



Convex Hull for Dynamic Data

Convex Hull and Parallel Tree Contraction

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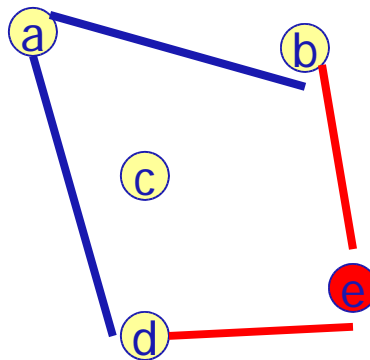
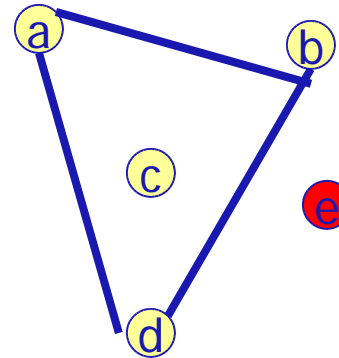
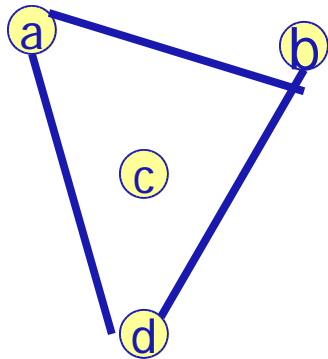
(joint work with Umut Acar, and Guy Blelloch)



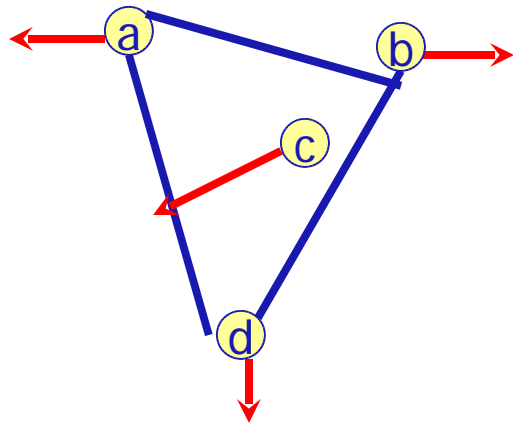
Motivation

- Application data is dynamic
 - word processors: slowly changing text
 - graphics: render similar images
 - mobile phone networks: continuously moving hosts
- Important to handle dynamic data efficiently

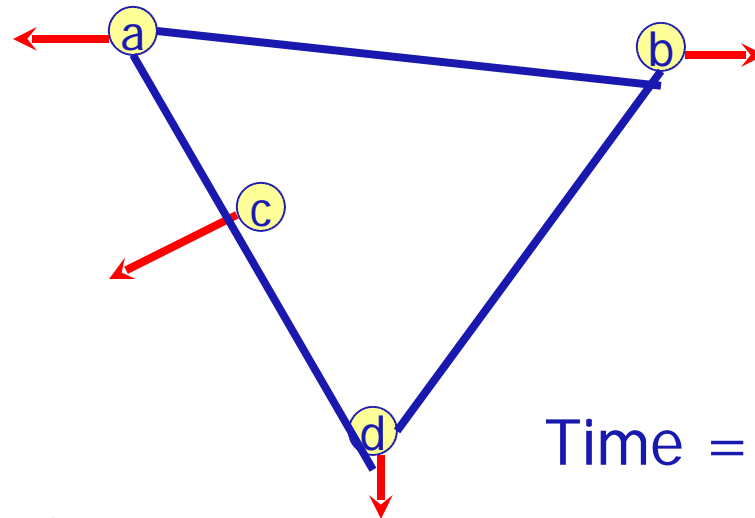
Dynamic Algorithms: Changing Data



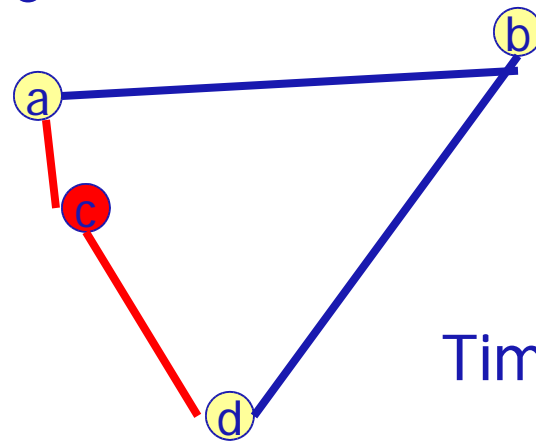
Kinetic Algorithms: Moving Data



Time = 0



Time = 1



Time = $1 + \epsilon \dots \infty$



How to invent Dynamic/Kinetic Algorithms

- Just like any other algorithm. Think, ponder, divide, conquer...
- Or, use adaptivity...



Adaptivity

- Makes a standard algorithm dynamic or kinetic
- Requires little change to the standard algorithm
- Can be done semi-automatically
- Not all algorithms yield efficient adaptive algorithms
 - Will talk about this more

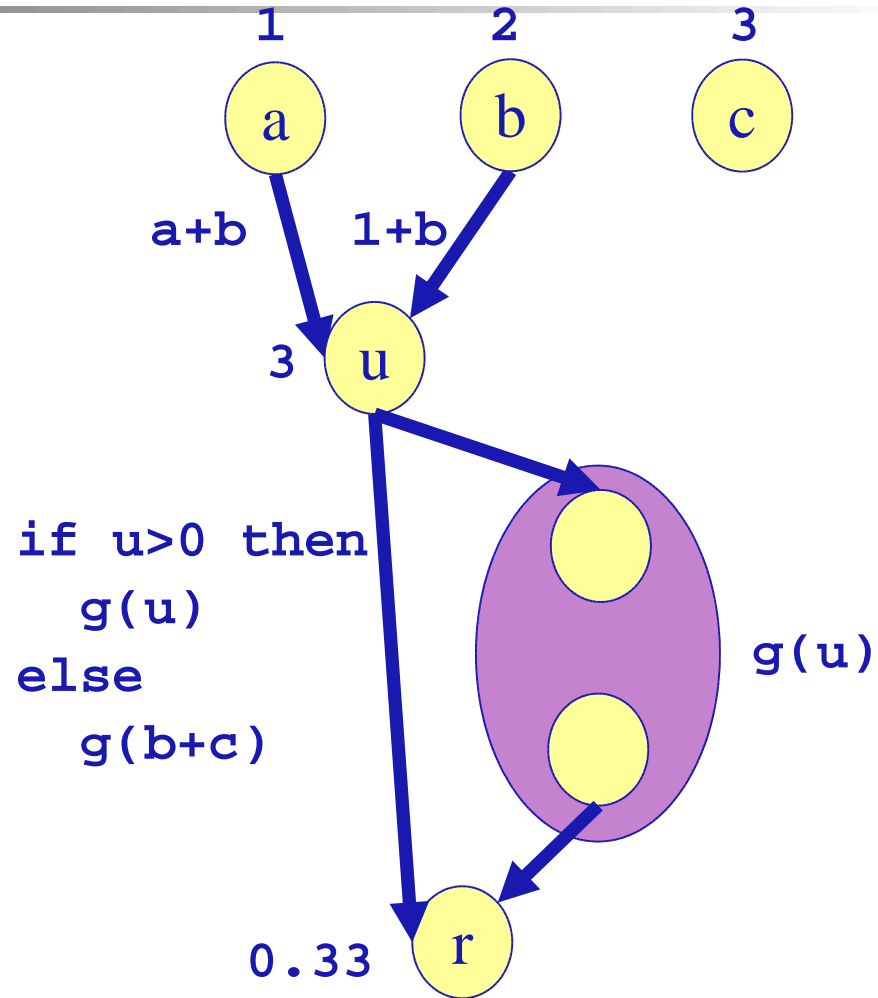


How does Adaptivity Work?

- Represent a computation with a dynamic dependency graph
- nodes \ni data, edges = dependencies
- Sources = input, sinks = output,
- The user can
 - change the input,
 - update the output
- Update:
 - Take a changed node,
 - Update all its children (the children are now changed)
 - Repeat until no more changed nodes

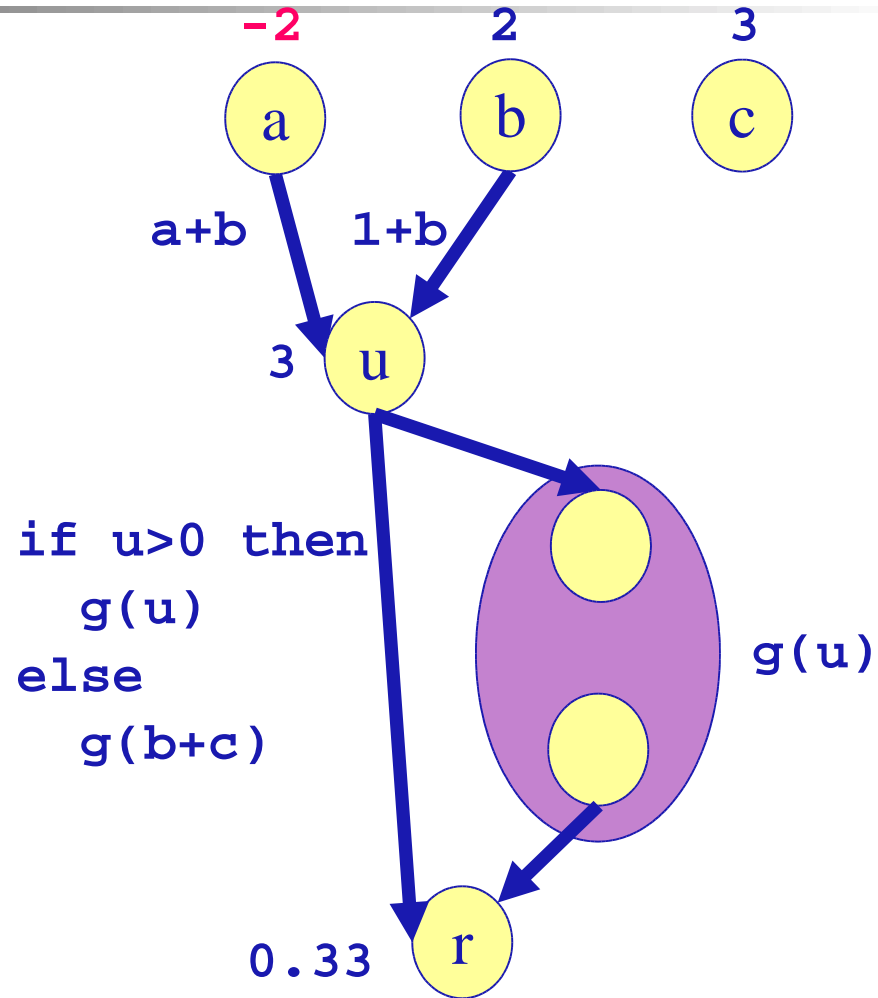
Adaptivity Example

```
fun f (a,b,c) =  
  let  
    u = a+b  
  in  
    if (u > 0) then  
      g(u)  
    else  
      g(b+c)  
  end
```



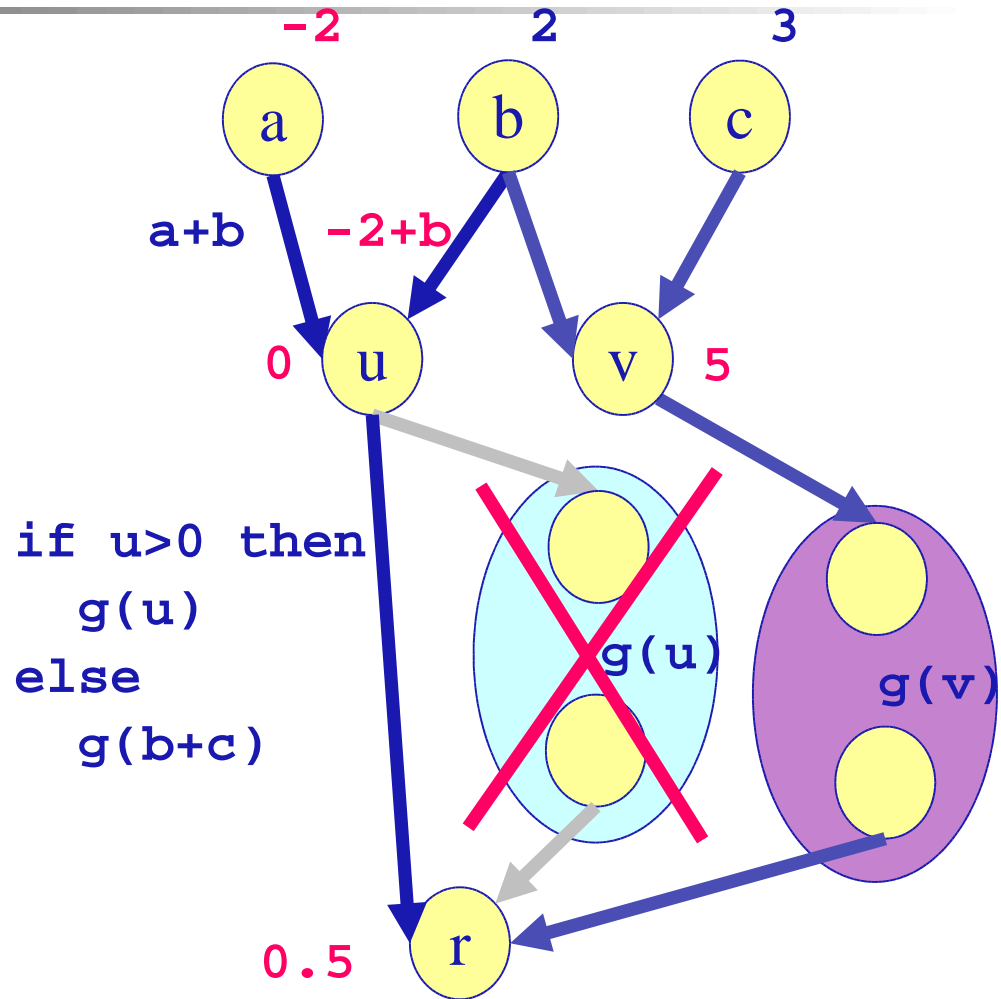
Adaptivity Example: Change

```
fun f (a,b,c) =  
  let  
    u = a+b  
  in  
    if (u > 0) then  
      g(u)  
    else  
      g(b+c)  
  end
```



