Talk Outline

- Spatial join and issues
- Background: R-tree
- The seeded tree method
- Seeded tree construction
- Experiment results
- Conclusions

Objective

- Spatial joins with no pre-computed spatial indices.
- No existing solutions.
- Approach: construct spatial index structure at join time.

Previous Spatial Joins

- Cannot use relational join alg.
 - Spatial data lack total order.
 - Spatial joins are more than natural joins.
- Use spatial indices designed for spatial selections.
 - E.g. R-trees, R*-trees, ...
- Spatial indices must exist for datasets at time of join

[Beckmann et al. 90, Brinkhoff et al. 93, Faloutsos et al. 87, Gunther 93, Guttman 84, Sellis et al. 87].

Expensive to construct dynamically.

Spatial Join

• Spatial data:

Data with spatial extent.

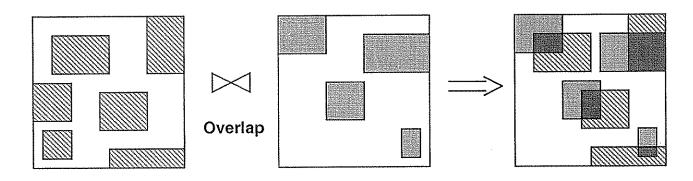
E.g.: points, lines, regions

• Spatial Join:

Given spatial data sets A and B, find all (a,b), $a \in A$ and $b \in B$, such that

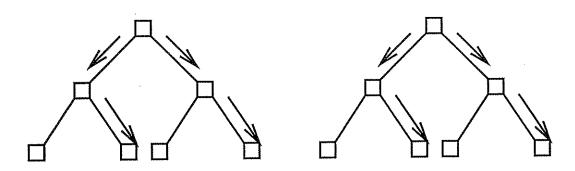
 $spatial_predicate(a, b) = TRUE$

• Commonly encountered predicate: overlap(a, b).



Background: Tree Matching

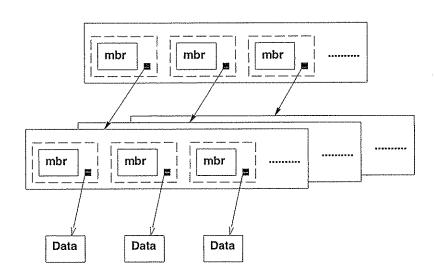
- R-tree join algorithm [Brinkhoff et al. 93]: Between two pre-computed R-trees
 - 1. Two nodes match iff their mbrs overlap.
 - 2. Recursively descend both trees finding pairs of matching nodes.
 - 3. Report results at the leaf level.



Background: R-Trees

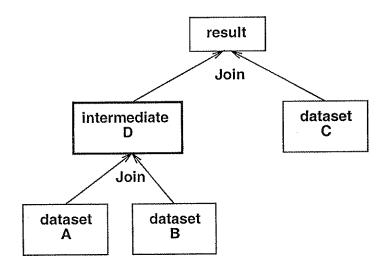
- B-tree-like data structure.
- Node contains array of (mbr, cp).

 - Minimal bounding box:
- Expensive to construct when large.
 - Possibility of memory thrashing.



Problem with Pre-computed Indices

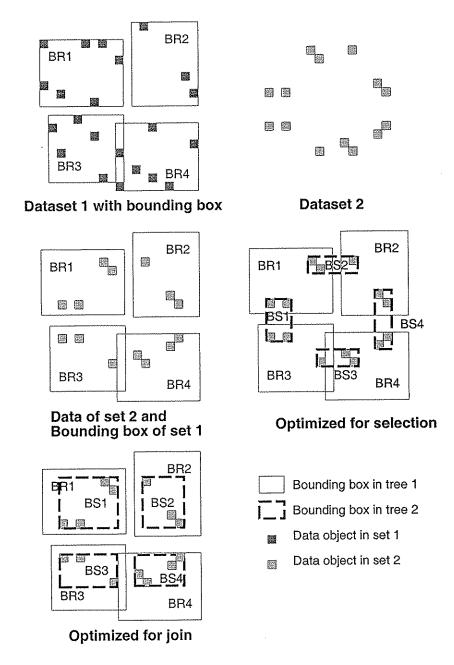
- Spatial indices not exist for all datasets.
- Queries with non-spatial selections.
 - E.g. Find all <u>government-owned</u>
 buildings that overlap residential areas.
- Queries with multiple spatial joins.
 - Input many be intermediate results.



