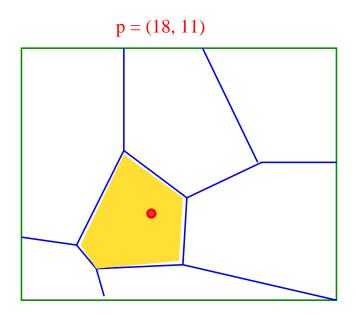
#### Point Location

• Preprocess a planar, polygonal subdivision for point location queries.

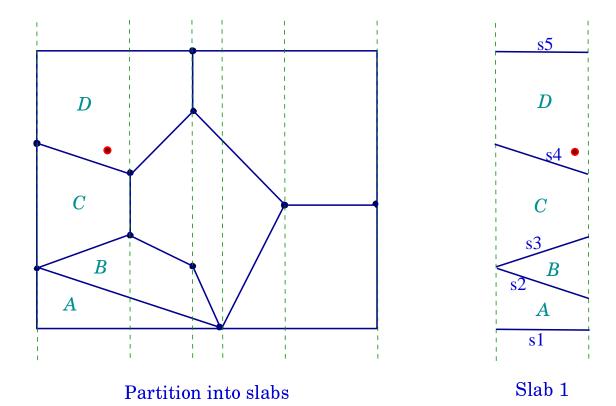


- Input is a subdivision S of complexity n, say, number of edges.
- Build a data structure on S so that for a query point p = (x, y), we can find the face containing p fast.
- Important metrics: space and query complexity.

#### The Slab Method

- Draw a vertical line through each vertex.

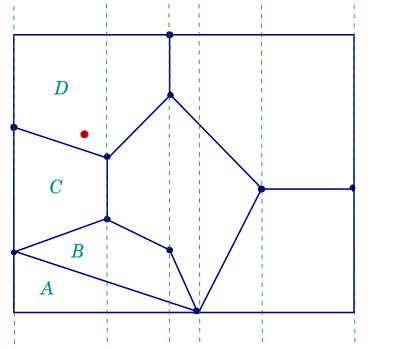
  This decomposes the plane into slabs.
- In each slab, the vertical order of line segments remains constant.



• If we know which slab p = (x, y) lies, we can perform a binary search, using the sorted order of segments.

#### The Slab Method

- To find which slab contains p, we perform a binary search on x, among slab boundaries.
- A second binary search in the slab determines the face containing p.





s5

D

 $\boldsymbol{C}$ 

 $\boldsymbol{B}$ 

 $\boldsymbol{A}$ 

s1

s4 •

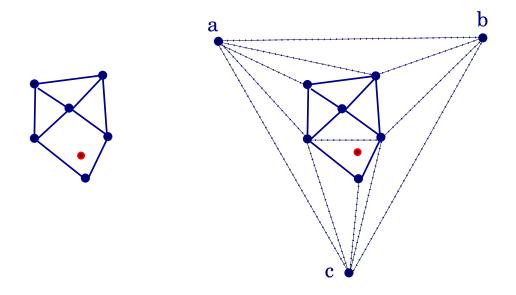
- Thus, the search complexity is  $O(\log n)$ .
- But the space complexity is  $\Theta(n^2)$ .

## **Optimal Schemes**

- There are other schemes (kd-tree, quad-trees) that can perform point location reasonably well, they lack theoretical guarantees. Most have very bad worst-case performance.
- Finding an optimal scheme was challenging. Several schemes were developed in 70's that did either  $O(\log n)$  query, but with  $O(n \log n)$  space, or  $O(\log^2 n)$  query with O(n) space.
- Today, we will discuss an elegant and simple method that achieved optimality,  $O(\log n)$  time and O(n) space [D. Kirkpatrick '83].
- Kirkpatrick's scheme however involves large constant factors, which make it less attractive in practice.
- Later we will discuss a more practical, randomized optimal scheme.

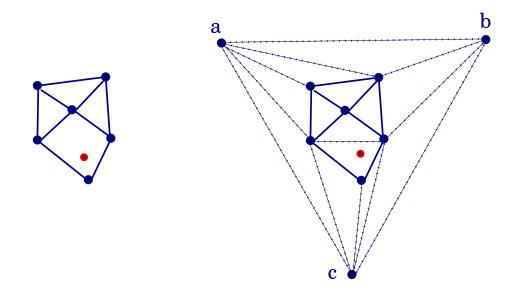
## Kirkpatrick's Algorithm

- Start with the assumption that planar subdivision is a triangulation.
- If not, triangulate each face, and label each triangular face with the same label as the original containing face.
- If the outer face is not a triangle, compute the convex hull, and triangulate the pockets between the subdivision and CH.
- Now put a large triangle *abc* around the subdivision, and triangulate the space between the two.