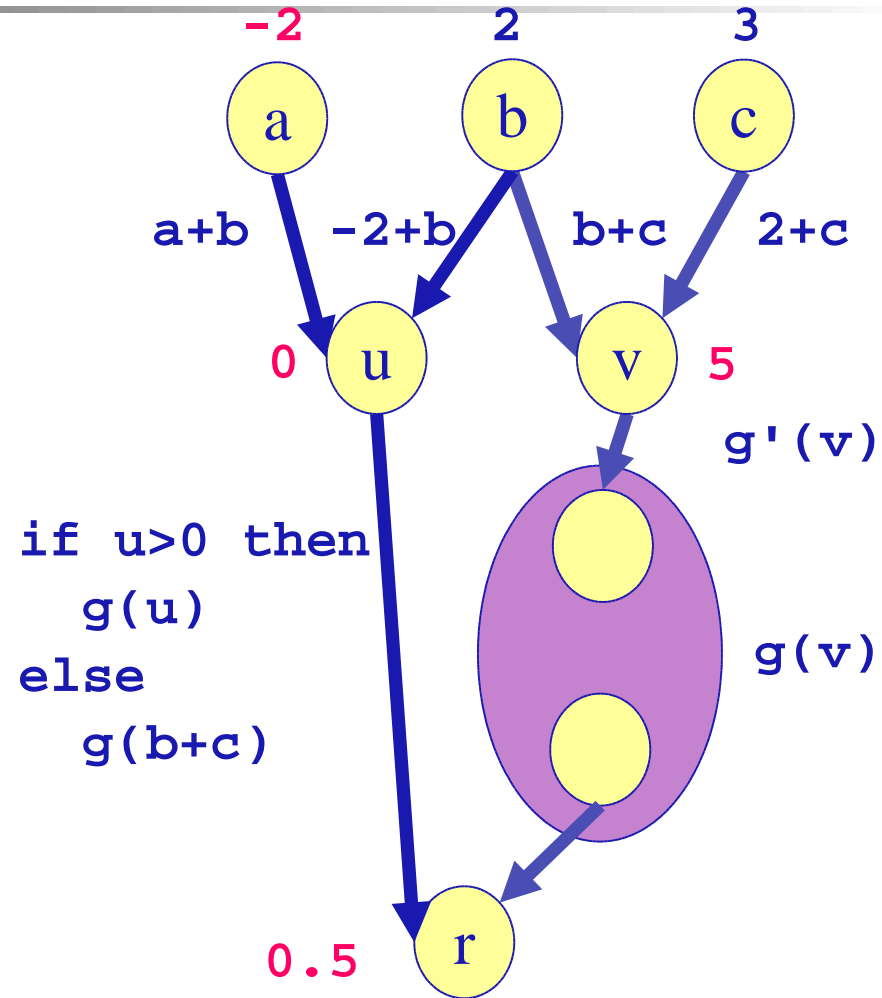
Update

```
fun f (a,b,c) =  
  let  
    u = a+b  
  in  
    if (u > 0) then  
      g(u)  
    else  
      g(b+c)  
  end
```



Update

```
fun f (a,b,c) =  
  let  
    u = a+b  
  in  
    if (u > 0) then  
      g(u)  
    else  
      g(b+c)  
  end
```





Adaptivity and Stability

- Adaptivity updates the result by rerunning the parts of the computation affected by the input change
- It is efficient when the computation is “stable”, i.e., computations on “similar” inputs are “similar”
- We apply the adaptivity technique to convex hulls
- Result: Efficient Dynamic and Kinetic convex hulls



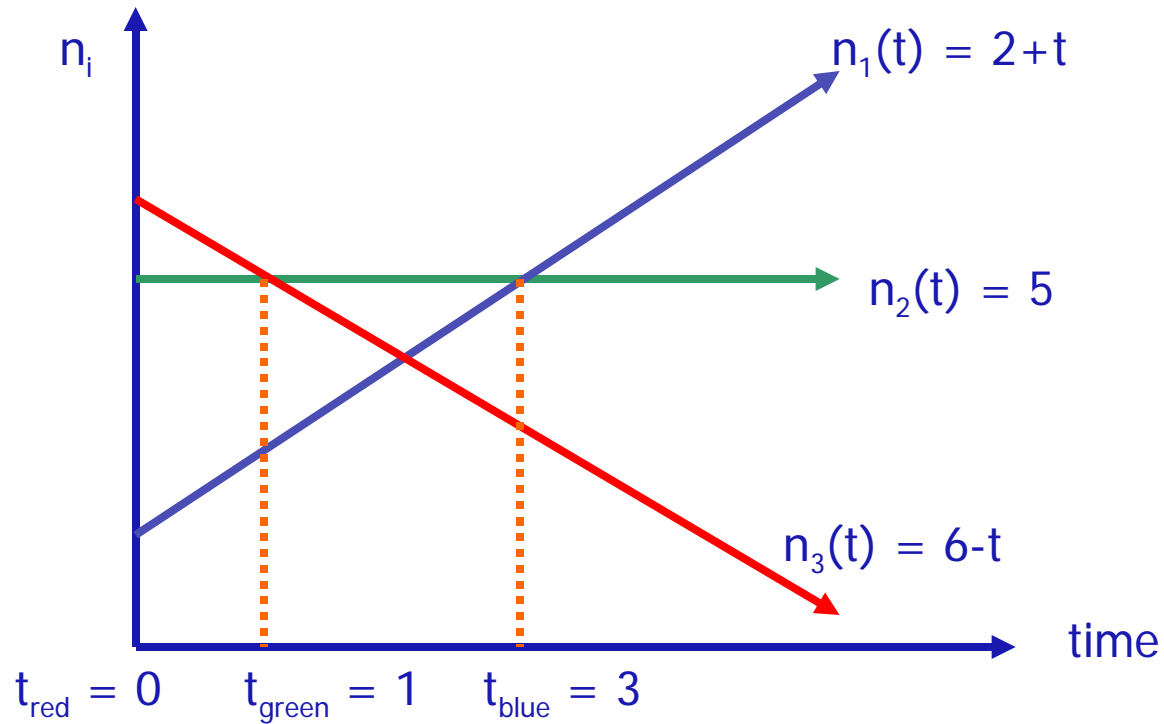
1-D Convex Hull: Max and Min

- Just consider upper hull: Finding the maximum
- Consider two algorithms:
 - The March: March through the list
 - The Tournament: pair up the elements and take the max of each pair

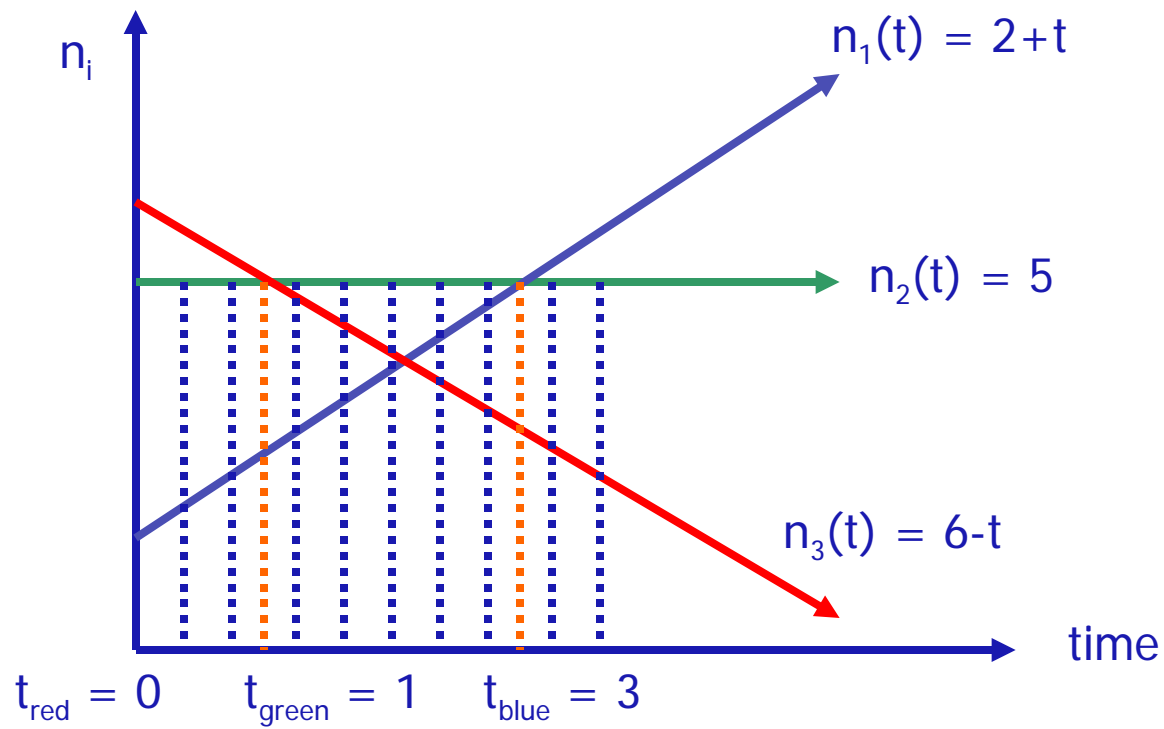


Kinetic Maximum

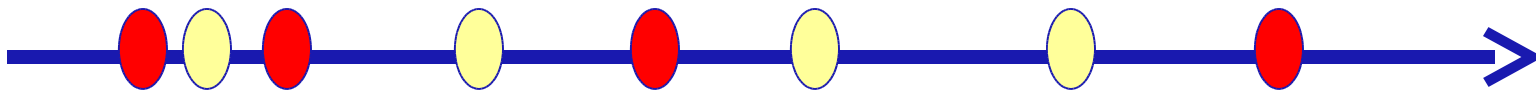
- Numbers increase/decrease continuously in time $n_i(t) = n_i + c_i t$



Sampling



Internal and External Events



 External event: Final result changes

 Internal event: Final result does not change
but a test fails



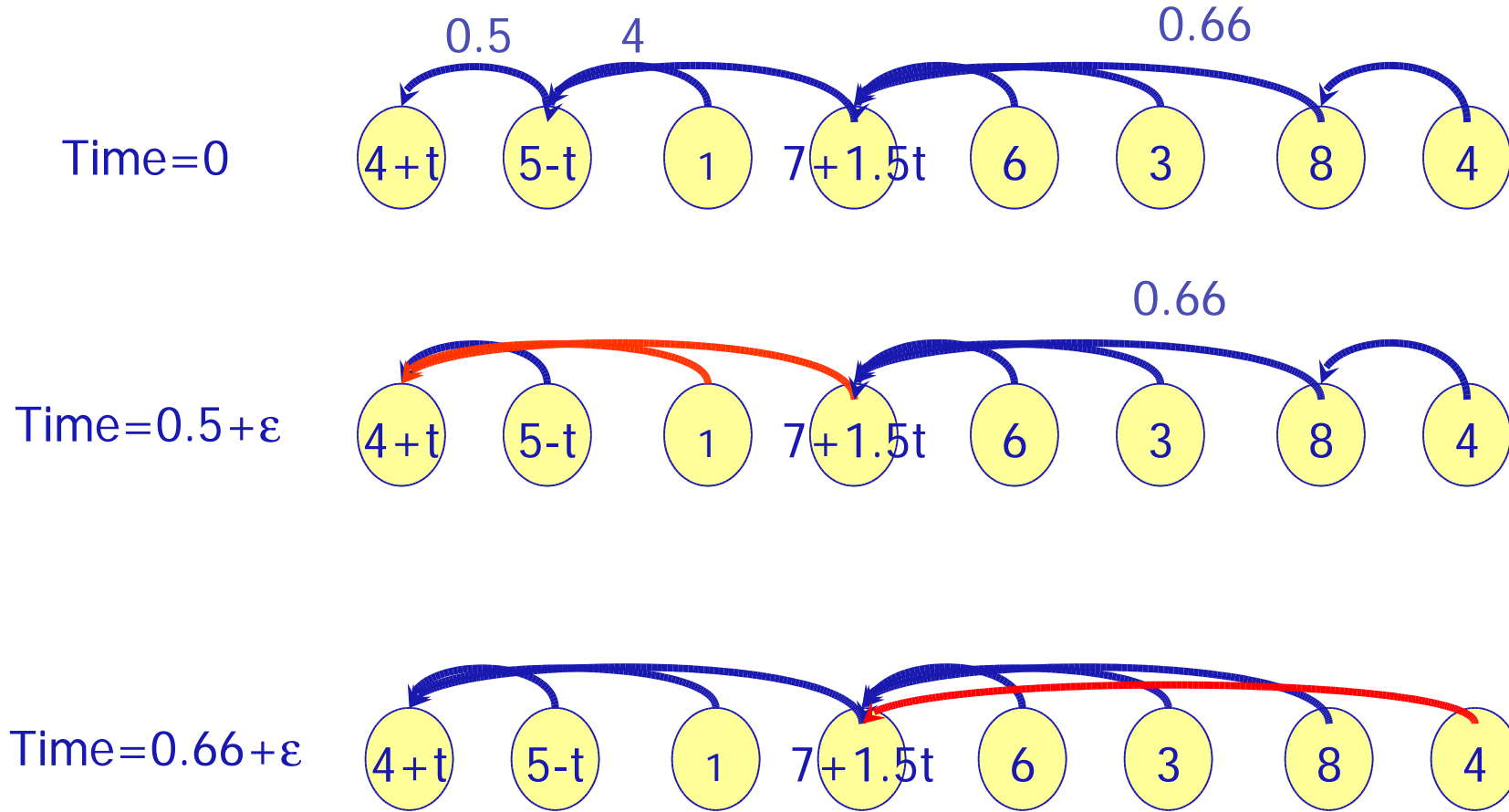
Proof Simulation via Certificates

- A set of comparisons that prove the current maximum
- Associate a **certificate** with each comparison
 - certificate = comparison result + **failure time**
- Consider the times that a certificate fails