Our Approach: Seeded Trees

- Dynamically build indices at join time.
- Principles:
 - 1. Index optimized for join, not selection.
 - 2. Exploit information about join.
 - 3. Low construction costs.
- Working assumption:
 - R-tree exists for one dataset.
 - Construct a seeded tree to join Rtree.

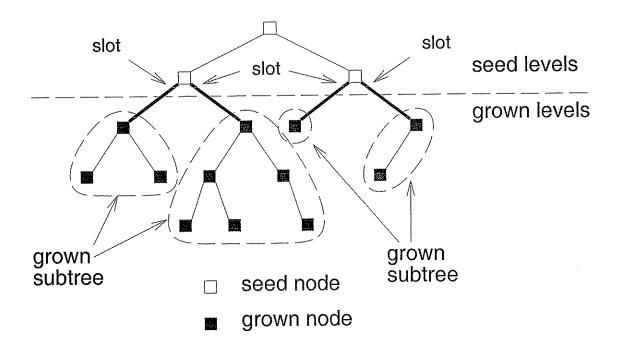
Joins vs. Selections



Seeded Tree Structure

- Tailored for join with a given R-tree.
- Upper levels: Seed levels.
- Grown level: grown subtrees are R-trees.

Slots

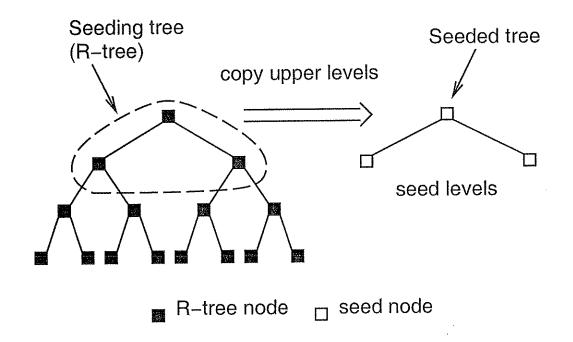


Seeded Tree Life Cycle

- Tree construction
 - Seeding phase
 - growing phase
 - cleanup phase
- Tree matching

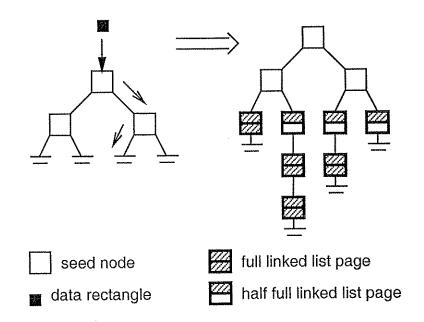
Seeding Phase

- \bullet Copy upper n levels of R-tree.
- Copied nodes may be transformed.



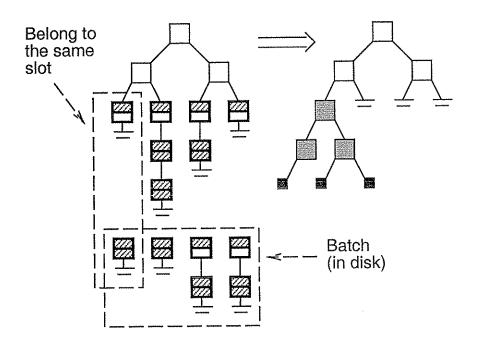
Growing Phase: Build Linked Lists

- Insert each object through seed levels, choosing appropriate slot.
- Build linked lists at the slots.
- When buffer full, batch-write linked lists to disk.