## Modifying Subdivision

- By Euler'e formula, the final size of this triangulated subdivision is still O(n).
- This transformation from S to triangulation can be performed in  $O(n \log n)$  time.



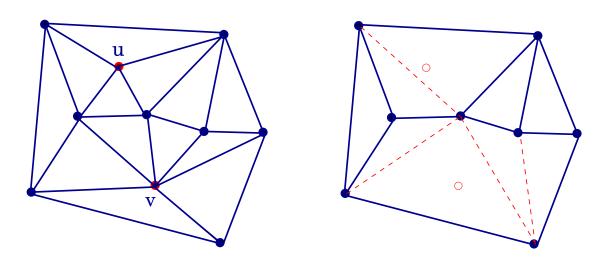
• If we can find the triangle containing p, we will know the original subdivision face containing p.

#### Hierarchical Method

- Kirkpatrick's method is hierarchical: produce a sequence of increasingly coarser triangulations, so that the last one has O(1) size.
- Sequence of triangulations  $T_0, T_1, \ldots, T_k$ , with following properties:
  - 1.  $T_0$  is the initial triangulation, and  $T_k$  is just the outer triangle abc.
  - **2.** k **is**  $O(\log n)$ .
  - 3. Each triangle in  $T_{i+1}$  overlaps O(1) triangles of  $T_i$ .
- Let us first discuss how to construct this sequence of triangulations.

## Building the Sequence

- Main idea is to delete some vertices of  $T_i$ .
- Their deletion creates holes, which we re-triangulate.



Vertex deletion and re-triangulation

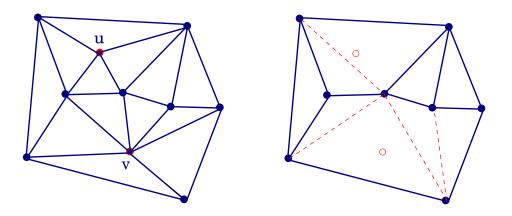
- We want to go from O(n) size subdivision  $T_0$  to O(1) size subdivision  $T_k$  in  $O(\log n)$  steps.
- Thus, we need to delete a constant fraction of vertices from  $T_i$ .
- A critical condition is to ensure each new triangle in  $T_{i+1}$  overlaps with O(1) triangles of  $T_i$ .

## Independent Sets

- Suppose we want to go from  $T_i$  to  $T_{i+1}$ , by deleting some points.
- Kirkpatrick's choice of points to be deleted had the following two properties:

[Constant Degree] Each deletion candidate has O(1) degree in graph  $T_i$ .

- If p has degree d, then deleting p leaves a hole that can be filled with d-2 triangles.
- When we re-triangulate the hole, each new triangle can overlap at most d original triangles in  $T_i$ .

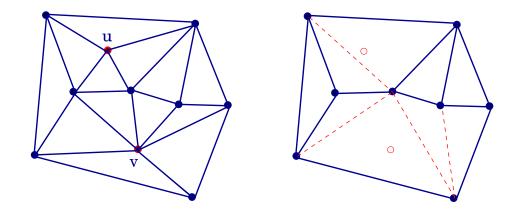


Vertex deletion and re-triangulation

## Independent Sets

[Independent Sets] No two deletion candidates are adjacent.

• This makes re-triangulation easier; each hole handled independently.



