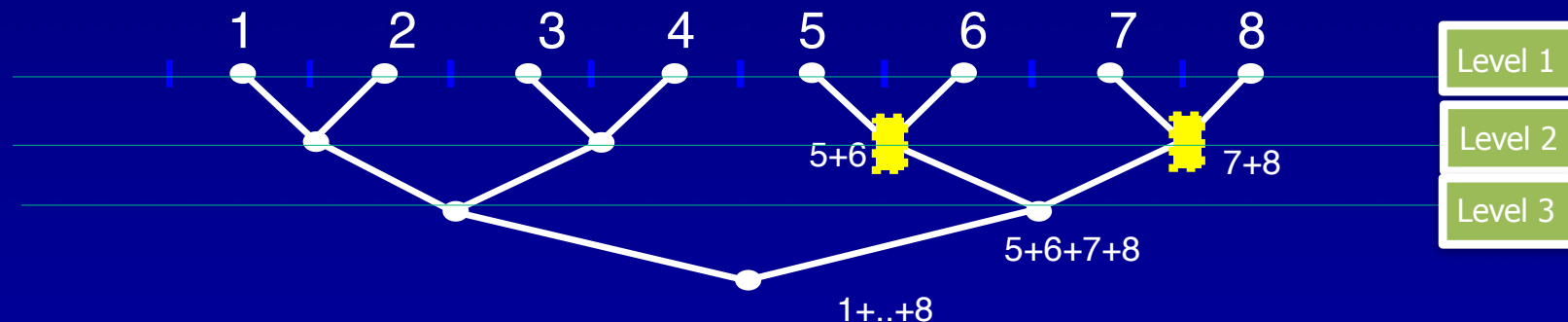


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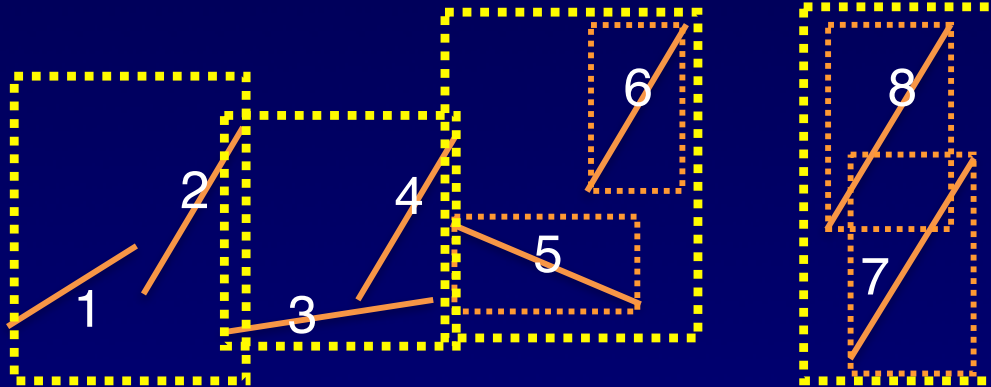


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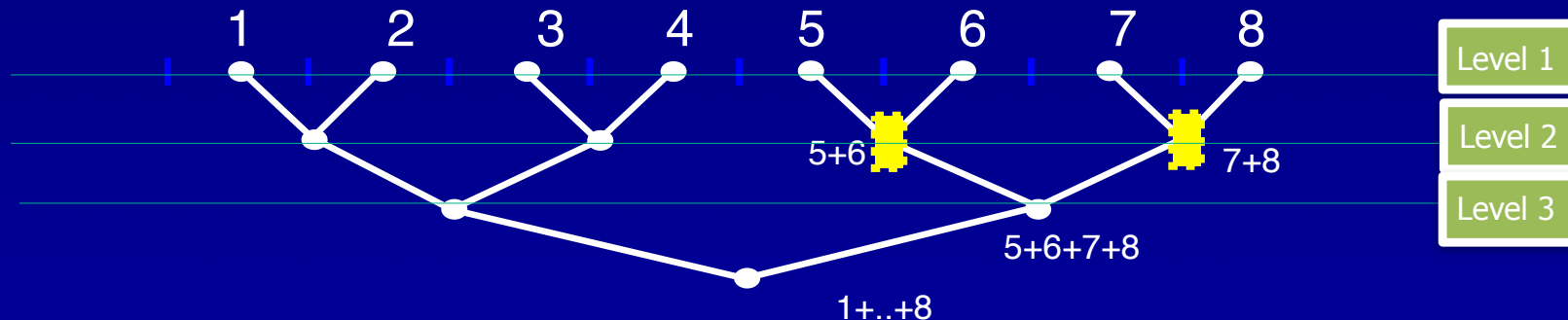