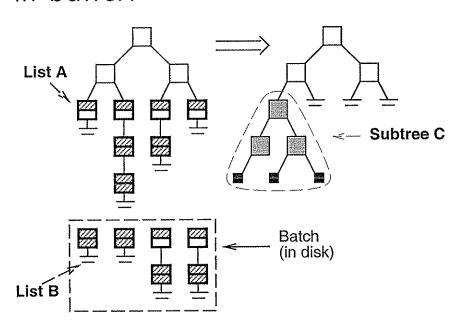
Growing Phase: Build Grown Subtrees

- Convert linked lists into grown subtrees.
- Construct one subtree at a time.
 - Avoids memory thrashing.
- Construct subtree under i using linked lists attached to i.



Tree Construction Buffer & I/O Management

- Building linked lists: write batches.
- building grown subtrees:
 - Read units: linked lists
 - Write units: subtrees
 - At most needs 1 linked list & 1 subtree in buffer.

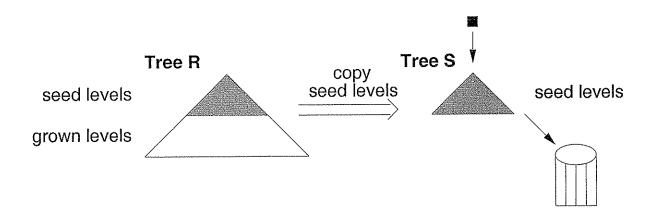


Cleanup Phase and Tree Matching

- Cleanup Phase: housing keeping.
 - Final adjustment of mbrs if necessary.
 - Delete empty slots.
- Tree matching: produce join result.

Seed-Level Filtering

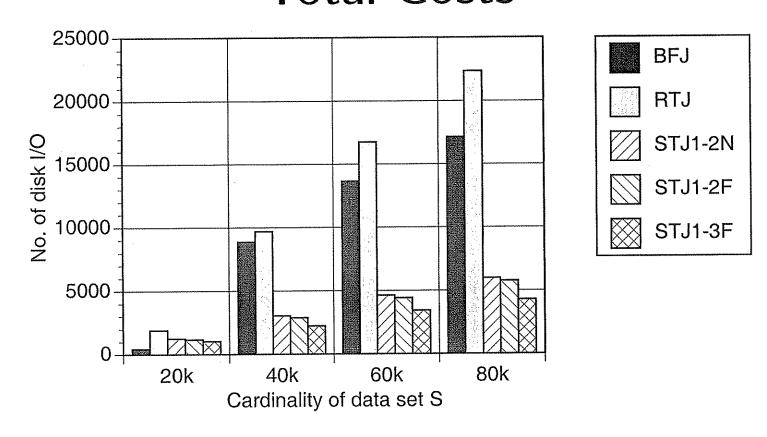
- ◆ Object overlaps some leaf of tree
 ⇒ overlaps some node in each level.
- With copy-seeding if object overlaps no seed level
 - It overlaps no leaf of the seeding tree.
 - Don't consider it anymore.
- Reduces seeded tree size.