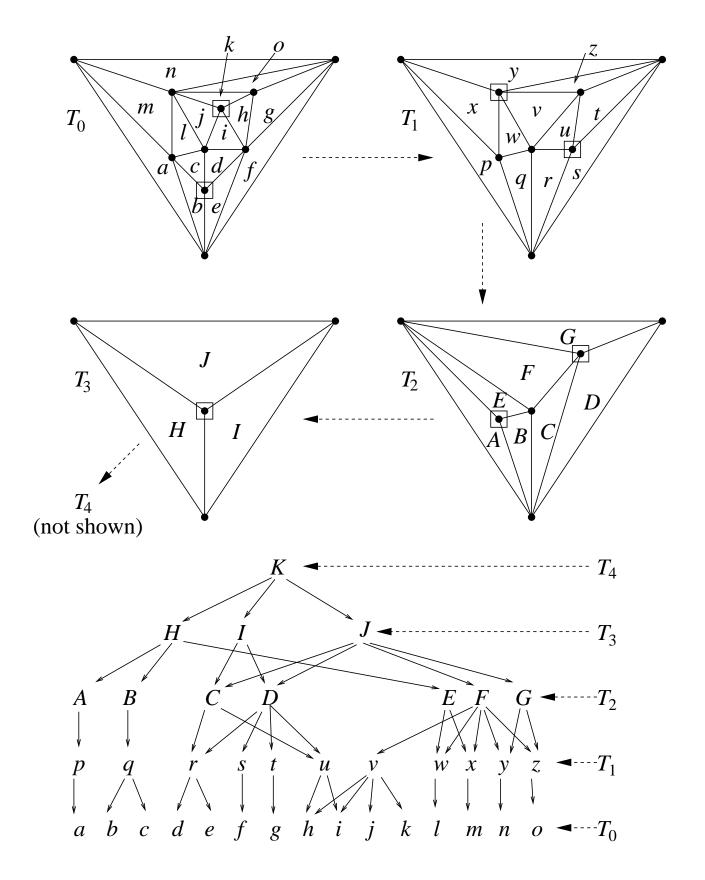
Vertex deletion and re-triangulation

#### I.S. Lemma

Lemma: Every planar graph on n vertices contains an independent vertex set of size n/18 in which each vertex has degree at most 8. The set can be found in O(n) time.

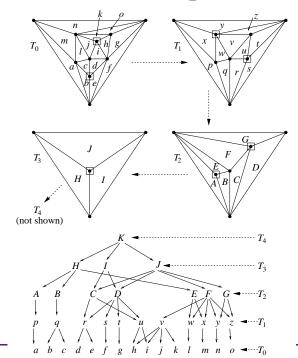
- We prove this later. Let's use this now to build the triangle hierarchy, and show how to perform point location.
- Start with  $T_0$ . Select an ind set  $S_0$  of size n/18, with max degree 8. Never pick a, b, c, the outer triangle's vertices.
- Remove the vertices of  $S_0$ , and re-triangulate the holes.
- Label the new triangulation  $T_1$ . It has at most  $\frac{17}{18}n$  vertices. Recursively build the hierarchy, until  $T_k$  is reduced to abc.
- The number of vertices drops by 17/18 each time, so the depth of hierarchy is  $k = \log_{18/17} n \approx 12 \log n$

## Illustration

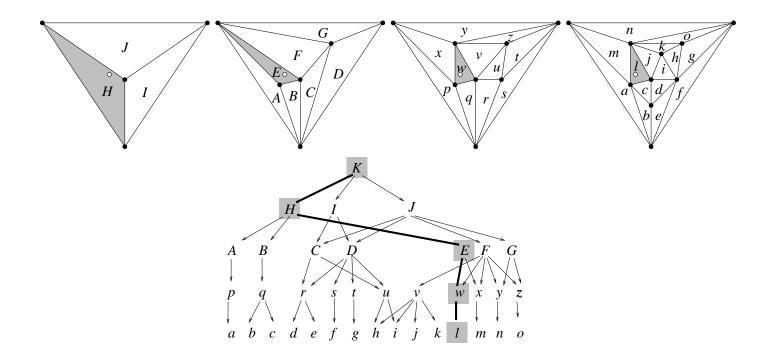


#### The Data Structure

- Modeled as a DAG: the root corresponds to single triangle  $T_k$ .
- The nodes at next level are triangles of  $T_{k-1}$ .
- Each node for a triangle in  $T_{i+1}$  has pointers to all triangles of  $T_i$  that it overlaps.
- To locate a point p, start at the root. If p outside  $T_k$ , we are done (exterior face). Otherwise, set  $t = T_k$ , as the triangle at current level containing p.

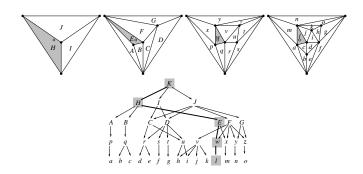


#### The Search



- Check each triangle of  $T_{k-1}$  that overlaps with t—at most 6 such triangles. Update t, and descend the structure until we reach  $T_0$ .
- Output t.