- We need an algorithm that updates the result as well as the certificates

- Use adaptivity to obtain to do the update efficiently

Kinetic March



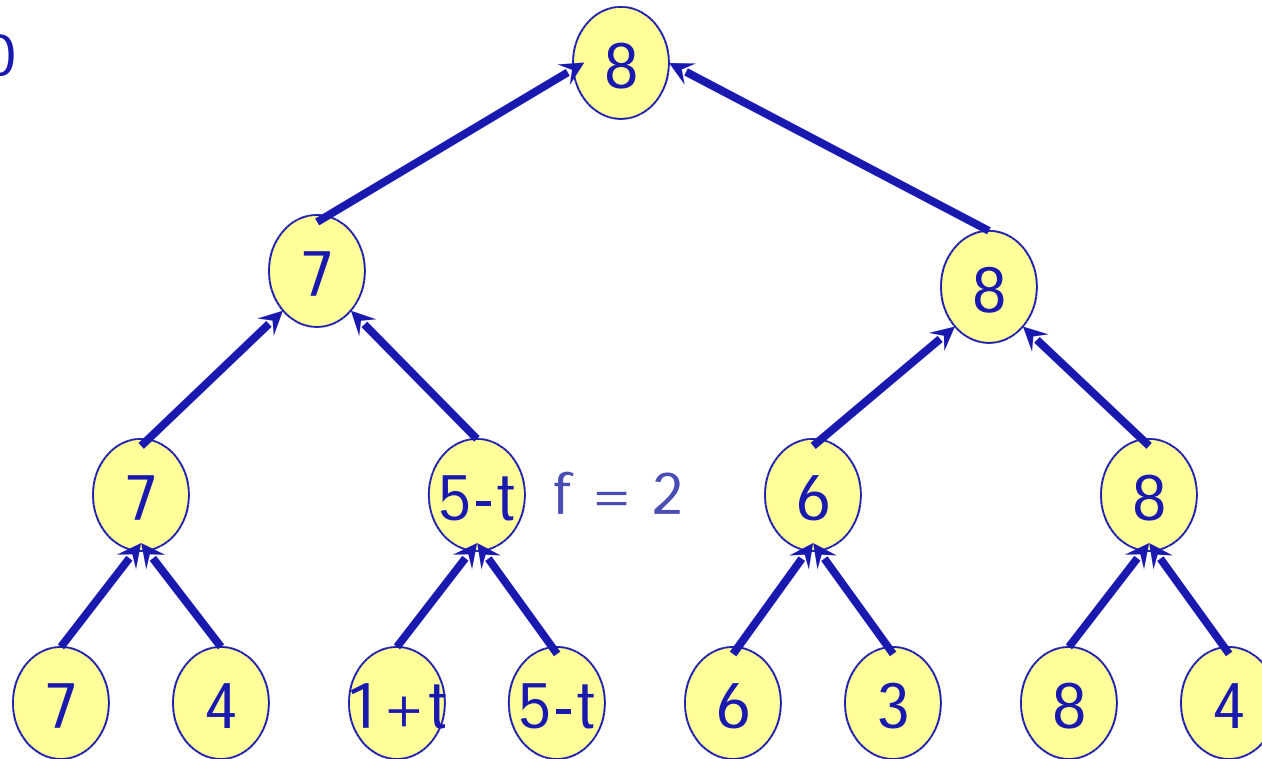


Kinetic March Performance

- $O(n)$: Because an item in the beginning of the list could become the maximum and it will be compared to the rest of the list
- Not acceptable because computing from scratch takes $O(n)$

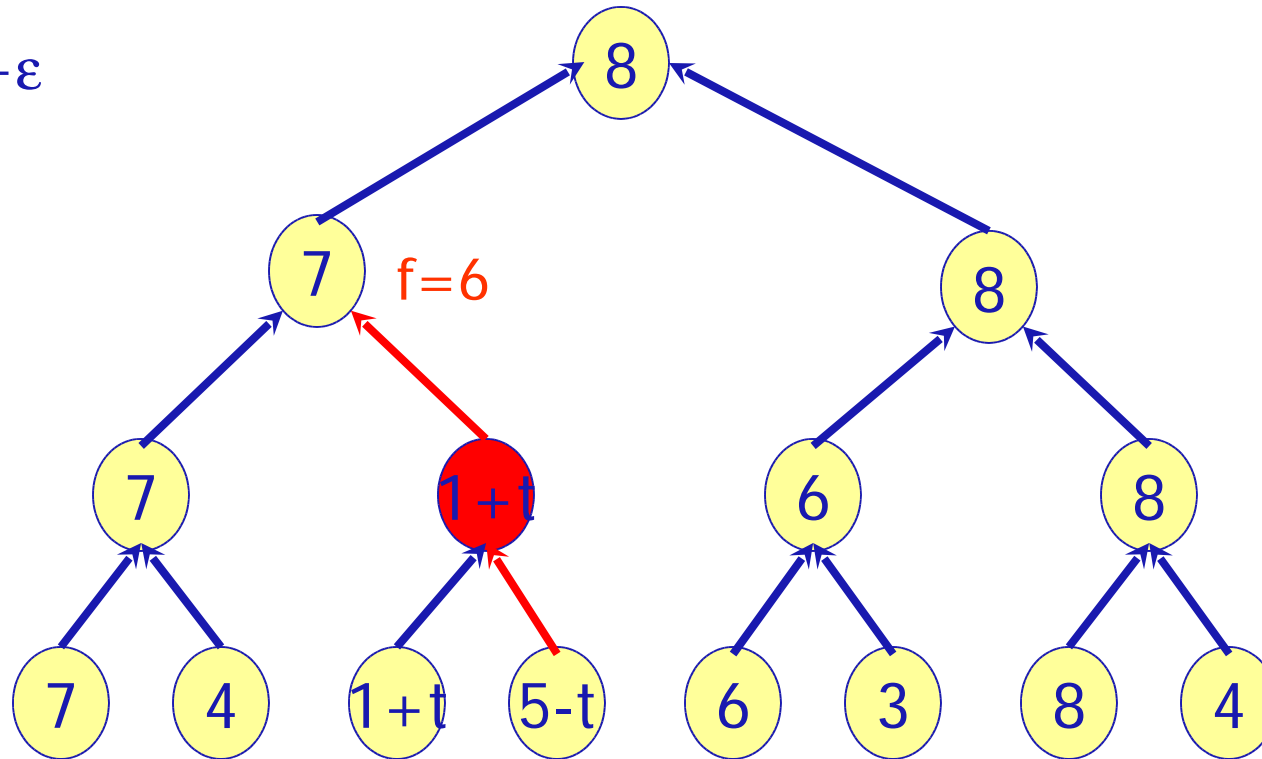
The Kinetic Tournament

Time = 0



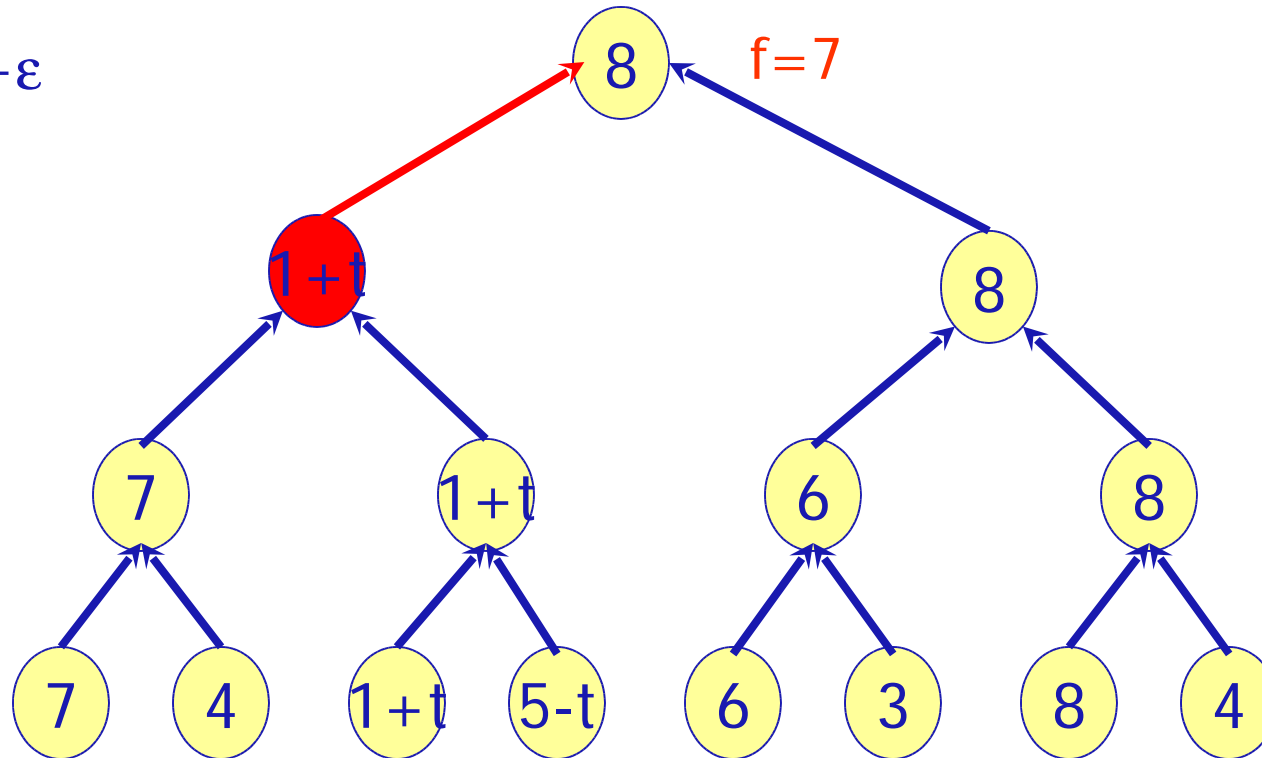
The Kinetic Tournament

Time = $2 + \epsilon$



The Kinetic Tournament

Time = $6 + \epsilon$



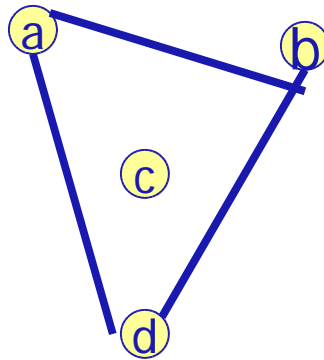


Performance of Kinetic Tournament

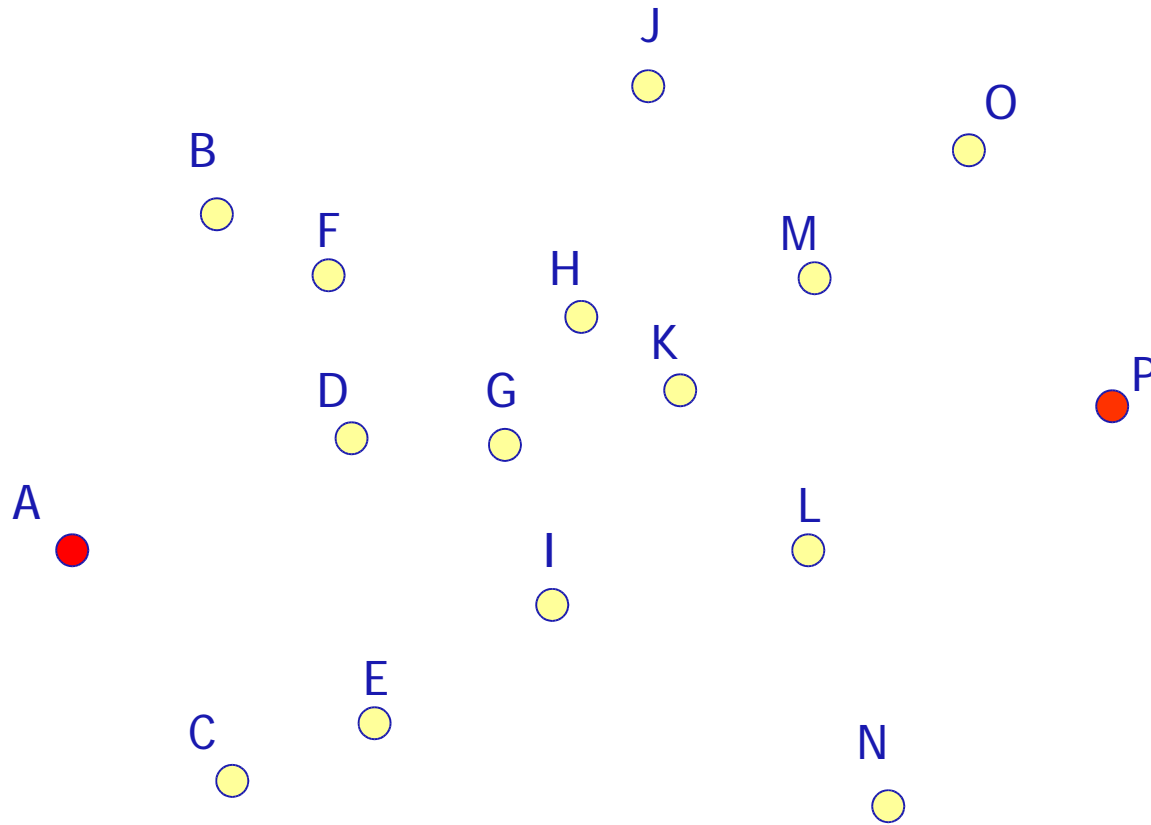
- Worst case $\log n$ time per event
- This kinetic algorithm is an adaptive version of the standard tournament algorithm for finding maximum

2-D Convex Hull

- Many algorithms: Quick Hull, Graham Scan, Incremental, Merge Hull, Ultimate, Improved Ultimate...
- We will focus on the Quick Hull algorithm
- Input: A list of points P
- Output: The boundary points on the hull of P
- Example: Input = $[a,b,c,d]$ Output = $[a,b,d]$