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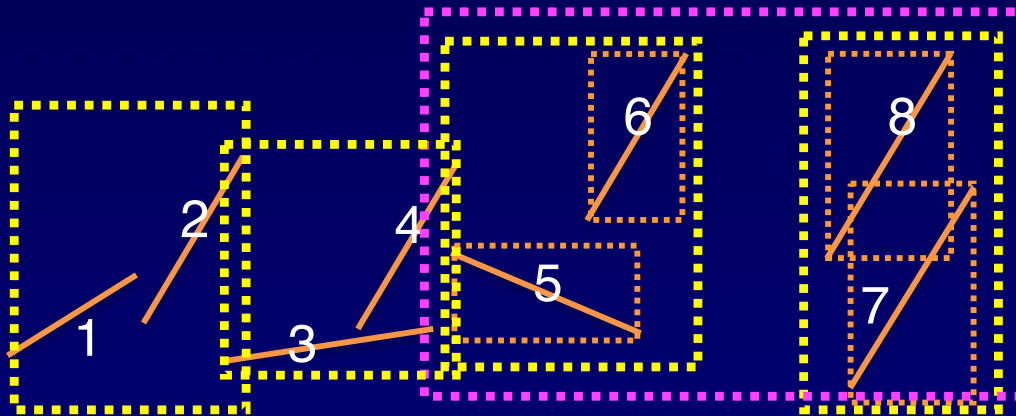


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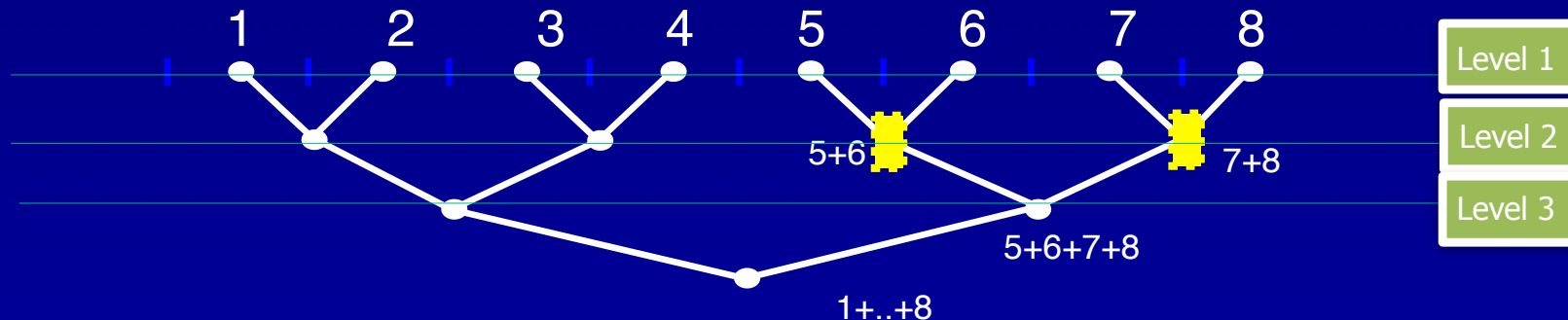


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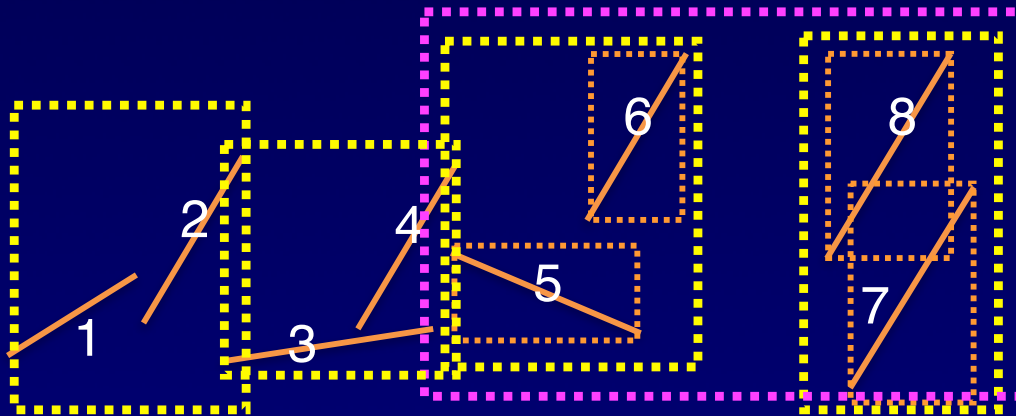