## **Analysis**



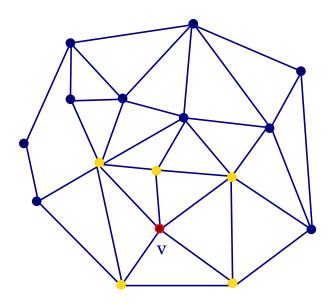
- Search time is  $O(\log n)$ —there are  $O(\log n)$  levels, and it takes O(1) time to move from level i to level i-1.
- Space complexity requires summing up the sizes of all the triangulations.
- Since each triangulation is a planar graph, it is sufficient to count the number of vertices.
- The total number of vertices in all triangulations is

$$n\left(1+(17/18)+(17/18)^2+(17/18)^3+\cdots\right)\leq 18n.$$

• Kirkpatrick structure has O(n) space and  $O(\log n)$  query time.

## Finding I.S.

- We describe an algorithm for finding the independent set with desired properties.
- Mark all nodes of degree  $\geq 9$ .
- While there is an unmarked node, do
  - 1. Choose an unmarked node v.
  - 2. Add v to IS.
  - 3. Mark v and all its neighbors.
- Algorithm can be implemented in O(n) time—keep unmarked vertices in list, and representing T so that neighbors can be found in O(1) time.

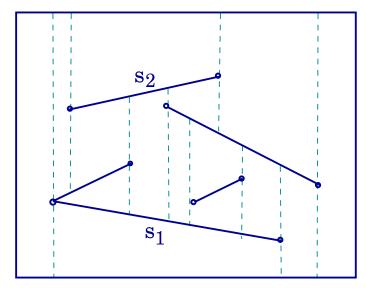


## I.S. Analysis

- Existence of large size, low degree IS follows from Euler's formula for planar graphs.
- A triangulated planar graph on n vertices has e = 3n 6 edges.
- Summing over the vertex degrees, we get  $\sum_{v} deg(v) = 2e = 6n 12 < 6n.$
- We now claim that at least n/2 vertices have degree  $\leq 8$ .
- Suppose otherwise. Then n/2 vertices all have degree  $\geq 9$ . The remaining have degree at least 3. (Why?)
- Thus, the sum of degrees will be at least  $9\frac{n}{2} + 3\frac{n}{2} = 6n$ , which contradicts the degree bound above.
- So, in the beginning, at least n/2 nodes are unmarked. Each chosen v marks at most 8 other nodes (total 9 counting itself.)
- Thus, the node selection step can be repeated at least n/18 times.
- So, there is a I.S. of size  $\geq n/18$ , where each node has degree  $\leq 8$ .

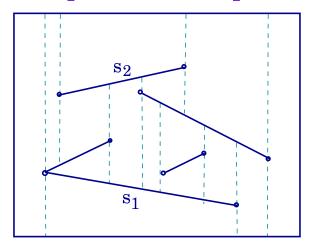
## Trapezoidal Maps

- A randomized point location scheme, with (expected) query  $O(\log n)$ , space O(n), and construction time  $O(n \log n)$ .
- The expectation does not depend on the polygonal subdivision. The bounds holds for any subdivision.
- It appears simpler to implement, and its constant factors are better than Kirkpatrick's.
- The algorithm is based on trapezoidal maps, or decompositions, also encountered earlier in triangulation.



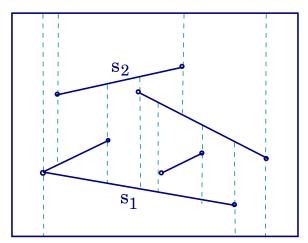
## Trapezoidal Maps

- Input a set of non-intersecting line segments  $S = \{s_1, s_2, \dots, s_n\}$ .
- Query: given point p, report the segment directly above p.
- The region label can be easily encoded into the line segments.
- Map is created by shooting a ray vertically from each vertex, up and down, until a segment is hit.
- In order to avoid degeneracies, assume that no segment is vertical.
- The resulting rays plus the segments define the trapezoidal map.



## Trapezoidal Maps

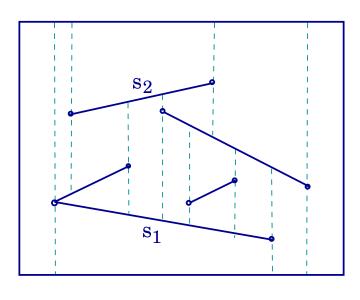
- Enclose S into a bounding box to avoid infinite rays.
- All faces of the subdivision are trapezoids, with vertical sides.
- Size Claim: If S has n segments, the map has at most 6n+4 vertices and 3n+1 traps.