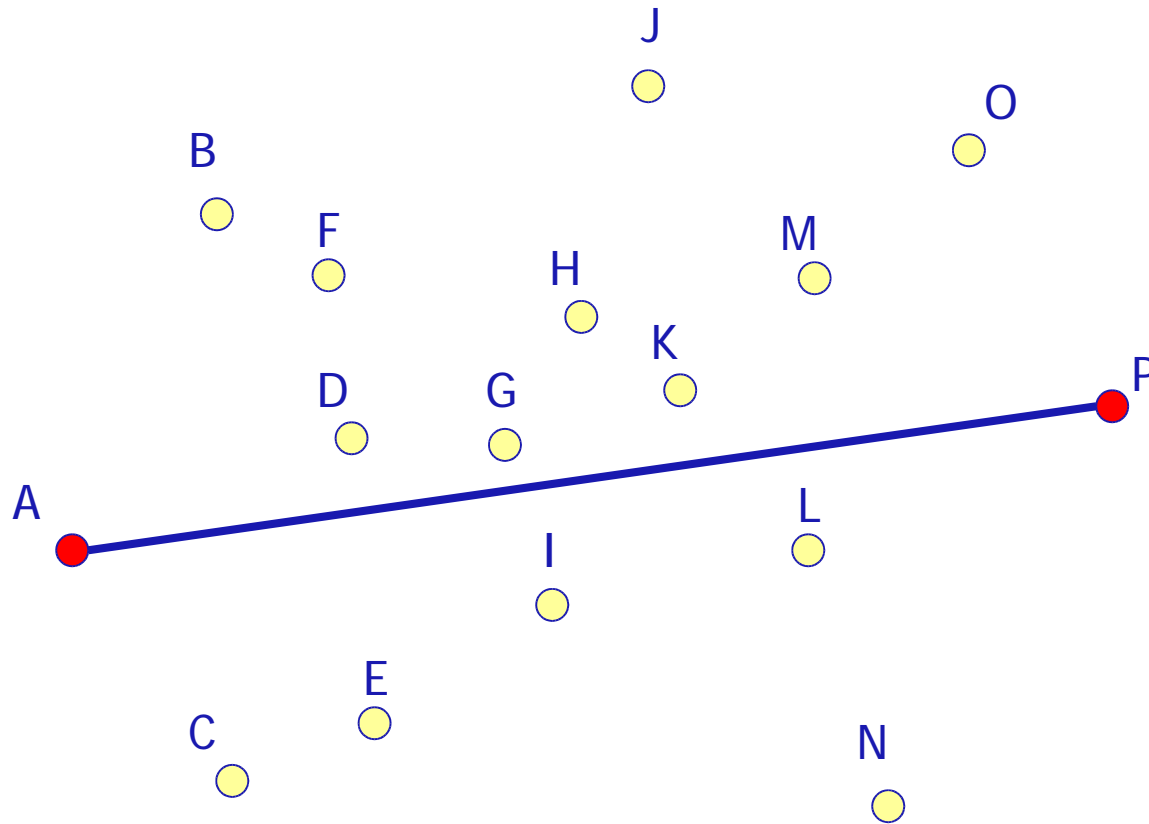


Quick Hull Example



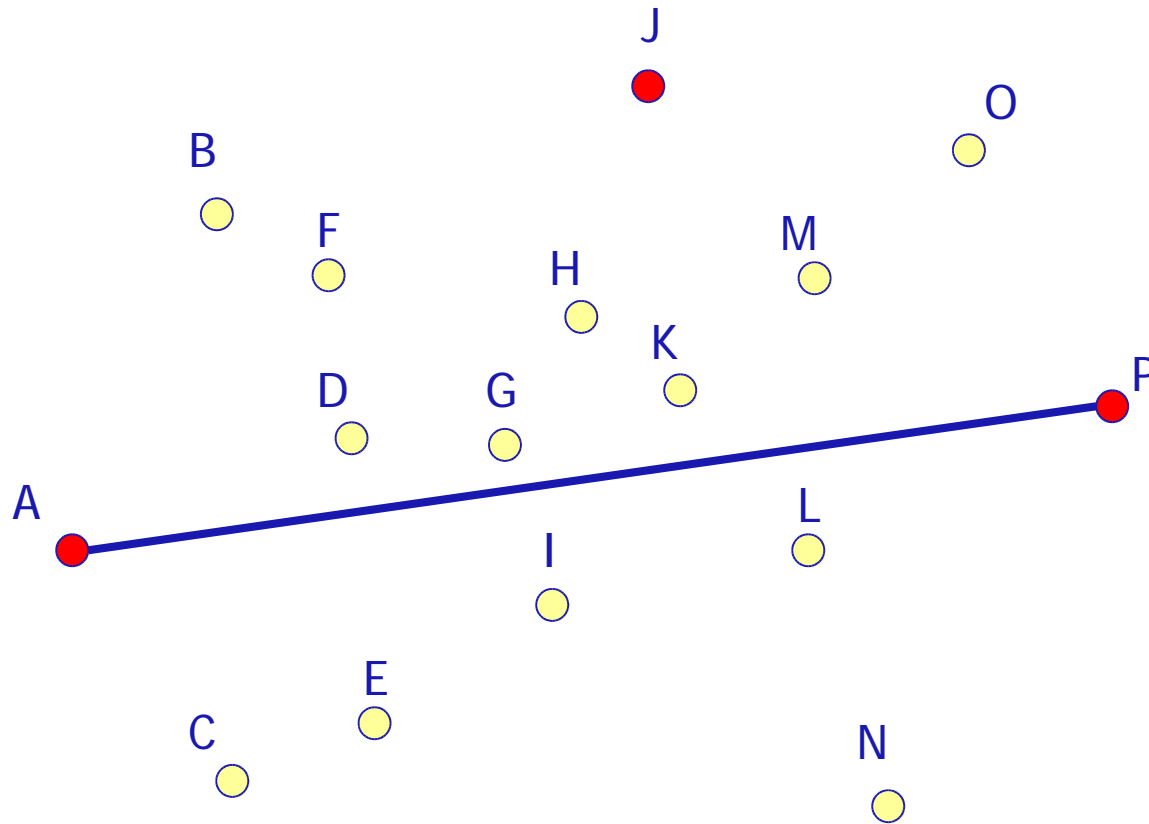
[A B C D E F G H I J K L M N O P]

Quick Hull Example - Filter



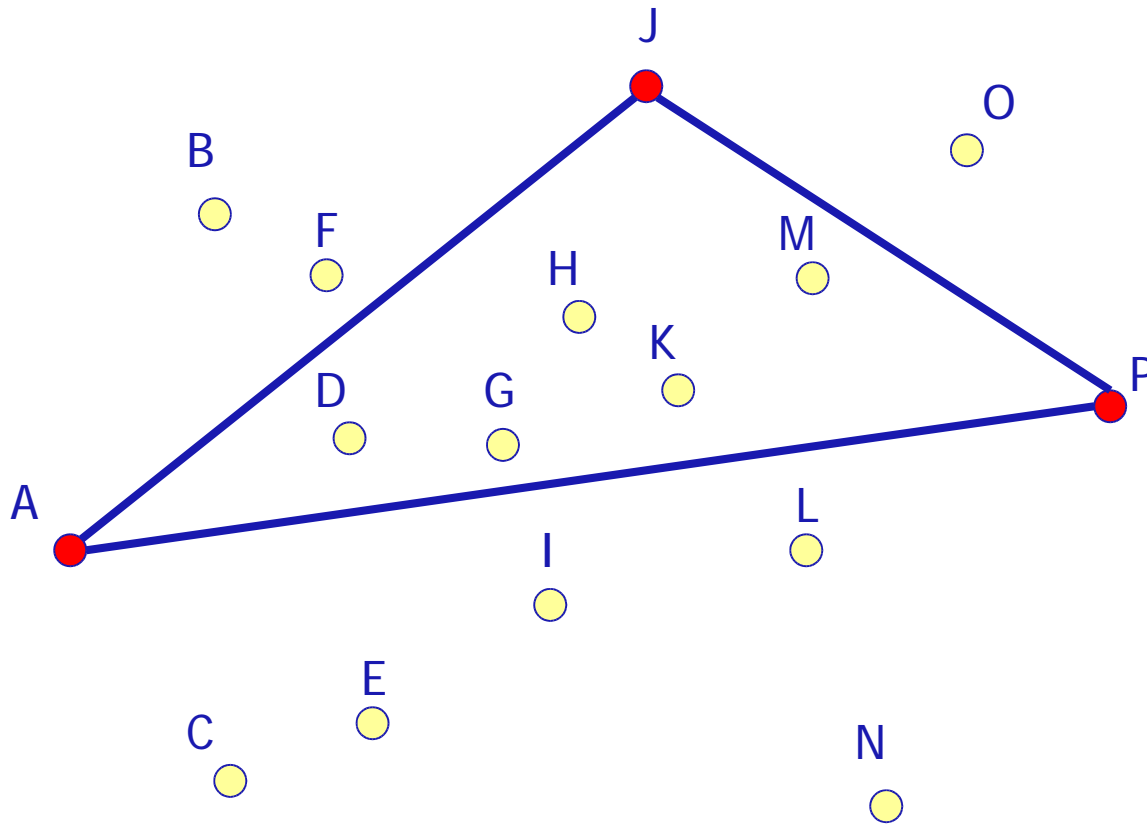
[A B D F G H J K M O P]

Quick Hull Example - Maximum



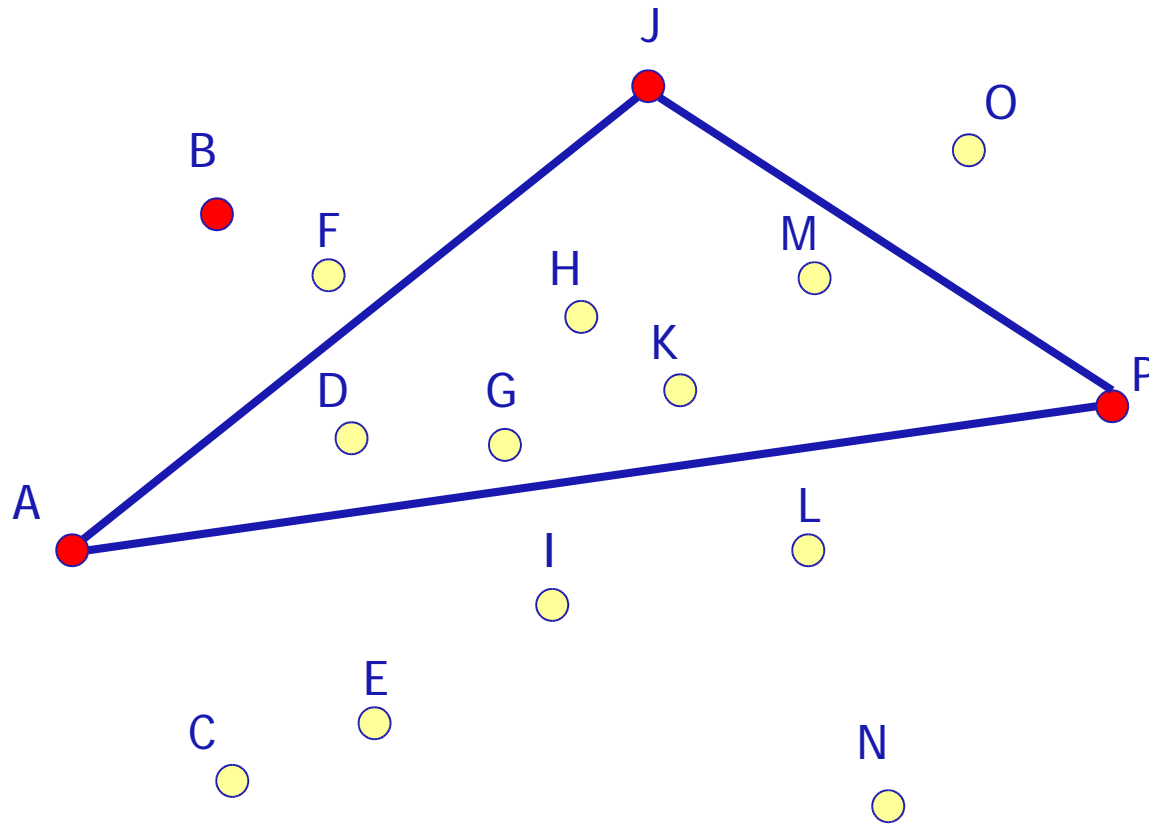
[A B D F G H J K M O P]

Quick Hull Example - Filter



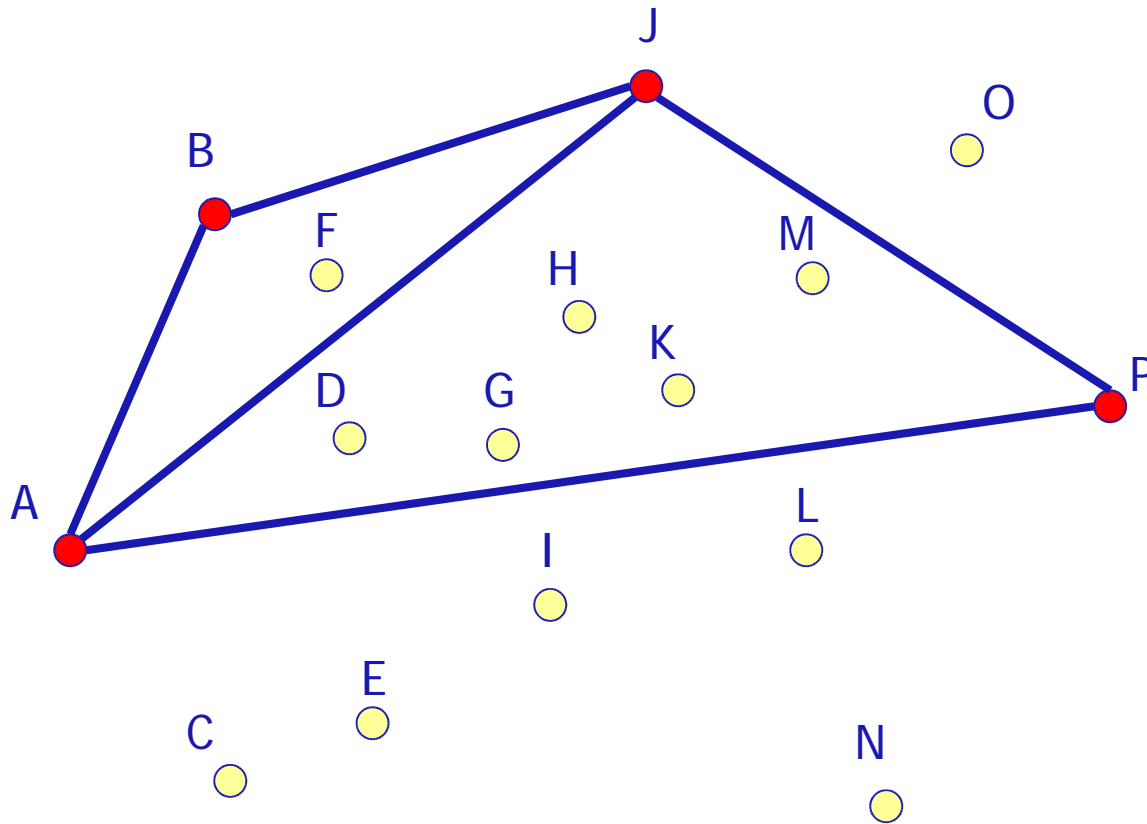
[[A B F J] [J O P]]

Quick Hull Example - Maximum



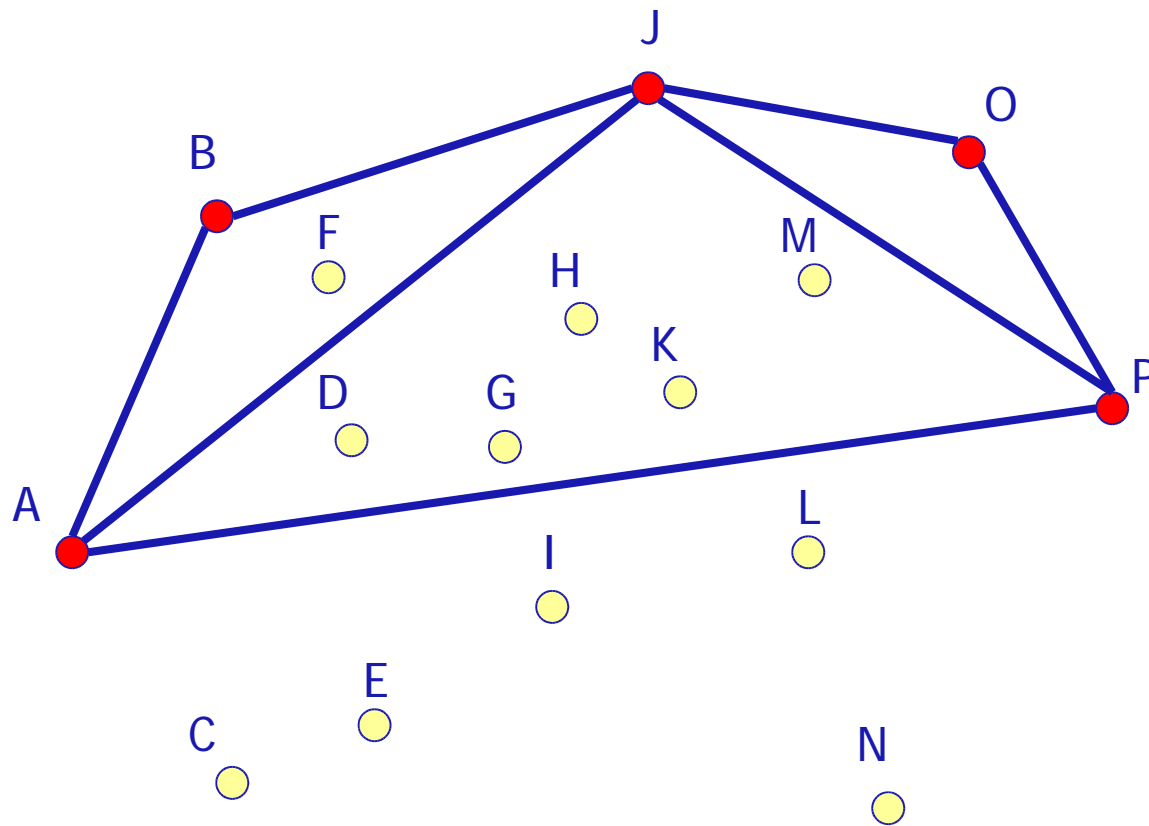
[[A B F J] [J O P]]