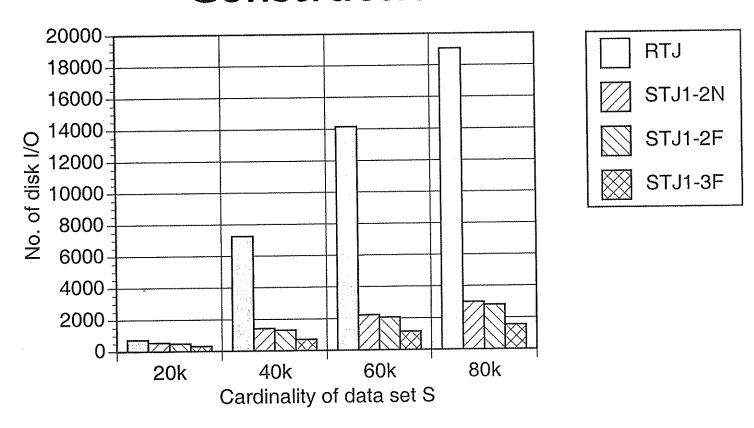
Experiments

- **STJ** (Seeded Tree Join): construct a seeded tree, and match with existing R-tree.
- RTJ (R-Tree Join): construct an R-tree, and match with existing R-tree. (variation of [Brinkhoff et al. 93])
- **BFJ** (Brute Force Join): perform a series of window queries (i.e. spatial selections).
- Experiment series 1:
 - vary data set size.
- Experiment series 2:
 - vary degree of spatial clustering.