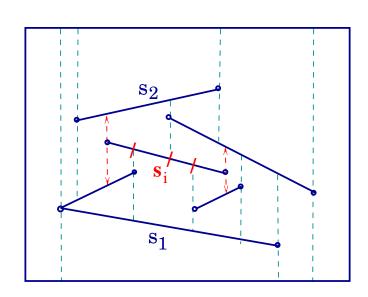
- Each vertex shoots one ray, each resulting in two new vertices, so at most 6n vertices, plus 4 for the outer box.
- The left boundary of each trapezoid is defined by a segment endpoint, or lower left corner of enclosing box.
- The corner of box acts as leftpoint for one trap; the right endpoint of any segment also for one trap; and left endpoint of any segment for at most 2 trapezoids. So total of 3n + 1.

#### Construction

- Plane sweep possible, but not helpful for point location.
- Instead we use randomized incremental construction.
- Historically, invented for randomized segment intersection. Point location an intermediate problem.
- Start with outer box, one trapezoid. Then, add one segment at a time, in an arbitrary, not sorted, order.



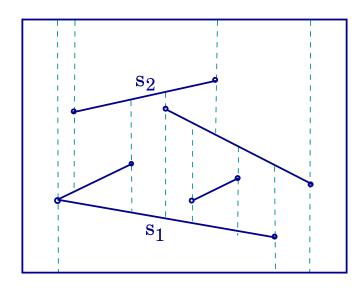


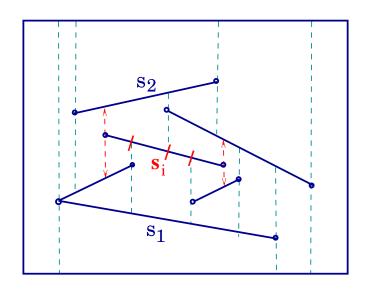


After inserting s

#### Construction

- Let  $S_i = \{s_1, s_2, \dots, s_i\}$  be first i segments, and  $\mathcal{T}_i$  be their trapezoidal map.
- Suppose  $\mathcal{T}_{i-1}$  built, and we add  $s_i$ .
- Find the trapezoid containing the left endpoint of  $s_i$ . Defer for now: this is point location.
- Walk through  $\mathcal{T}_{i-1}$ , identifying trapezoids that are cut. Then, "fix them up".
- Fixing up means, shoot rays from left and right endpoints of  $s_i$ , and trim the earlier rays that are cut by  $s_i$ .





Before

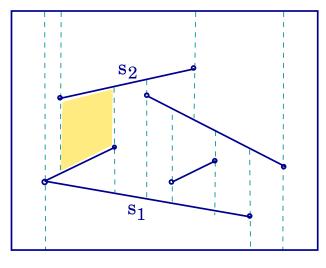
After inserting s

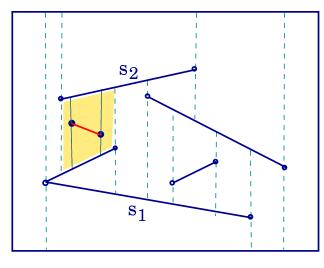
## **Analysis**

- Observation: Final structure of trap map does not depend on the order of segments. (Why?)
- Claim: Ignoring point location, segment i's insertion takes  $O(k_i)$  time if  $k_i$  new trapezoids created.

#### • Proof:

- Each endpoint of  $s_i$  shoots two rays.
- Additionally, suppose  $s_i$  interrupts K existing ray shots, so total of K+4 rays need processing.
- If K = 0, we get exactly 4 new trapezoids.
- For each interrupted ray shot, a new trapezoid created.
- With DCEL, update takes O(1) per ray.

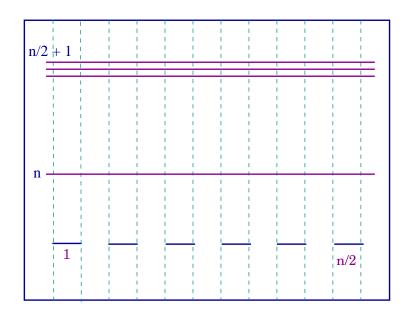




Before After

#### Worst Case

- In a worst-case,  $k_i$  can be  $\Theta(i)$ . This can happen for all i, making the worst-case run time  $\sum_{i=1}^{n} i = \Theta(n^2)$ .
- Using randomization, we prove that if segments are inserted in random order, then expected value of  $k_i$  is O(1)!
- So, for each segment  $s_i$ , the expected number of new trapezoids created is a constant.
- Figure below shows a worst-case example. How will randomization help?