Quick Hull Example - Base Case



[[A B] [B J] [J O P]]

Quick Hull Example - Done



[[A B] [B J] [J O] [O P]]

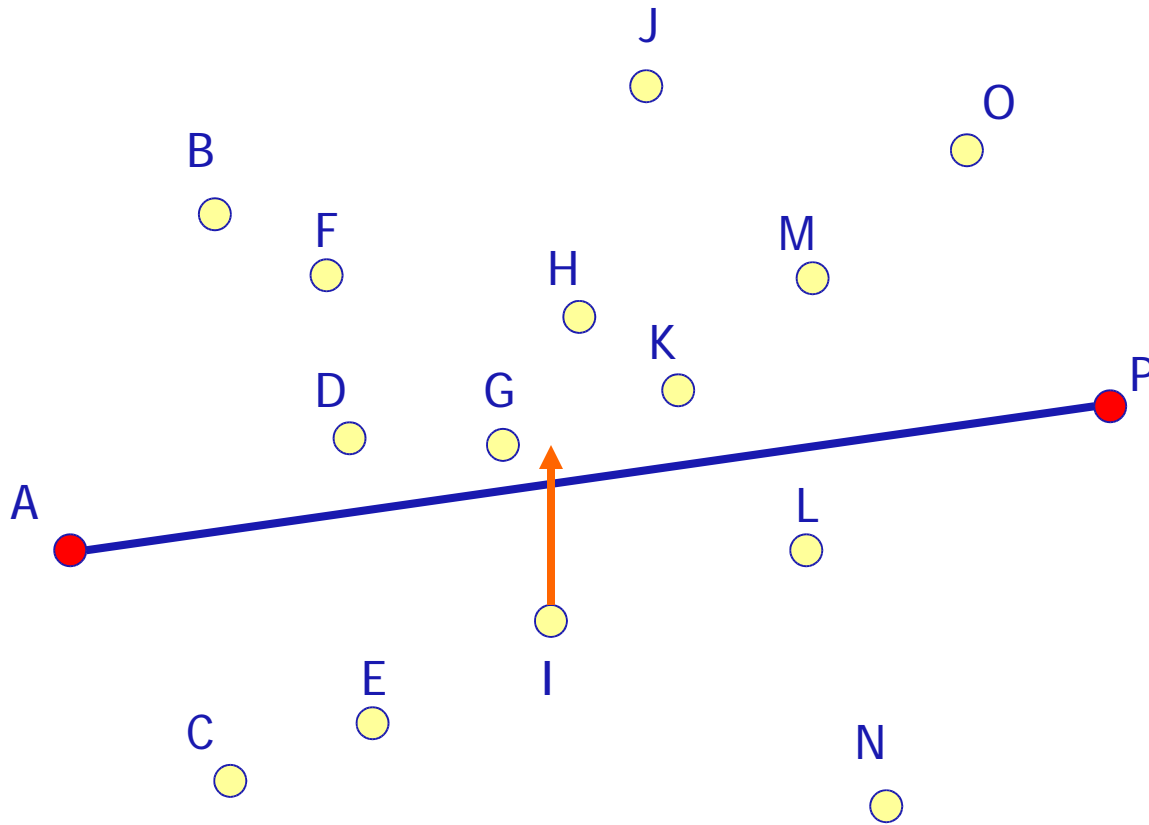


Kinetic Quick Hull

- Two kinds of tests: Line-side and distance comparisons
- Filtering => Line Side
- Finding the furthest point => Distance comparisons

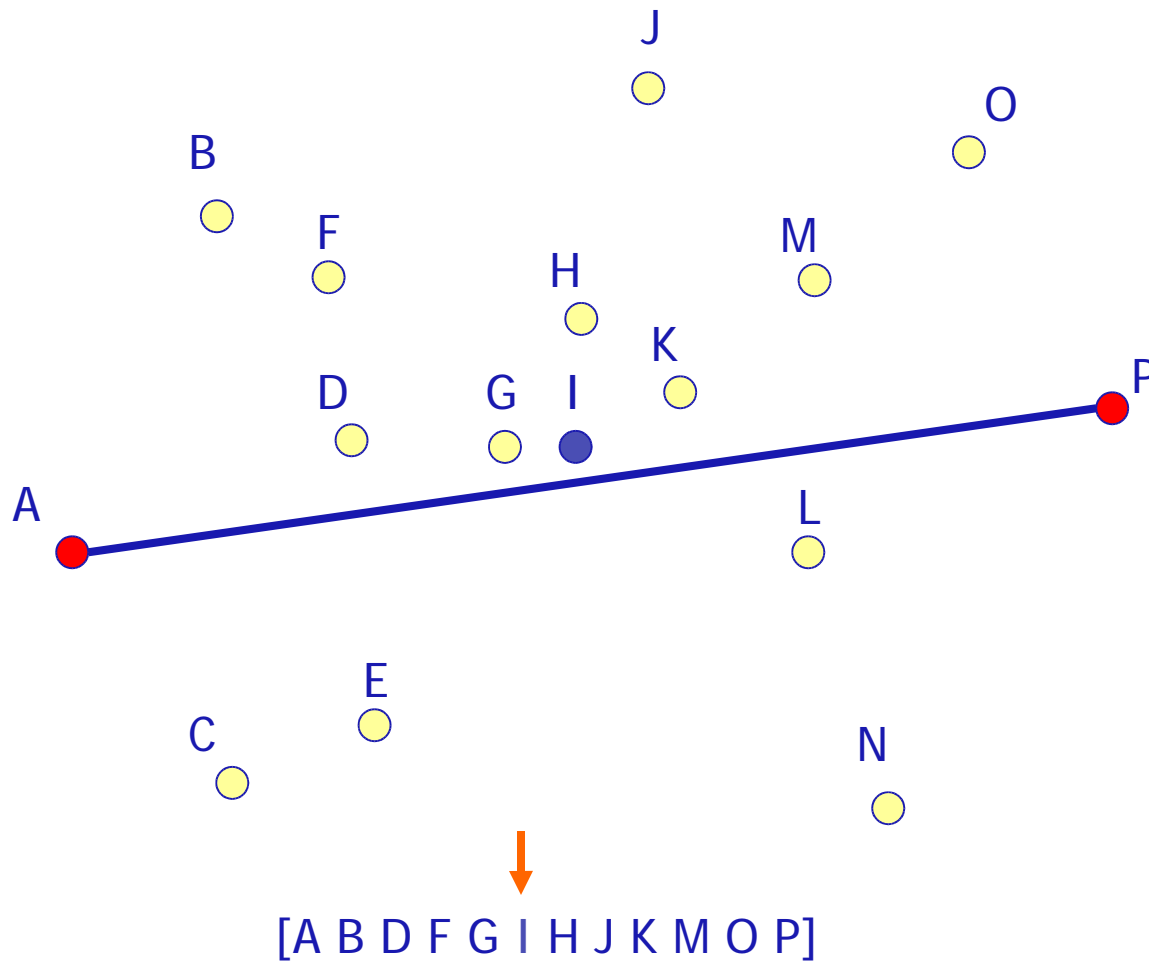
- Have certificates for these two events that is all

Line Side Test Fails

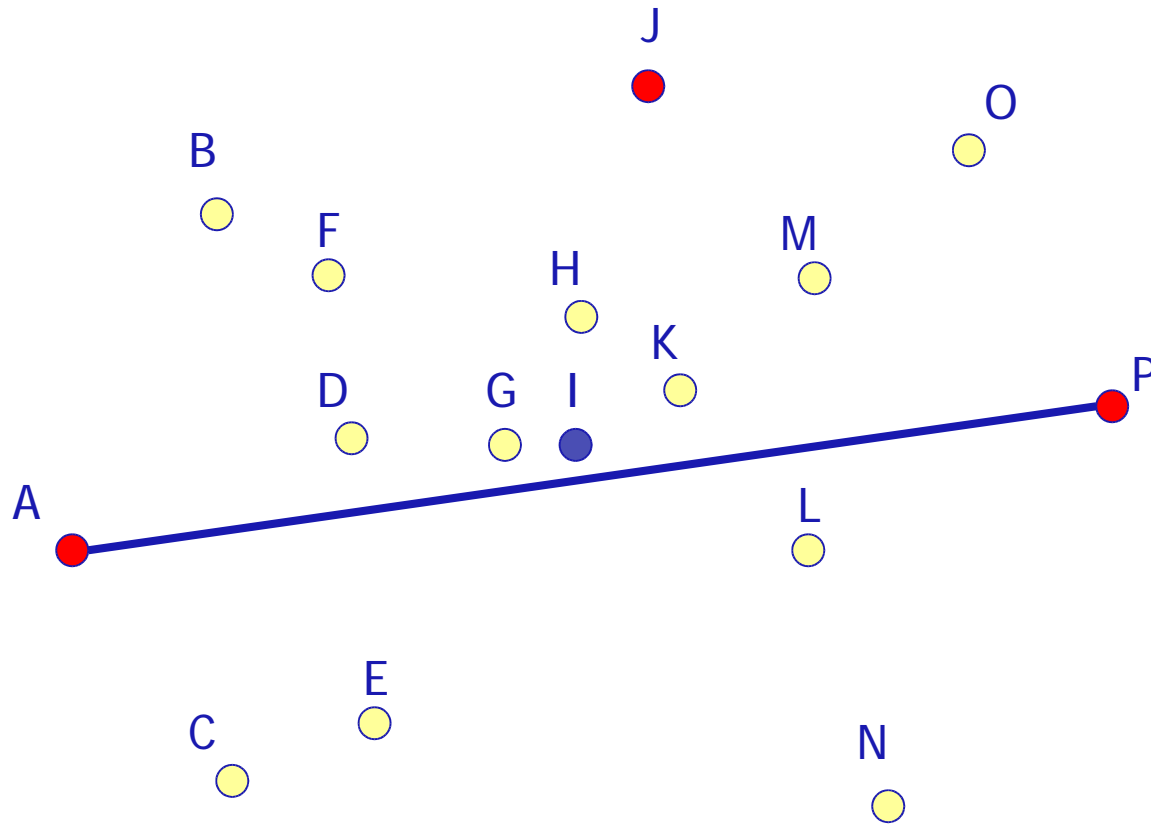


[A B D F G H J K M O P]