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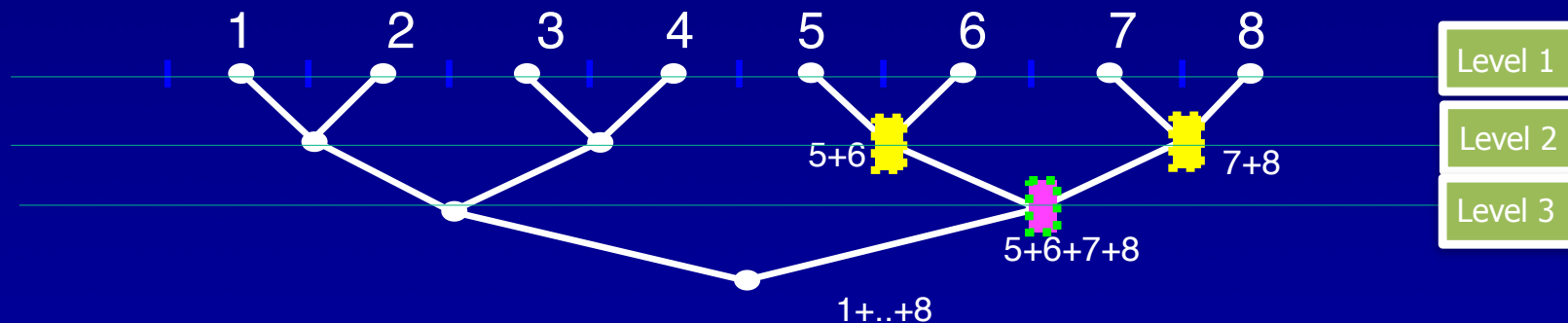


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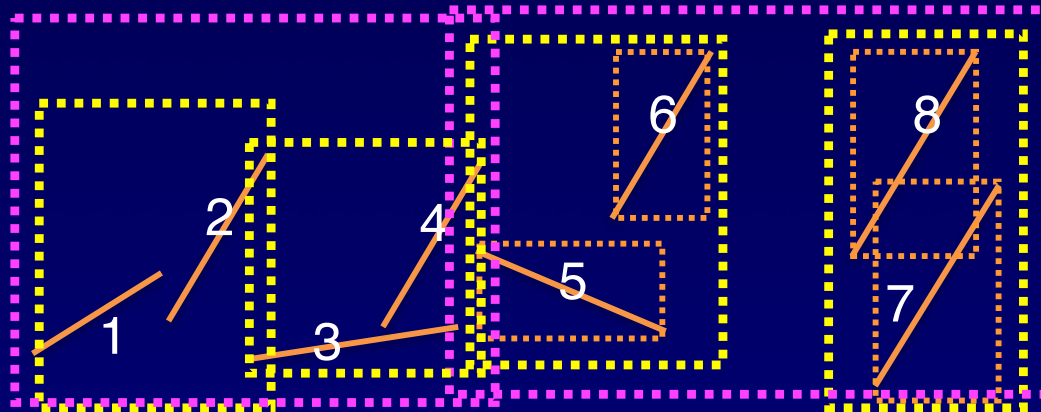