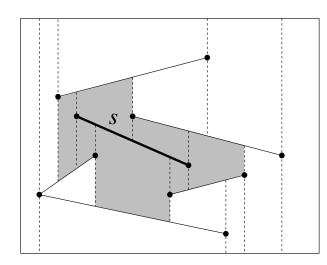


### Randomization

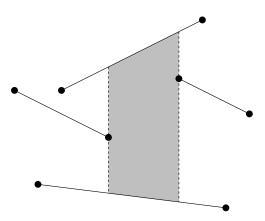
• Theorem: Assume  $s_1, s_2, \ldots, s_n$  is a random permutation. Then,  $E[k_i] = O(1)$ , where  $k_i$  trapezoids created upon  $s_i$ 's insertion, and the expectation is over all permutations.

#### • Proof.

- 1. Consider  $\mathcal{T}_i$ , the map after  $s_i$ 's insertion.
- 2.  $\mathcal{T}_i$  does not depend on the order in which segments  $s_1, \ldots, s_i$  were added.
- 3. Reshuffle  $s_1, \ldots, s_i$ . What's the probability that a particular s was the last segment added?
- 4. The probability is 1/i.
- 5. We want to compute the number of trapezoids that would have been created if s were the last segment.



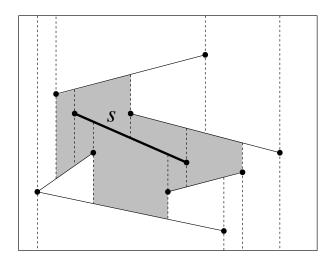
The trapezoids that depend on s

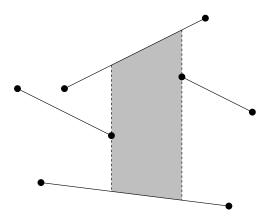


The segments that the trapezoid depends on.

#### **Proof**

- Say trapezoid  $\Delta$  depends on s if  $\Delta$  would be created by s if s were added last.
- Want to count trapezoids that depend on each segment, and then find the average over all segments.
- Define  $\delta(\Delta, s) = 1$  if  $\Delta$  depends on s; otherwise,  $\delta(\Delta, s) = 0$ .





The trapezoids that depend on s

The segments that the trapezoid depends on.

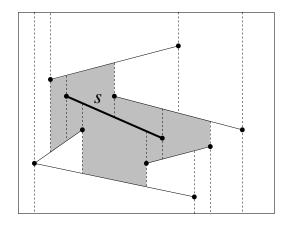
• The expected complexity is

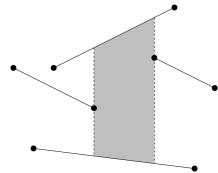
$$E[k_i] = \frac{1}{i} \sum_{s \in S_i} \sum_{\Delta \in \mathcal{T}_i} \delta(\Delta, s)$$

- Some segments create a lot of trapezoids; others very few.
- Switch the order of summation:

$$E[k_i] \; = \; \frac{1}{i} \sum_{\Delta \in \mathcal{T}_i} \sum_{s \in S_i} \delta(\Delta, s)$$

### **Proof**





The trapezoids that depend on s

The segments that the trapezoid depends on.

• Now we are counting number of segments each trapezoid depents on.

$$E[k_i] \ = \ \frac{1}{i} \sum_{\Delta \in \mathcal{T}_i} \sum_{s \in S_i} \delta(\Delta, s)$$

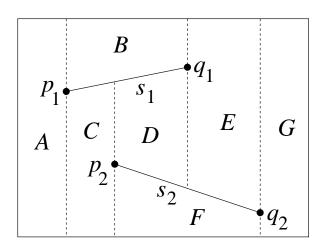
- This is much easier—each  $\Delta$  depends on at most 4 segments.
- Top and bottom of  $\Delta$  defined by two segments; if either of them added last, then  $\Delta$  comes into existence.
- Left and right sides defined by two segments endpoints, and if either one added last,  $\Delta$  is created.
- Thus,  $\sum_{s \in S_i} \delta(\Delta, s) \leq 4$ .
- $\mathcal{T}_i$  has O(i) trapezoids, so