"I" is inserted in the middle of the list

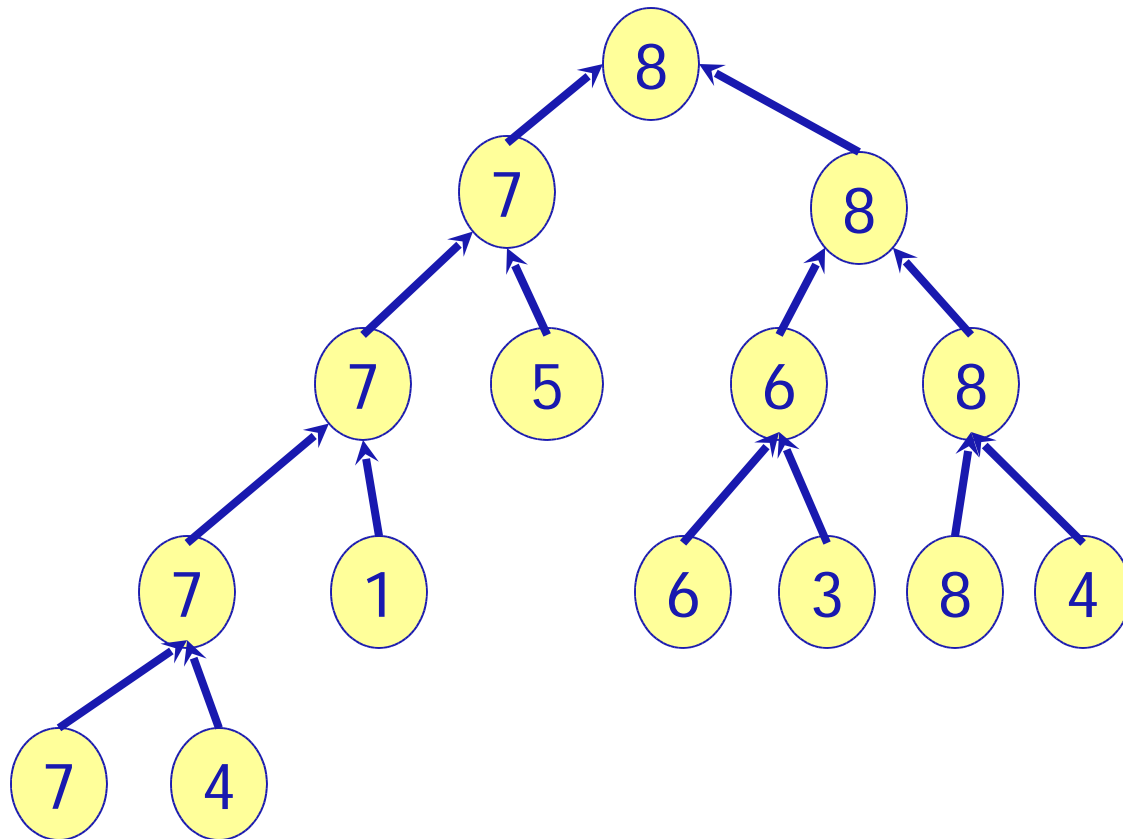


Recompute Maximum

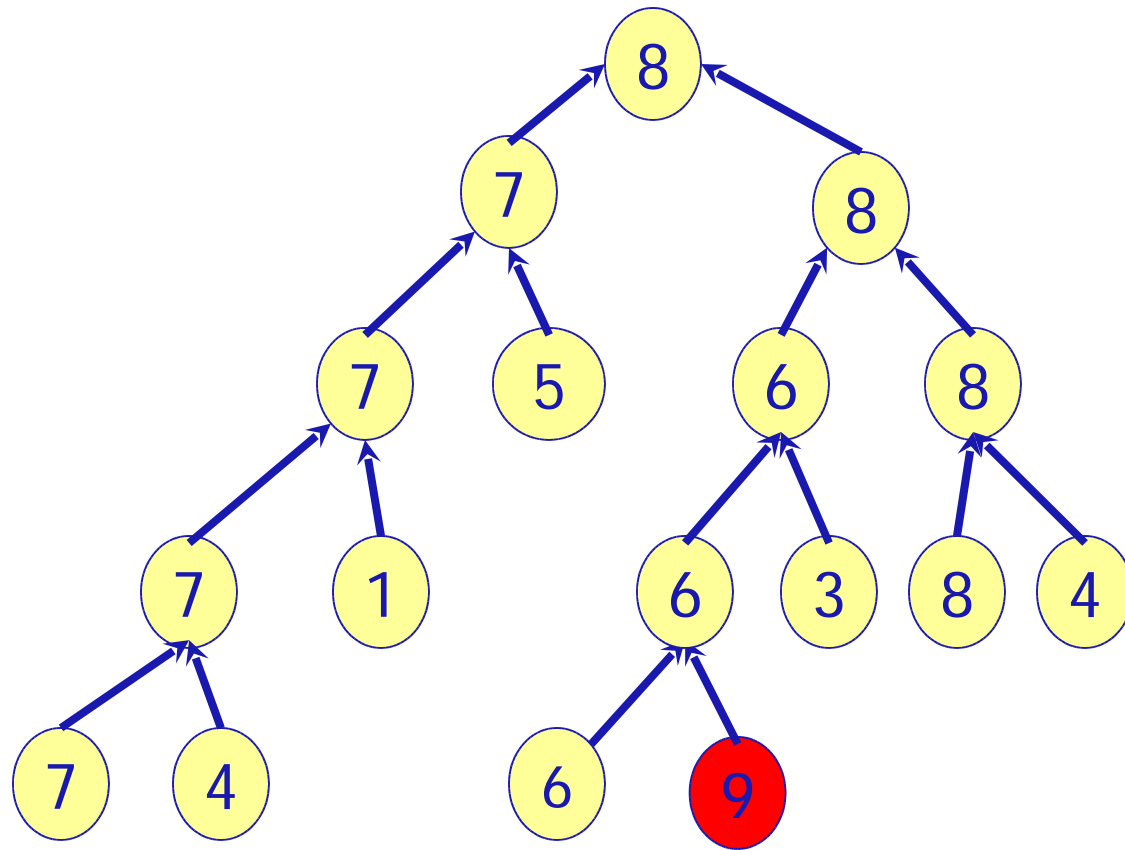


[A B D F G I H J K M O P]

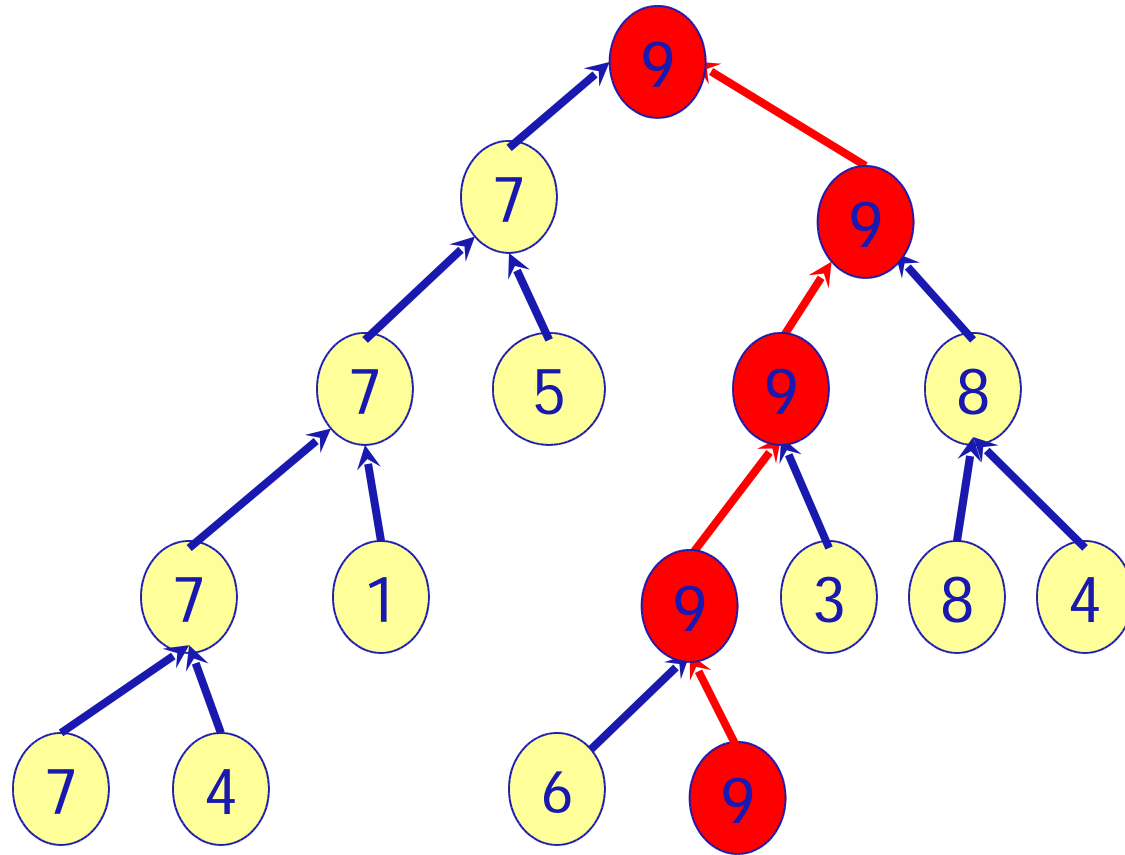
Dynamic Tournament - Random Trees



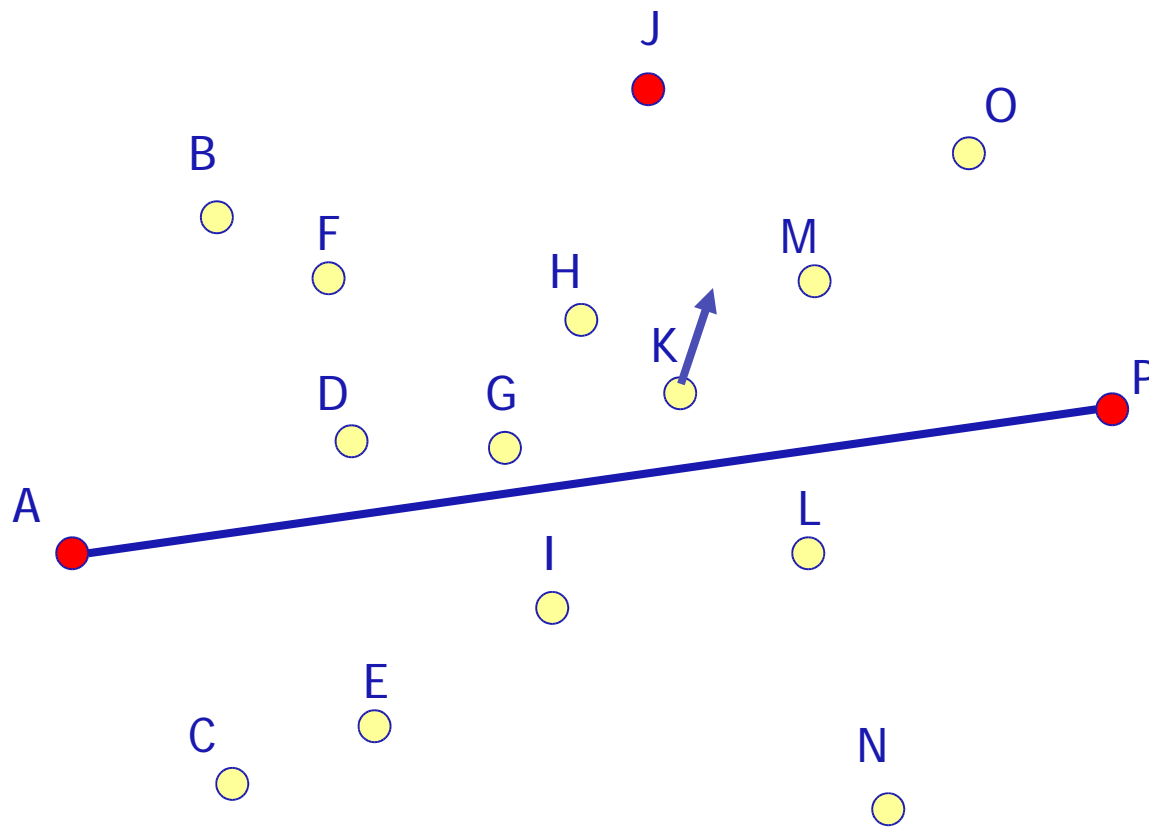
Dynamic Tournament - Random Trees



Dynamic Tournament - Random Trees

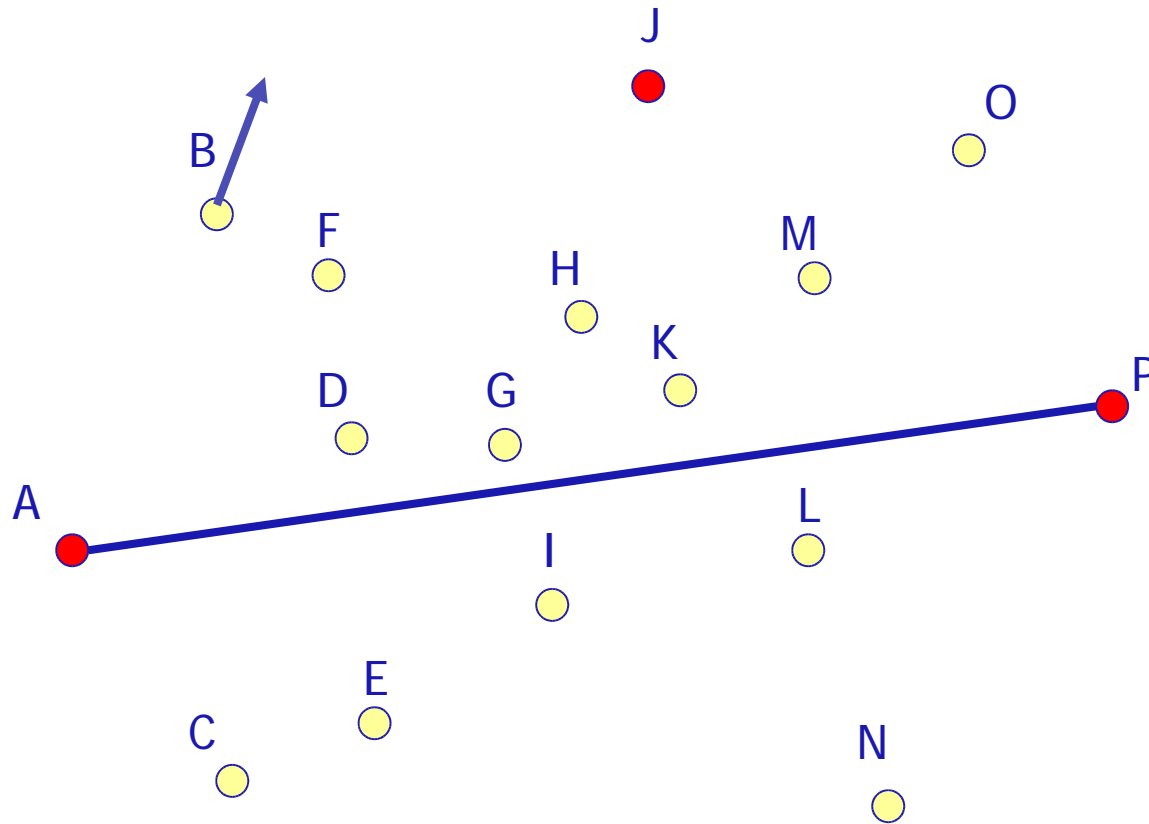


Distance Comparison Fails - Case 1



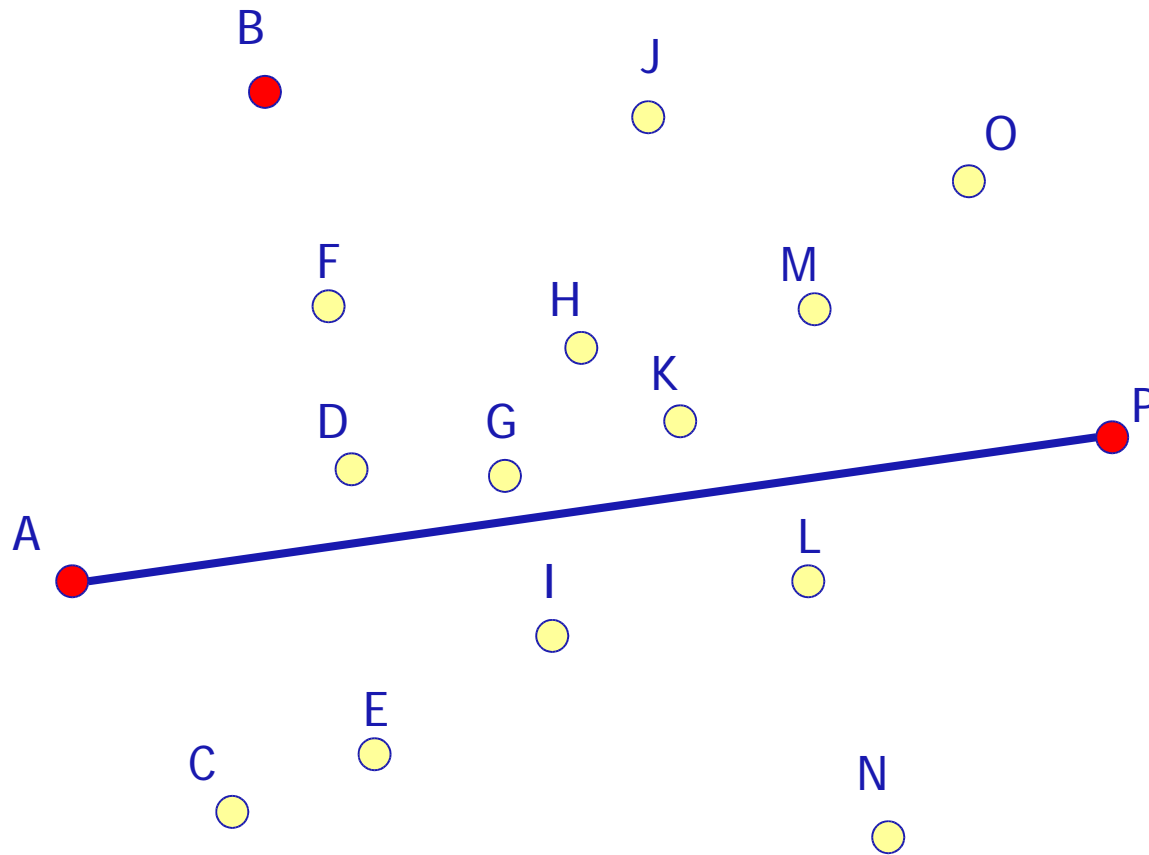
[A B D F G H J K M O P]