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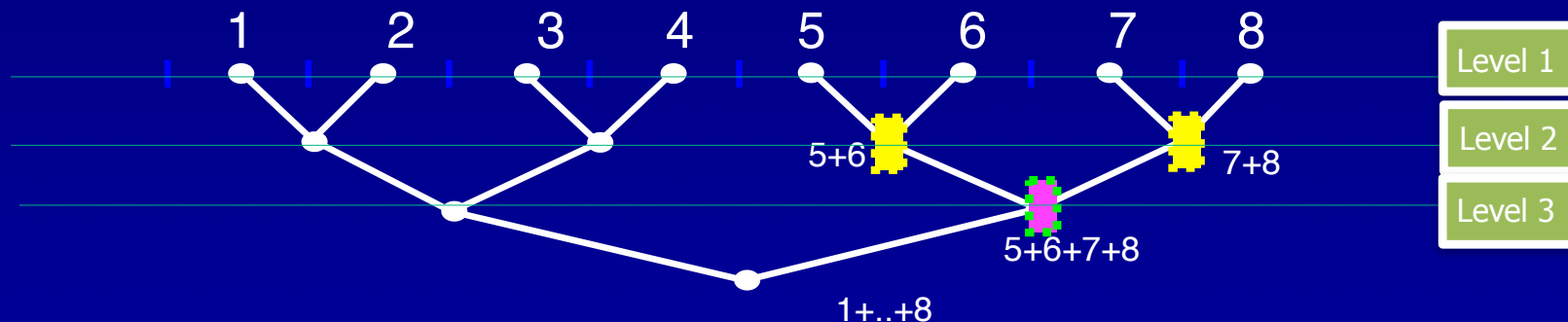


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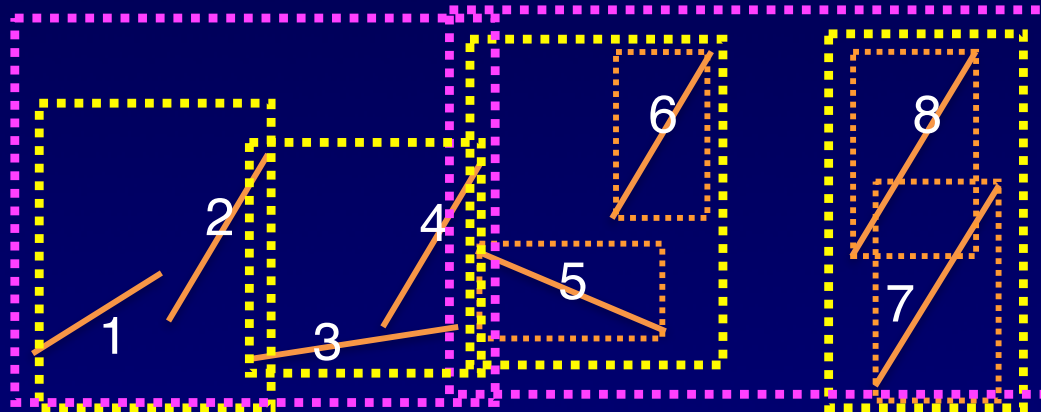


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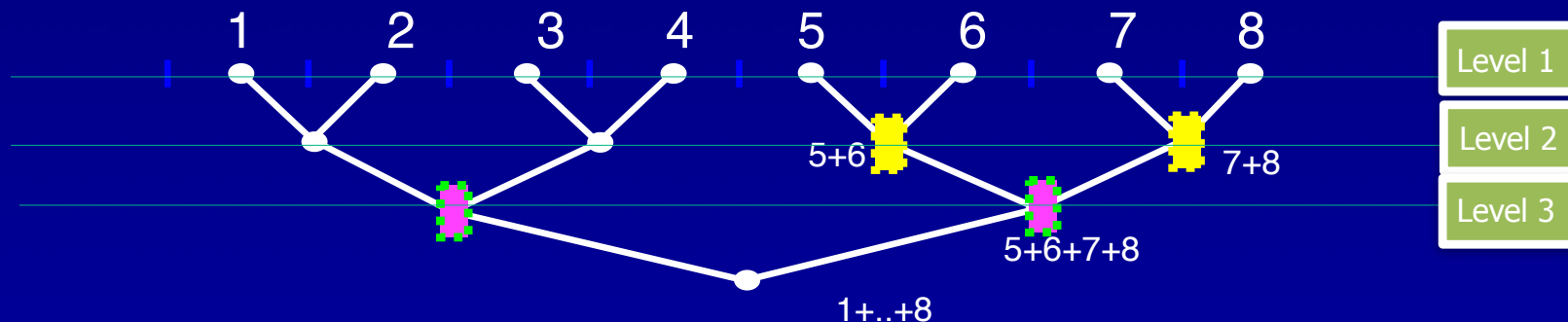