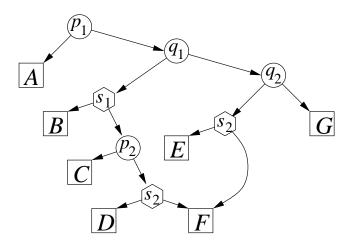
$$E[k_i] = \frac{1}{i} \sum_{\Delta \in \mathcal{T}_i} 4 = \frac{1}{i} 4 |\mathcal{T}_i| = \frac{1}{i} O(i) = O(1).$$

• End of proof.

#### Point Location

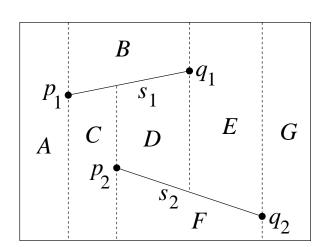
- Like Kirkpatrick's, point location structure is a rooted directed acyclic graph.
- To query processor, it looks like a binary tree, but subtree may be shared.
- Tree has two types of nodes:
  - x-node: contains the x-coordinate of a segment endpoint. (Circle)
  - y-node: pointer to a segment. (Hexagon)
- A leaf for each trapzedoid.

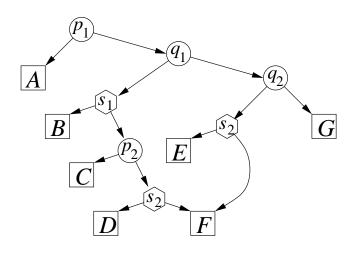




#### Point Location

- Children of x-node correspond to points lying to the left and right of x coord.
- Children of *y*-node correspond to space below and above the segment.
- y-node searched only when query's x-coordinate is within segment's span.
- $\bullet$  Example: query in region D.

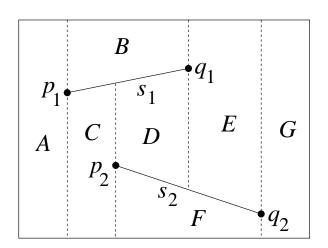


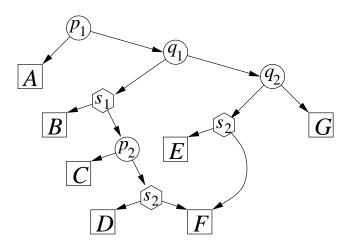


• Encodes the trap decomposition, and enables point location during the construction as well.

## Building the Structure

- Incremental construction, mirroring the trapezoidal map.
- When a segment s added, modify the tree to account for changes in trapezoids.
- Essentially, some leaves will be replaced by new subtrees.
- Like Kirkpatrick's, each old trapezoid will overlap O(1) new trapezoids.

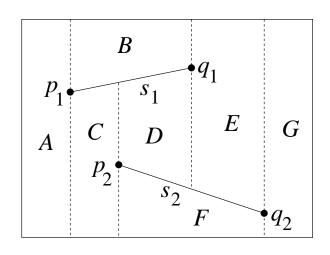


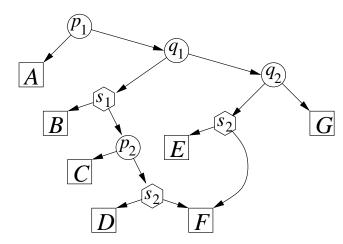


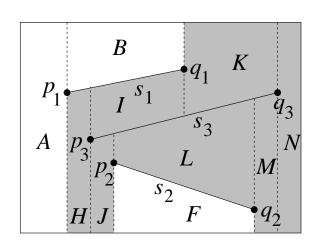
• Each trapezoid appears exactly once as a leaf. For instance, *F*.

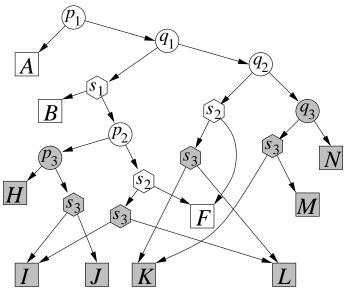
# Adding a Segment

• Consider adding segment  $s_3$ .



