

So Far So Good

- Yesterday we discussed "dimension 1.5" problems:
 - Interval stabbing and point location
- We developed a number of useful tools/techniques
 - Logarithmic method
 - Weight-balanced B-trees
 - Global rebuilding
- On Thursday we also discussed several tools/techniques
 - B-trees
 - Persistent B-trees
 - Construction using buffer technique

Interval Management

• Maintain *N* intervals with unique endpoints dynamically such that stabbing query with point *x* can be answered efficiently



- Solved using external interval tree
- We obtained the same bounds as for the 1d case
 - Space: *O*(*N*/*B*)
 - Query: $O(\log_B N + T/B)$
 - Updates: $O(\log_B N)$ I/Os

Interval Management

- External interval tree:
 - Fan-out $\Theta(\sqrt{B})$ weight-balanced B-tree on endpoints
 - Intervals stored in O(B) secondary structure in each internal node
 - Query efficiency using filtering
 - Bootstrapping used to avoid O(B) search cost in each node
 - * Size $O(B^2)$ underflow structure in each node
 - * Constructed using sweep and persistent B-tree
 - * Dynamic using global rebuilding





3-Sided Range Searching

• Interval management corresponds to simple form of 2d range search



- More general problem: Dynamic 3-sidede range searching
 - Maintain set of points in plane such that given query (q_1, q_2, q_3) , all points (x,y) with $q_1 \le x \le q_2$ and $y \ge q_3$ can be found efficiently



3-Sided Range Searching : Static Solution

- Construction: Sweep top-down inserting *x* in persistent B-tree at (*x*,*y*)
 - -O(N/B) space
 - $-O(\frac{N}{B}\log_B N)$ I/O construction using buffer technique
- Query (q_1, q_2, q_3) : Perform range query with $[q_1, q_2]$ in B-tree at $q_3 O(\log_B N + T/B)$ I/Os
- Dynamic using logarithmic method
 - Insert: $O(\log_B^2 N)$
 - Query: $O(\log_B^2 N + T/B)$
- Improve to $O(\log_B N)$? Deletes?





- Decreasing *y* values on root-leaf path
- -(x,y) on path from root to leaf holding x
- If v holds point then parent(v) holds point







- Natural idea: Block tree
- Problem:
 - $-O(\log_B N)$ I/Os to follow paths to to q_1 and q_2
 - But O(T) I/Os may be used to visit other nodes ("overshooting")
 - $\Rightarrow O(\log_B N + T)$ query



- Base tree: Weight-balanced B-tree on *x*-coordinates (*a*,*k*=*B*)
- Points in "heap order":
 - Root stores B top points for each of the $\Theta(B)$ child slabs
 - Remaining points stored recursively
- Points in each node stored in " $O(B^2)$ -structure"

– Persistent B-tree structure for static problem

Linear space

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- Query with (q_1, q_2, q_3) starting at root v:
 - Query $O(B^2)$ -structure and report points satisfying query
 - Visit child v if
 - * *v* on path to q_1 or q_2
 - * All points corresponding to *v* satisfy query



Analysis:

- $O(\log_B B^2 + \frac{T_v}{R}) = O(1 + \frac{T_v}{R})$ I/Os used to visit node v
- $-O(\log_B N)$ nodes on path to q_1 or q_2
- For each node v not on path to q_1 or q_2 visited, B points reported in *parent*(*v*)

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- Insert (*x*, *y*) (assuming fixed *x*-coordinate set static base tree):
 - Find relevant node *v*:
 - * Query O(B²)-structure to find
 B points in root corresponding
 to node u on path to x
 - * If *y* smaller than *y*-coordinates of all *B* points then recursively search in *u*



- Insert (*x*, *y*) in $O(B^2)$ -structure of *v*
- If $O(B^2)$ -structure contains >B points for child u, remove lowest point and insert recursively in u
- Delete: Similarly

- Analysis:
 - Query visits $O(\log_B N)$ nodes
 - $O(B^2)$ -structure queried/updated in each node
 - * One query
 - * One insert and one delete
- $O(B^2)$ -structure analysis:
 - Query: $O(\log_B B^2 + B/B) = O(1)$
 - Update in O(1) I/Os using update
 block and global rebuilding



 $O(\log_B N)$ I/Os

Removing Fixed *x***-coordinate Set Assumption**

- Deletion:
 - Delete point as previously
 - Delete *x*-coordinate from base
 - tree using global rebuilding
- $\Rightarrow O(\log_B N)$ I/Os amortized
- Insertion:
 - Insert *x*-coordinate in base tree and rebalance (using splits)
 - Insert point as previously



Split: Boundary in *v* becomes boundary in *parent*(*v*)