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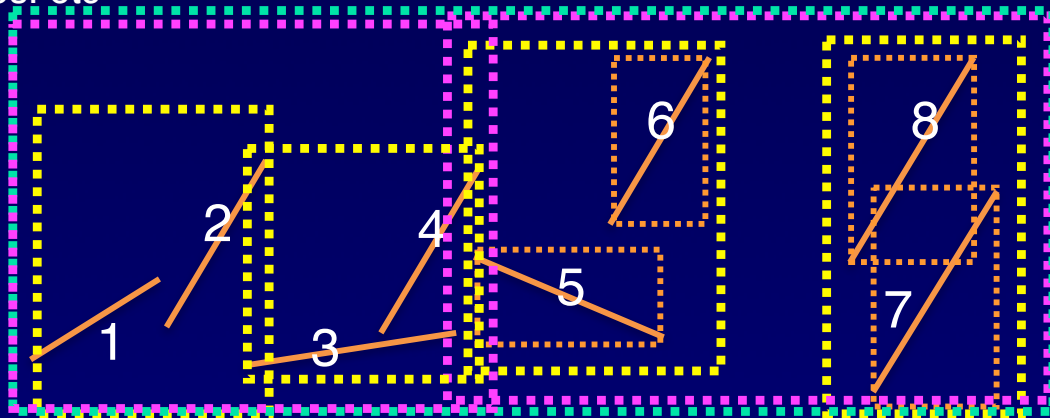


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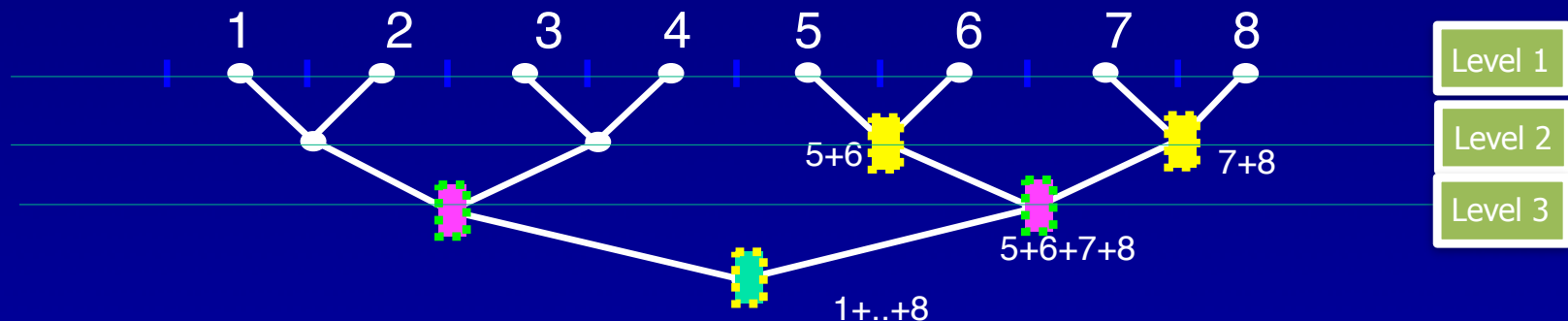


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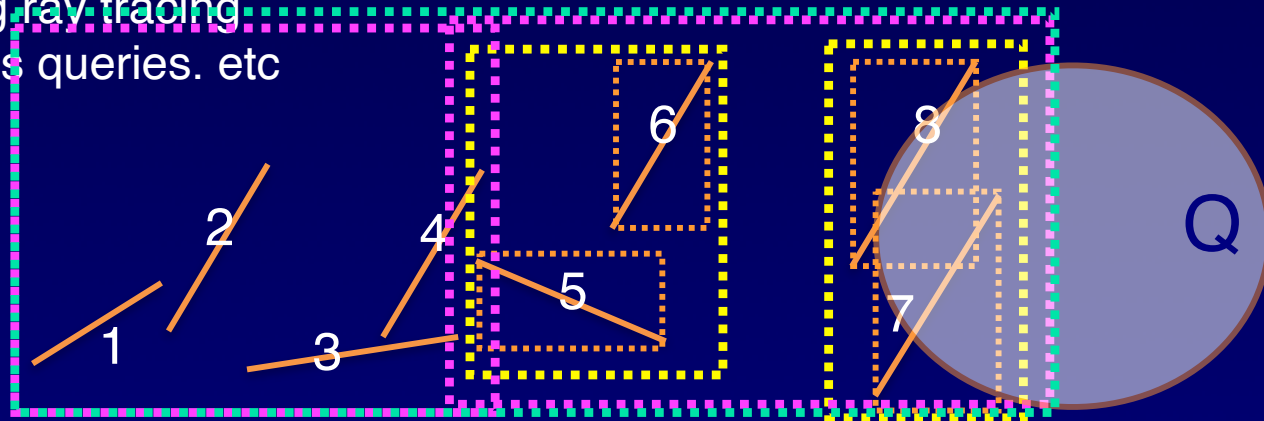