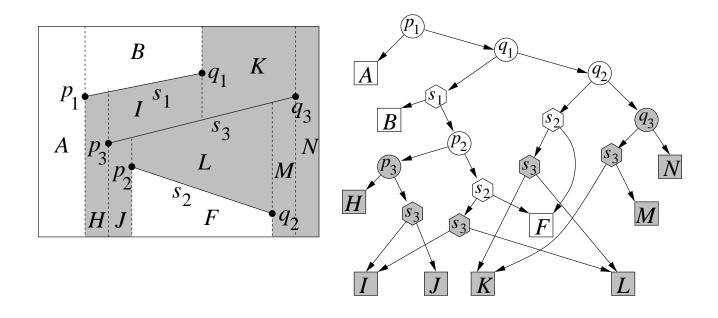






## Adding a Segment

- Changes are highly local.
- If segment s passes entirely through an old trapezoid t, then t is replaced by two traps t', t''.
  - During search, we need to compare query point to s to decide above/below.
  - So, a new y-node added which is the parent of t' and t''.
- If an endpoint of s lies in t, then we add a x-node to decide left/right and a y-node for the segment.

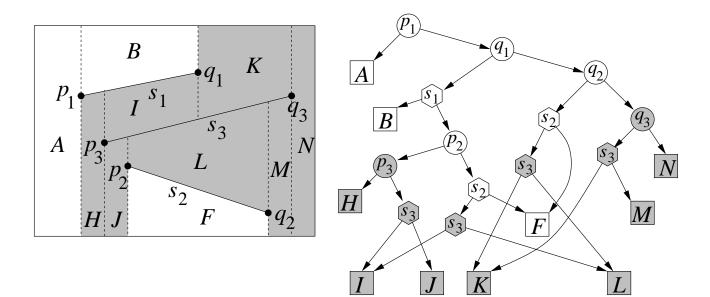


## **Analysis**

- Space is O(n), and query time is  $O(\log n)$ , both in expectation.
- Expected bound depends on the random permutation, and not on the choice of input segments or the query point.
- The data structure size  $\propto$  number of trapezpoids, which is O(n), since O(1) expected number of traps created when a new segment inserted.
- In order to analyze query bound, fix a query q.
- We consider how q moves incrementally through the trapezoidal map as new segments are inserted.
- Search complexity  $\propto$  number of trapezoids encountered by q.

## Search Analysis

- Let  $\Delta_i$  be trapezoid containing q after insertion of ith segment.
- If  $\Delta_i = \Delta_{i-1}$  then new insertion does not affect q's trapezoid. (E.g.  $q \in B$  and  $s_3$ 's insertion.)
- If  $\Delta_i \neq \Delta_{i-1}$ , then new segment deleted q's trapezoid, and q needs to locate itself among the (at most 4) new traps.
- q could fall 3 levels in the tree. E.g.  $q \in C$  falling to J after  $s_3$ 's insertion.

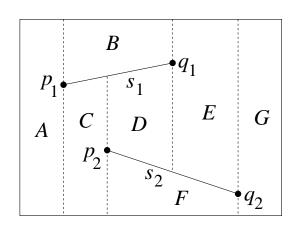


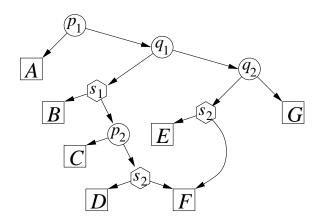
## Search Analysis

- Let  $P_i$  be probability that  $\Delta_i \neq \Delta_{i-1}$ , over all random permutation.
- Since q can drop  $\leq 3$  levels, expected search path length is  $\sum_{i=1}^{n} 3P_i$ .
- We will show that  $P_i \leq 4/i$ . That will imply that expected search path length is

$$3\sum_{i=1}^{n} \frac{4}{i} = 12\sum_{i=1}^{n} \frac{1}{i} = 12\ln n$$

- Why is  $P_i \leq 4/i$ ? Use backward analysis.
- The trapezoid  $\Delta_i$  depends on at most 4 segments. The probability that *i*th segment is one of these 4 is at most 4/i.





#### Final Remarks

- Expectation only says that average search path is small. It can still have large variance.
- The trapezoidal map data structure has bounds on variance too. See the textbook for complete analysis.

Theorem: For any  $\lambda > 0$ , the probability that depth of the randomized seach structure exceeds  $3\lambda \ln(n+1)$  is at most

$$\frac{2}{(n+1)^{\lambda \ln 1.25 - 3}}$$

• More careful analysis can provide better constants for the data structure.