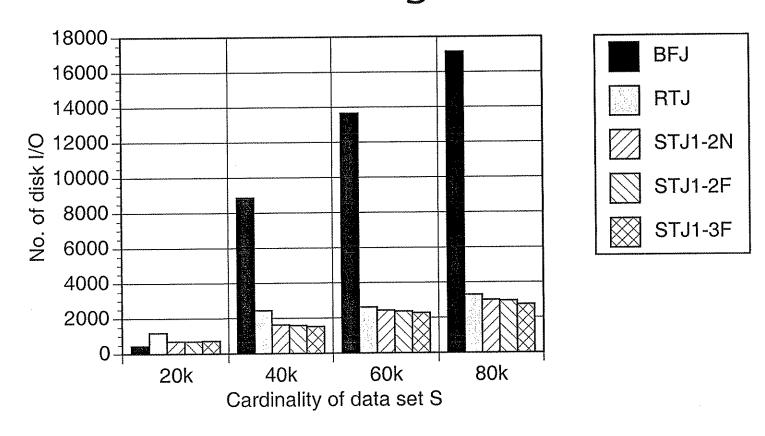
Experiment Series 1: Total Costs



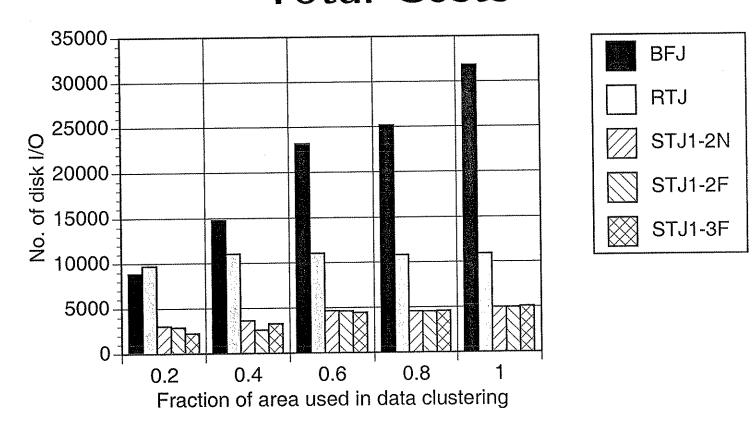
Experiment Series 1: Construction Cost



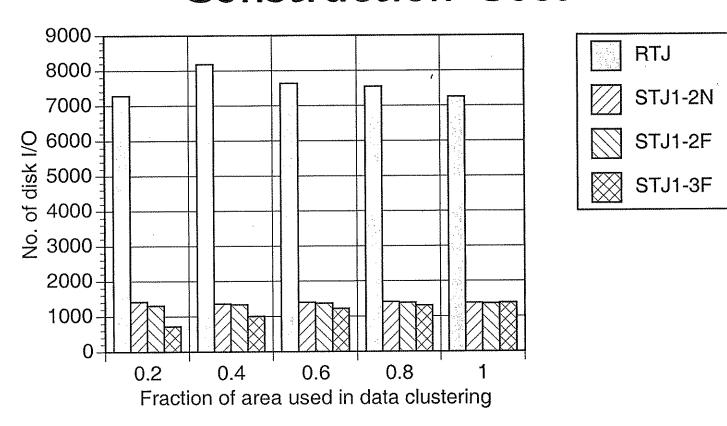
Experiment Series 1: Matching Cost



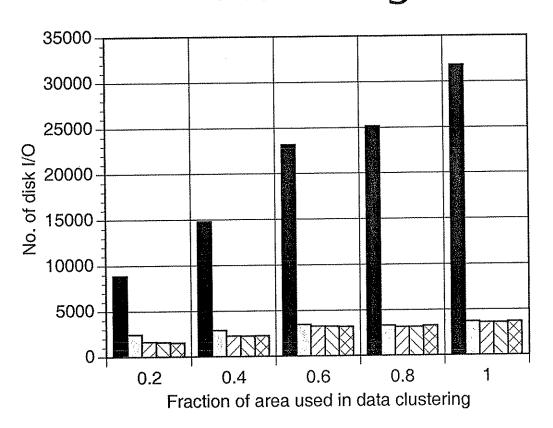
Experiment Series 2: Total Costs

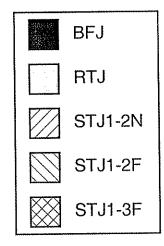


Experiment Series 2: Construction Cost



Experiment Series 2: Matching Cost





Conclusions

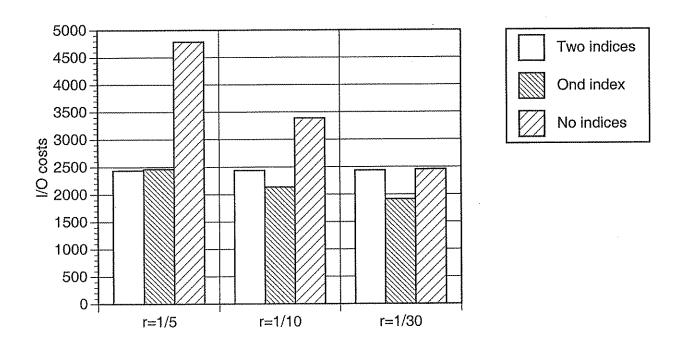
- New method to address situations where existing spatial indices are not applicable.
- Dynamic index construction at very low cost.
- Significant performance win over other methods.
- ▶ Dynamically constructing indices for joins
 ⇒ Doable for spatial databases.
- Extensions:
 - 2-seeded tree joins.
 - Reducing matching costs spatial hash joins.

New Problem 2-Seeded-Tree Join

- No spatial index for either dataset.
- Must dynamically construct 2 seeded trees.
- Difficulty:
 No R-tree to copy seed levels from.
- Solution: Don't copy!
 - Bootstrap seeding: determine topology and contents of seed levels from dataset.

Costs of 2-ST Joins

- Comparison <u>unfair</u> to seeded tree joins.
 - Two indices: given 2 pre-computed RT.
 - One index: given 1 pre-computed RT, build 1 ST.
 - No indices: build 2 ST.



Spatial Hash Join

- Relational hash join paradigm.
- Bootstrap Seeding to produce hash function.
- Solve multiple overlap problem.
- Good performance in our experiment.

Copying Strategies

 C_1 : Copy mbrs.

 C_2 : Copy the center points of mbrs.

 C_3 : Copy center points of mbrs at slot level. At other levels, mbr fields contain the true minimum bounding boxes of its children.

Update Strategies

 U_1 : No updates after insertions.

 U_2 : Update mbrs after each insertion to enclose inserted data objects and original seed mbrs.

 U_3 : Same as U_2 , but updated mbrs enclose only inserted data, not seed mbrs.

 U_4 : Update mbrs at slot level as in U_2 . Other mbrs untouched.

 U_5 : Update slot level mbrs as in U_3 . Other mbrs untouched.