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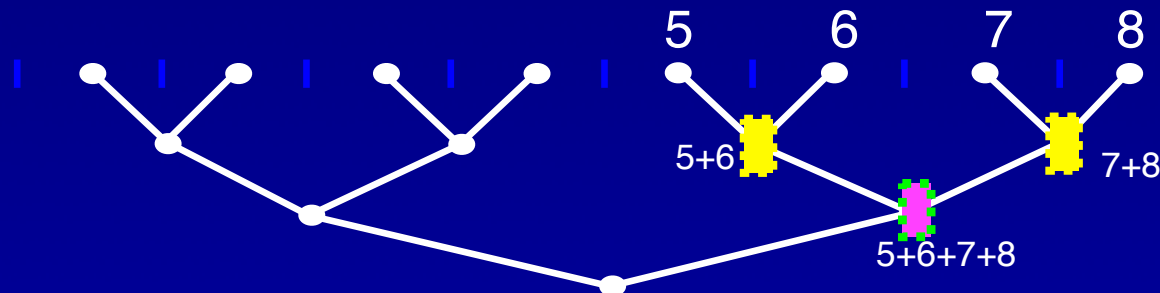
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Once a query region  $Q$  is given, we need to report the segments intersecting  $Q$ .

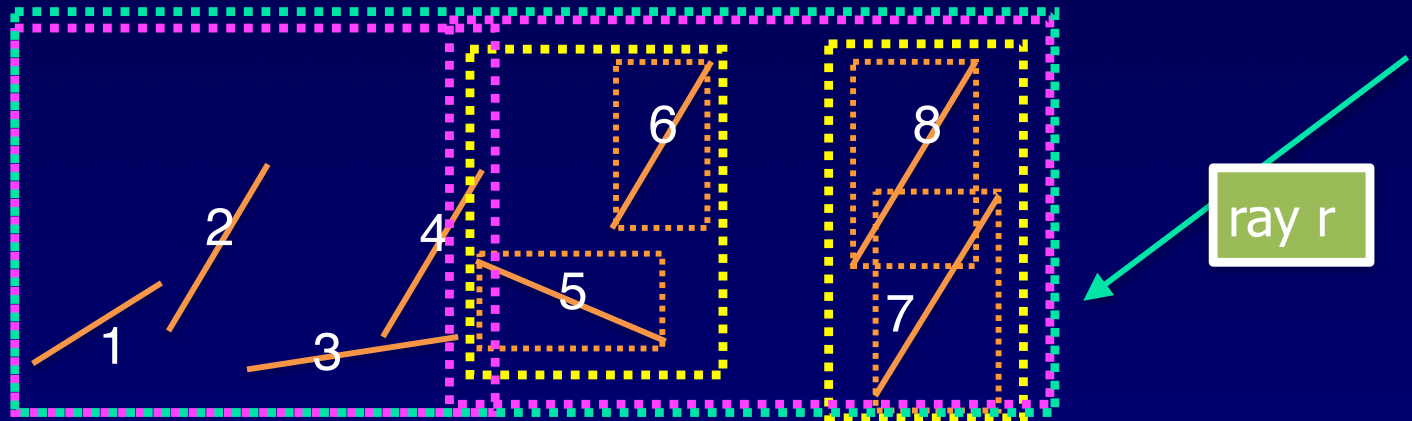
Check if  $Q$  intersects  $BB(\text{root})$

If not, we are done. If yes, check recursively if  $Q$  intersects  $BB(v.\text{left})$  and  $BB(v.\text{right})$





# R-trees



Analogously for a query ray  $r$