Distance Comparison Fails - Case 2



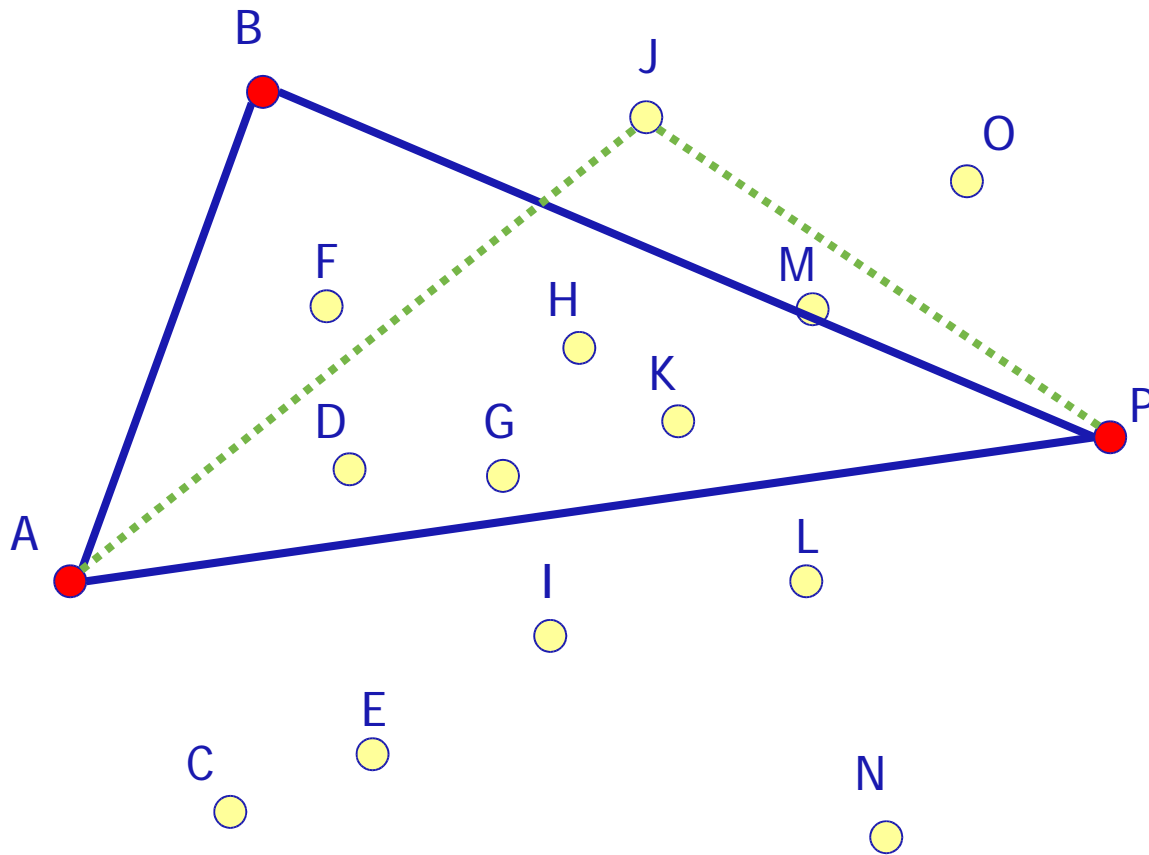
[A B D F G H J K M O P]

"B" is the new maximum



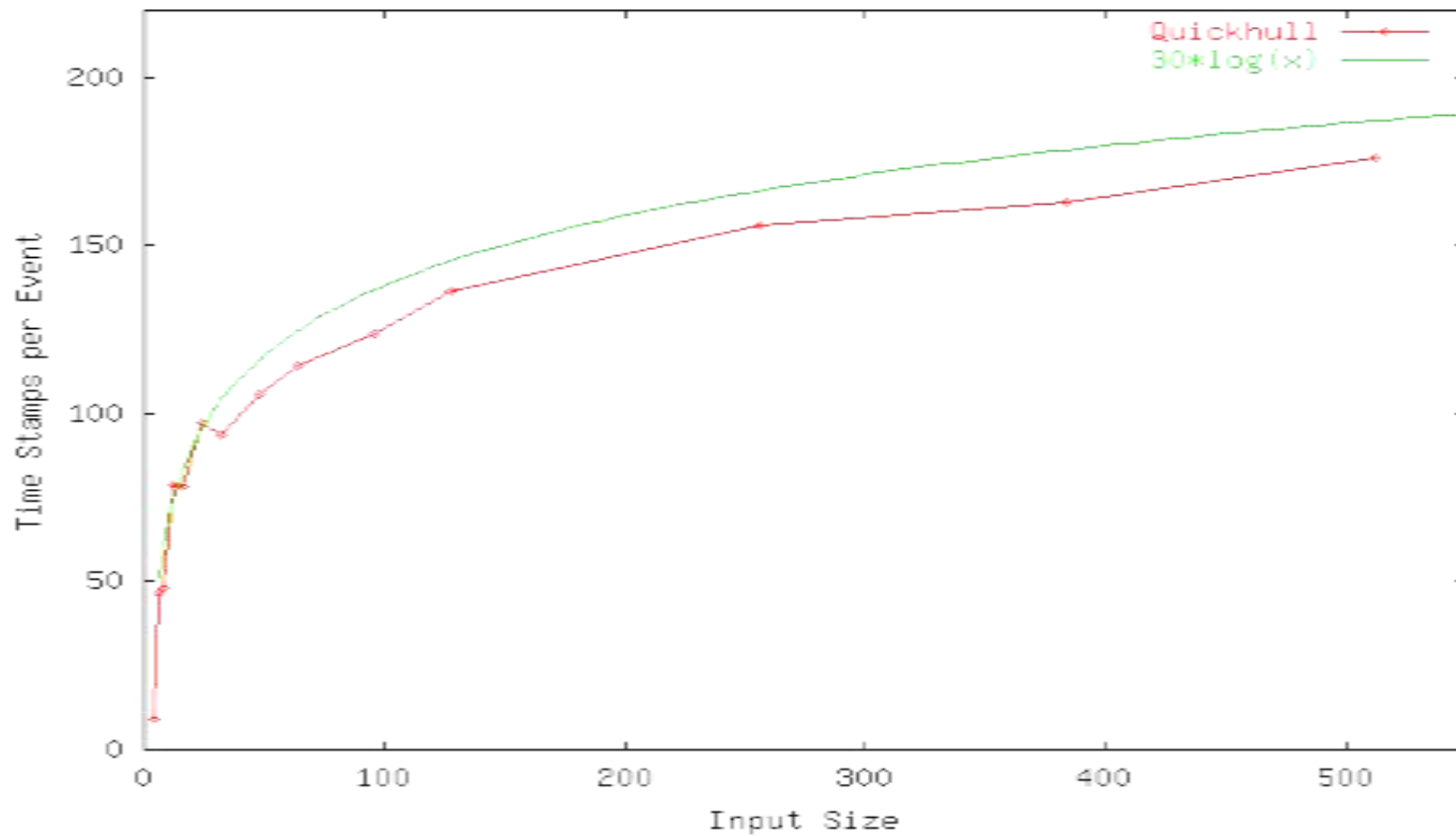
[A B D F G H J K M O P]

New recursive calls



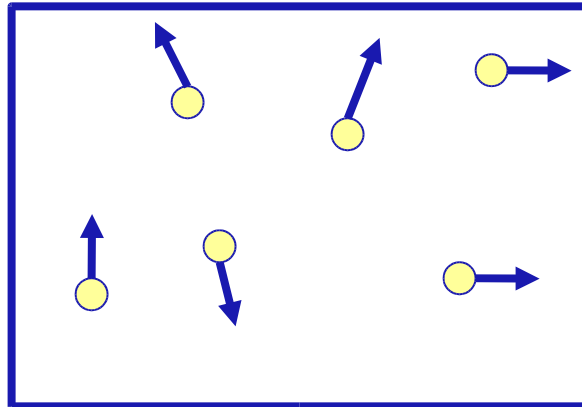
[[A B] [J M O]]

Experiments



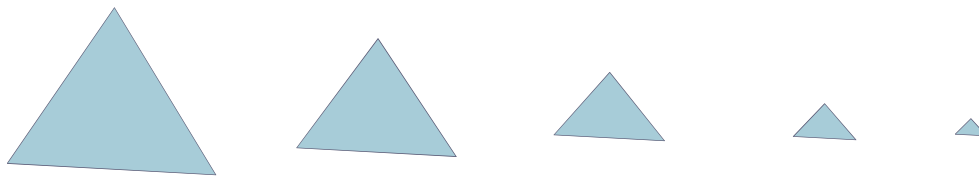
Summary of ConvexHull Work

- Kinetic Algorithms for convex hulls using adaptivity
 - Timothy Chan's $O(h \log n)$ algorithm: Improved "Ultimate Convex Hull": Have a working version
 - QuickHull
- Bounce events: Can maintain convex hull of points in a box - the points bounce off of the walls of the box
- Streamlined library for kinetic convex hulls in the SML language
 - A standard algorithm can be made kinetic in a few hours of work



Parallel Tree Contraction

- Fundamental technique [Miller & Reif '85]
- Contraction proceeds in rounds
 - Each round shrinks the tree by a constant factor
 - Expected $O(\log n)$ rounds



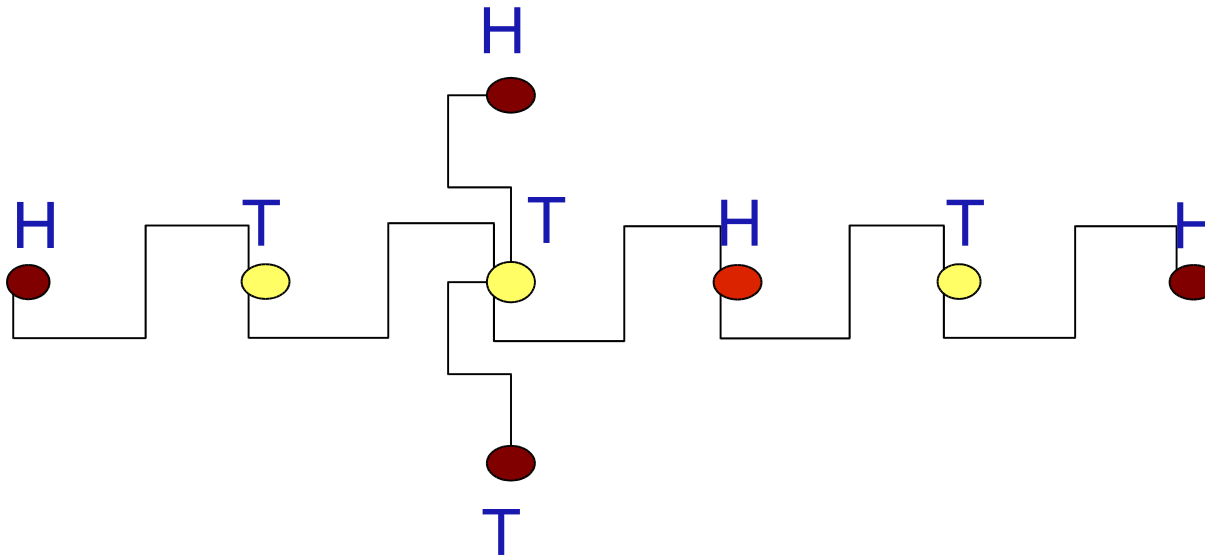
- Innovative Idea: Shrink the tree by local operations



Parallel Tree Contraction

- Start with a tree
- In each round:
 - Each node flips a coin
 - If leaf node then rake
 - If degree=2 and flip = H, and neighbors = T then contract
- Expected $O(\log n)$ rounds.

Contracting and Raking



Contracting and Raking




Contracting and Raking (cont.)





Done

H





Dynamic Trees Problem

- Given a forest of weighted trees
- Operations
 1. **Link**: edge insertion
 2. **cut**: edge deletion
 3. Queries
 - Heaviest edge in a subtree?
 - Heaviest edge on a path?



Data Structures for Dynamic Trees

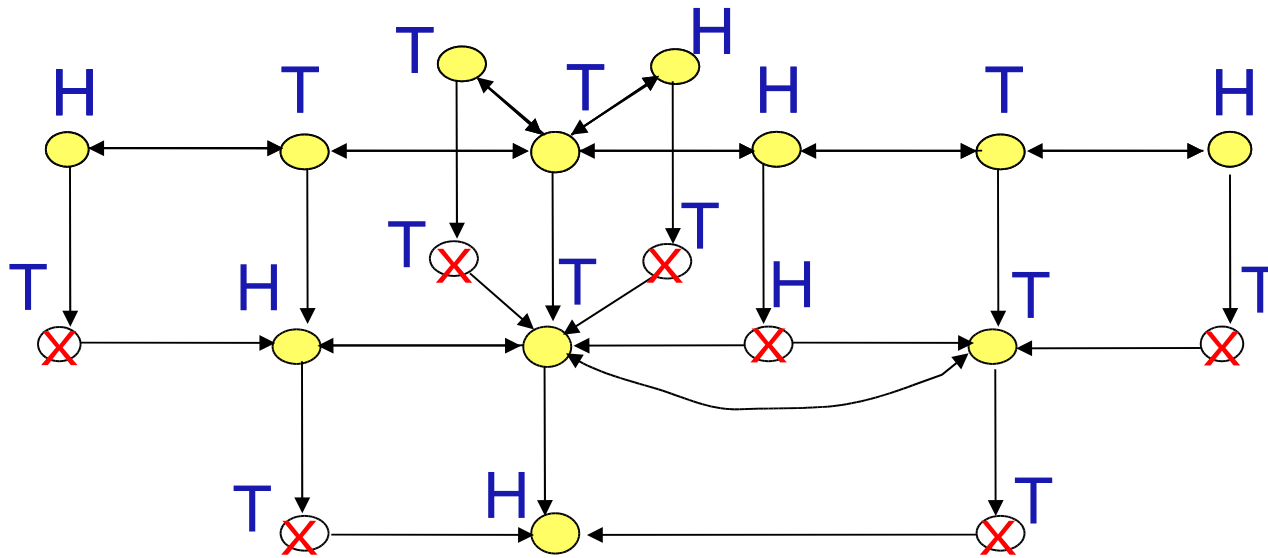
- Sleator Tarjan '85
 - Amortized $O(\log n)$ and worst-case $O(\log n)$
- Topology Trees [Frederickson '93]
 - Ternary (degree-tree) trees
 - Worst case $O(\log n)$
- Top Trees [AlstrupHoLiTh '97]
 - Generalize Topology Trees for arbitrary degree
- Idea: Trees as paths



Dynamic Parallel Tree Contraction

- Keep a copy of each round of the initial run.
- Each round affects next round.
- The nodes that “live” to the next round copy their neighbors scars, and pointers to them.
- Dependencies are based on what the node reads to do its work.