Removing Fixed *x***-coordinate Set Assumption**

- Split: When *v* splits *B* new points needed in *parent*(*v*)
- One point obtained from *v*' (*v*'') using "bubble-up" operation:
 - Find top point p in v'
 - Insert p in $O(B^2)$ -structure
 - Remove *p* from $O(B^2)$ -structure of *v*'
 - Recursively bubble-up point to v
- Bubble-up in $O(\log_B w(v))$ I/Os
 - Follow one path from *v* to leaf
 - Uses O(1) I/O in each node



Split in $O(B \log_B w(v)) = O(w(v))$ I/Os

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Removing Fixed *x***-coordinate Set Assumption**

- *O*(*1*) amortized split cost:
 - Cost: O(w(v))

– Weight balanced base tree: $\Omega(w(v))$ inserts below v between splits

- External Priority Search Tree
 - Space: *O*(*N*/*B*)
 - Query: $O(\log_B N + T/B)$
 - Updates: $O(\log_B N)$ I/Os amortized



- Amortization can be removed from update bound in several ways
 - Utilizing lazy rebuilding

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Summary: 3-sided Range Searching

- 3-sidede range searching
 - Maintain set of points in plane such that given query (q_1, q_2, q_3) , all points (x,y) with $q_1 \le x \le q_2$ and $y \ge q_3$ can be found efficiently



- We obtained the same bounds as for the 1d case
 - Space: *O*(*N*/*B*)
 - Query: $O(\log_B N + T/B)$
 - Updates: $O(\log_B N)$ I/Os

Summary: 3-sided Range Searching

• Main problem in designing external priority search tree was the increased fanout in combination with "overshooting"



- Same general solution techniques as in interval tree:
 - Bootstrapping:

* Use $O(B^2)$ size structure in each internal node

- * Constructed using persistence
- * Dynamic using global rebuilding
- Weight-balanced B-tree: Split/fuse in amortized O(1)
- Filtering: Charge part of query cost to output

Two-Dimensional Range Search

- We have now discussed structures for special cases of twodimensional range searching
 - Space: *O*(*N*/*B*)
 - Query: $O(\log_B N + T/B)$
 - Updates: $O(\log_B N)$



 q_3

 q_1

 q_2

- Cannot be obtained for general 2d range searching:
 - $-O(\log_B^c N)$ query requires $\Omega(\frac{N}{B} \frac{\log_B N}{\log_B \log_B N})$ space



• Base tree: Fan-out $\Theta(\log_B N)$ weight balanced tree on *x*-coordinates $\bigcup_{k=1}^{N}$

 $O(\frac{\log_B N}{\log_B \log_B N})$ height

- Points below each node stored in 4 linear space secondary structures:
 - "Right" priority search tree
 - "Left" priority search tree
 - B-tree on y-coordinates
 - Interval tree

$$\bigcup_{\Omega(\frac{N}{B}\frac{\log_B N}{\log_B \log_B N})}$$
 space



- Secondary interval tree structure:
 - Connect points in each slab in y-order
 - Project obtained segments in y-axis



- Intervals stored in interval tree
 - * Interval augmented with pointer to corresponding points in *y*-coordinate B-tree in corresponding child node

- Query with (q_1, q_2, q_3, q_4) answered in top node with q_1 and q_2 in different slabs v_1 and v_2
- Points in slab v_1
 - Found with 3-sided query in v_1 using right priority search tree
- Points in slab v_2
 - Found with 3-sided query in v_2 using left priority search tree
- Points in slabs between v_1 and v_2



- Answer stabbing query with q_3 using interval tree
 - \Rightarrow first point above q_3 in each of the $O(\log_B N)$ slabs
- Find points using y-coordinate B-tree in $O(\log_B N)$ slabs



• Query analysis:

- $-O(\log_B N)$ I/Os to find relevant node
- $-O(\log_B N + T/B)$ I/Os to answer two 3-sided queries
- $O(\log_B N + \frac{\log_B N}{B}) = O(\log_B N)$ I/Os to query interval tree
- $O(\log_B N + T/B)$ I/Os to traverse $O(\log_B N)$ B-trees



- Insert:
 - Insert *x*-coordinate in weight-balanced B-tree
 - * Split of v can be performed in $O(w(v)\log_B w(v))$ I/Os

$$\Rightarrow O(\frac{\log_B^2 N}{\log_B \log_B N})$$
 I/Os

- Update secondary structures in all $O(\frac{\log_B N}{\log_B \log_B N})$ nodes on one root-leaf path

* Update priority search trees

- * Update interval tree
- * Update B-tree

$$\Rightarrow O(\frac{\log_B^2 N}{\log_B \log_B N}) \text{I/Os}$$

• Delete:



– Similar and using global rebuilding

Summary: External Range Tree

- 2d range searching in $O(\frac{N}{B} \frac{\log_B N}{\log_R \log_R N})$ space
 - $O(\log_B N + T/B) I/O query$ $- O(\frac{\log_B N}{\log_B \log_B N}) I/O update$
- Optimal among $O(\log_B N + T/B)$ query structures





- kd-tree:
 - Recursive subdivision of point-set into two half using vertical/horizontal line
 - Horizontal line on even levels, vertical on uneven levels
 - One point in each leaf

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Linear space and logarithmic height



- Query:
 - Recursively visit node corresponding to regions intersected query
 - Report point in trees/nodes completely contained in query
- Analysis:
 - Number of regions intersecting horizontal line satisfy recurrence $Q(N) = 2 + 2Q(N/4) \Rightarrow Q(N) = O(\sqrt{N})$
 - Query intersects $4 \cdot O(\sqrt{N}) + T = O(\sqrt{N} + T)$ regions



• KdB-tree:

– Blocking of kd-tree but with *B* point in each leaf

• Query as before

- Analysis as before except that each region now contains *B* points $\downarrow = O(\sqrt{N/B} + T/B)$ I/O query





O-Tree Query

- Perform rangesearch with q_1 and q_2 in vertical B-tree
 - Query all kdB-trees in leaves of two horizontal B-trees with xinterval intersected but not spanned by query
 - Perform rangesearch with q_3 and q_4 horizontal B-trees with *x*-interval spanned by query
 - * Query all kdB-trees with range intersected by query



O-Tree Query Analysis

- Vertical B-tree query: $O(\log_B(\sqrt{N/B}/\log_B N)) = O(\sqrt{N/B})$
- Query of all kdB-trees in leaves of two horizontal B-trees: $O(\sqrt{N/B}/\log_B N) \cdot O(\sqrt{B\log_B^2 N/B} + \frac{T}{B}) = O(\sqrt{N/B} + \frac{T}{B})$
- Query $O(\sqrt{N_B}/\log_B N)$ horizontal B-trees: $O(\sqrt{N_B}/\log_B N) \cdot O(\log_B(\sqrt{N_B}/\log_B N)) = O(\sqrt{N_B})$
- Query $2 \cdot O(\sqrt{N/B} / \log_B N)$ kdB-trees not completely in query $2 \cdot O(\sqrt{N/B} / \log_B N) \cdot O(\sqrt{B \log_B^2 N/B} + \frac{T}{B}) = O(\sqrt{N/B} + \frac{T}{B})$
- Query in kdB-trees completely contained in query: O(^T/_B)
 ↓

$$O(\sqrt{N/B} + \frac{T}{B})$$
 I/Os



O-Tree Update

• Insert:

– Search in vertical B-tree: $O(\log_B N)$ I/Os

– Search in horizontal B-tree: $O(\log_B N)$ I/Os

- Insert in kdB-tree: $O(\log_B^2 (B \log_B^2 N)) = O(\log_B N)$ I/Os
- Use global rebuilding when structures grow too big/small
 - B-trees not contain $\Theta(\sqrt{N_B}/\log_B N)$ elements

- kdB-trees not contain $\Theta(B \log_B^2 N)$ elements

 $O(\log_B N)$ I/Os

 Deletes can be handled in O(log_B N) I/Os similarly

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• Can be extended to work in *d*-dimensions with optimal query bound $O((\frac{N}{B})^{1-\frac{1}{d}} + \frac{T}{B})$

Summary: 3 and 4-sided Range Search

3-sided 2d range searching: External priority search tree $-O(\log_B N + T/B)$ query, $O(\frac{N}{R})$ space, $O(\log_B N)$ update

General (4-sided) 2d range searching:

- External range tree: $O(\log_B N + T/B)$ query, $\Omega(\frac{N}{B} \frac{\log_B N}{\log_B \log_B N})$ space, $O(\frac{\log_B^2 N}{\log_B \log_B N})$ update
- O-tree: $\Omega(\sqrt{N/B} + T/B)$ query, $O(\frac{N}{B})$ space, $O(\log_B N)$ update

Techniques (one final time)

- Tools:
 - B-trees
 - Persistent B-trees
 - Buffer trees
 - Logarithmic method
 - Weight-balanced B-trees
 - Global rebuilding
- Techniques:
 - Bootstrapping
 - Filtering

Other results

- Many other results for e.g.
 - Higher dimensional range searching
 - Range counting
 - Halfspace (and other special cases) of range searching
 - Structures for moving objects
 - Proximity queries
- Many heuristic structures in database community
- Implementation efforts:
 - LEDA-SM (MPI)
 - TPIE (Duke)

External memory data structures