Dynamic Parallel Tree Contraction

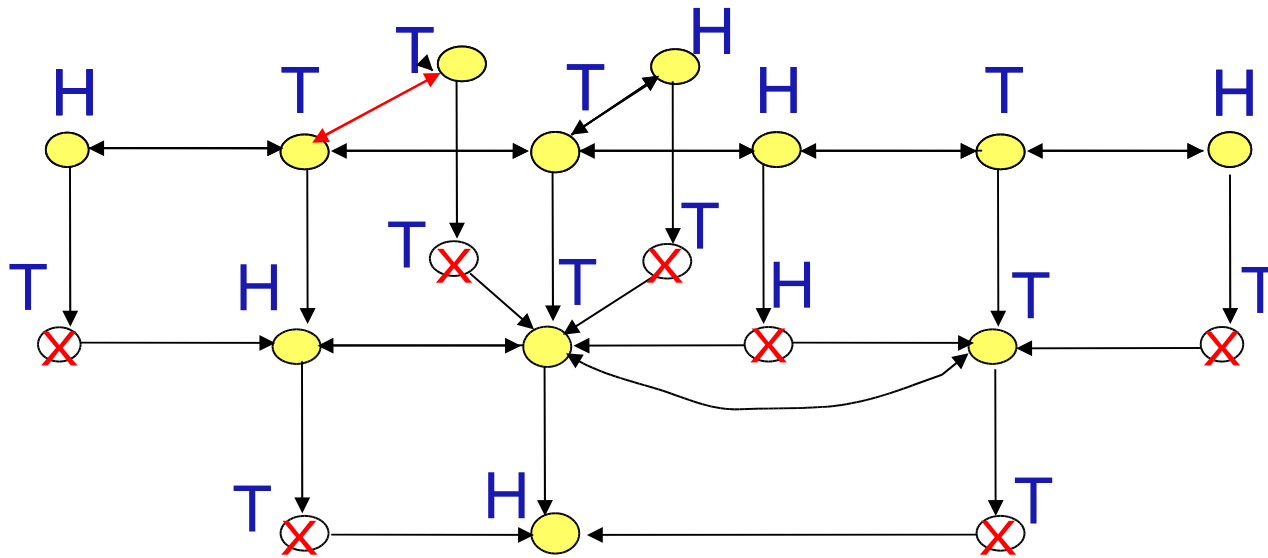




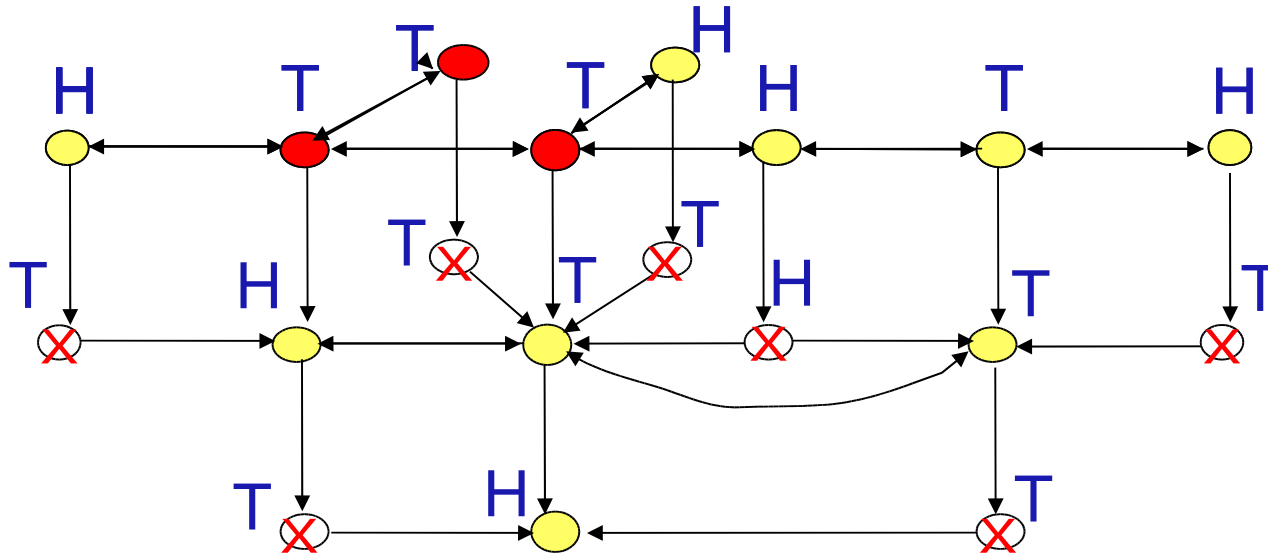
Propagation

- If any data changes nodes whose action depend on that data are woken up.
- Wake-up only those nodes that get affected by a change.
- Run same code as in original run.
- Expected constant amount of nodes woken up per round.

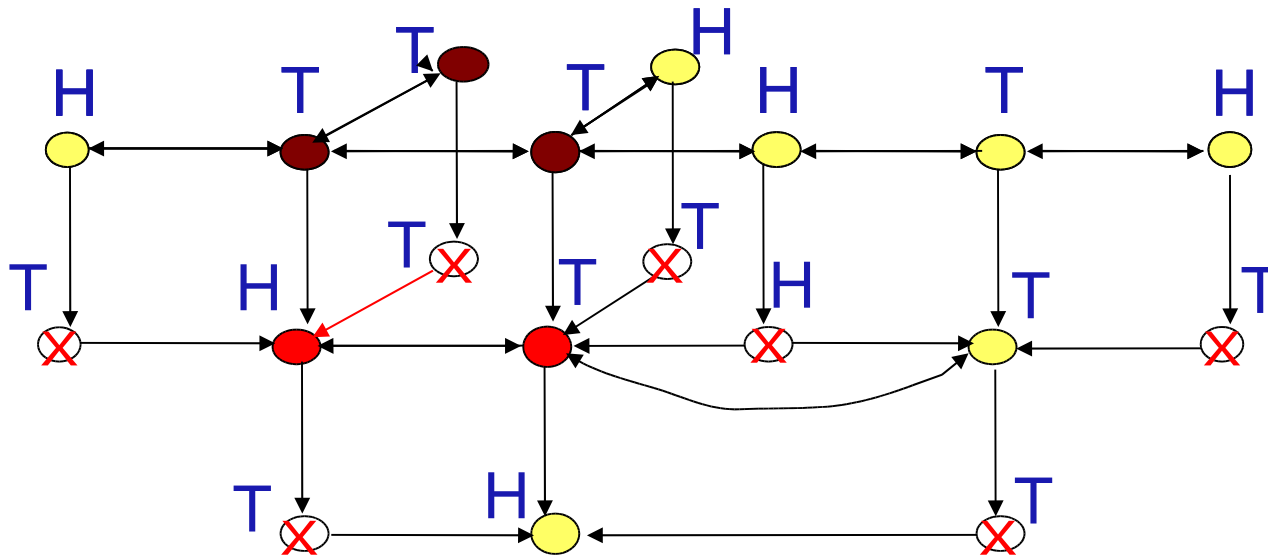
Change an edge



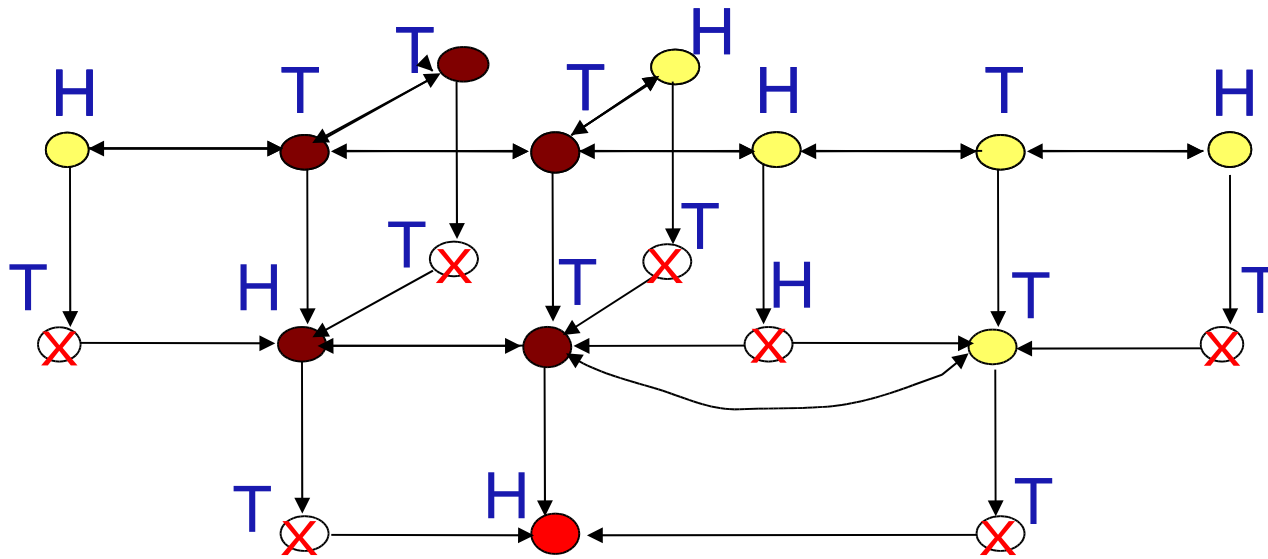
Three nodes woken up



Nodes rerun code, more nodes woken up

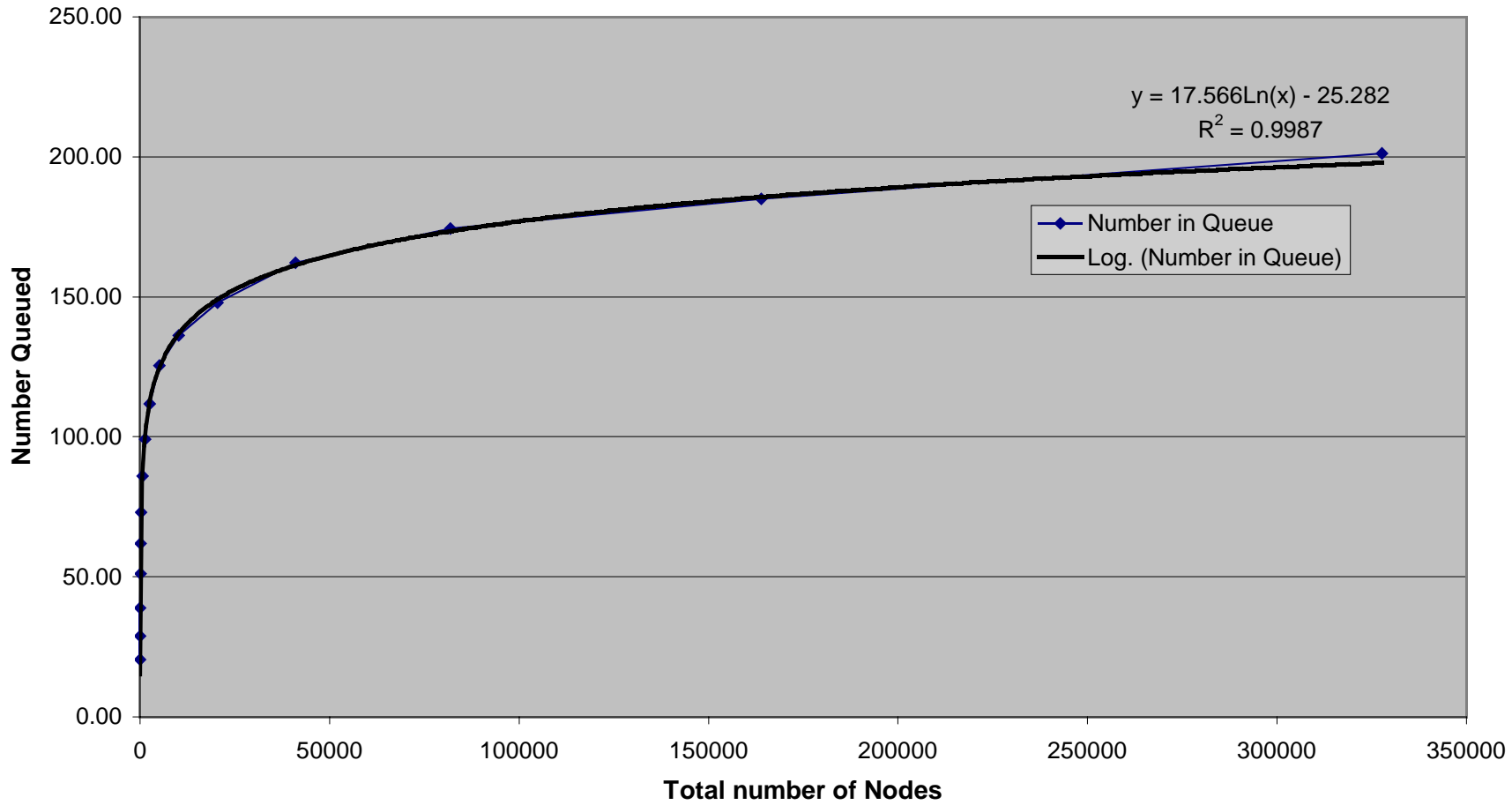


Propagation continued



Experimental Results

Number Queued when remarking Nodes





Work In Progress

- Analyzing Power of the data structure (what it can and cannot do)
- Different Applications
- Analyzing Running times for:
 - Different changes.
 - Unbalanced Trees.



Conclusion and Future Work

- **ConvexHull**
 - Used adaptivity to solve the kinetic convex hull problem.
 - Encouraging results.
 - Adaptivity makes writing dynamic/kinetic algorithms a simple edition on the standard algorithm
 - The quickhull algorithm updates based on events efficiently in the expected case.
- **Parallel Tree Contraction**
 - Efficient times $O(\log(n))$ expected time for an update.
 - Future Work:
 - More Applications