第六章 空间查询处理与优化

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第六章 空间查询处理与优化

- 6.1 查询处理与优化
- 6.2 空间查询处理算法
- 6.3 查询优化
- 6.4 发展趋势 (自学)

References:

Spatial Databases: A Tour, Chapter 5

空间数据库管理系统概论,第七章

Analogy of Automatic Transmission in Cars

- Manual transmission : automatic :: Java : SQL
- Ex. List facilities within 10km of Minneapolies (44.978, -93.265)

```
Public class Facility {
Protected String name;
Protected Point location;
}
```

Java JDK 1.8



SQL (Oracle spatial)

```
Select f.name From Facility f
where SDO_WITHIN_DISTANCE (f.shape,
SDO_GEOMETRY(2001,4326,SDO_POINT_TYPE(44.978,-93.265,NULL)
NULL,NULL),distance=10') = 'TRUE'
```

SQL: Java:: Automatic Transmission: Manual

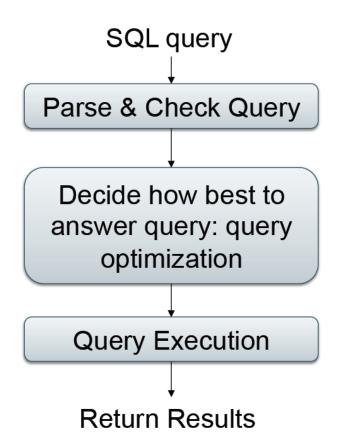
- SQL queries are declarative
 - Users do not specify algorithms and data-structures
 - Logical design and physical design are independent
 - No re-write needed for different users and data
 - DBMS needs to pick an algorithm to answer query
 - Analogy: automatic transmission choosing gear (1, 2, 3, ...)

Query Processing and Optimization (QPO)

- Basic idea of QPO
 - In SQL, queries are expressed in high level declarative form (关系代数)
 - QPO translates a SQL query to an execution plan (执行规划)
 - Over physical data model
 - Using operations on file-structures, indices, etc.
 - Goal: reduce run-time of execution plan
 - Answer query in as little time as possible
 - Constraints: QPO overheads are small
 - Computation time for QPO steps << that for execution plan</p>

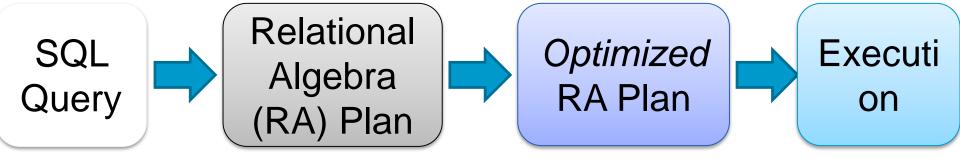
查询处理的步骤

- DBMS接收到SQL查询后,它的查询处理系统要将 查询转换为操作代码,一般分为四个步骤
 - _ 查询分析
 - _ 查询检查
 - _ 查询优化
 - _ 查询执行



查询处理的步骤

- SDBMS architecture
 - How does a SQL engine work?
 - Relational Algebra allows us to translate declarative (SQL) queries into precise and optimizable expressions



Declarative query (from user)

Translate to relational algebra expression

Find logically equivalent-but more efficient-RA expression

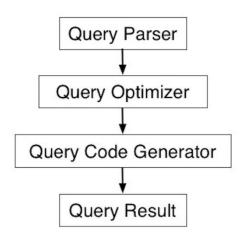
Execute each operator of the optimized plan

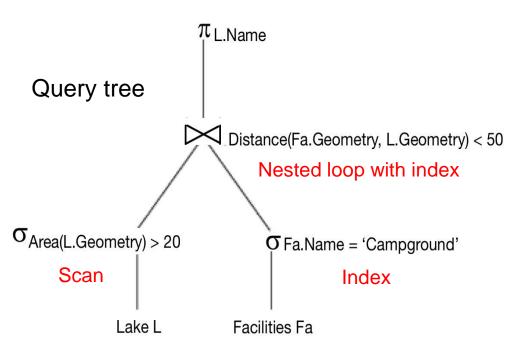
查询处理的步骤

- Query processing and optimization (QPO)
 - Picks algorithms to process a SQL query
- QPO: Physical data model :: automatic transmission : engine

```
Select L.Name
From Lake L, Facilities Fa

Where Area (L.Geometry) > 20 And
Fa.Name = 'campground' And
Distance(Fa.Geometry, L.Geometry < 50)
```





Why Learn About QPO?

- Why learn about automatic transmission in a car?
 - Identify cause of lack of power in a car
 - Is it the engine or the transmission?
 - Solve performance problem with manual override
 - Uphill, downhill driving => lower gears
- Why learn about QPO in a SDBMS?
 - Identify performance bottleneck for a query
 - Is it the physical data model or QPO?
 - How to help QPO speed up processing of a query?
 - Providing hints, rewriting query, etc.
 - How to enhance physical data model to speed up queries?
 - Add indices, change file structures, ...

Three Key Concepts in QPO

- 1. Building blocks
 - Most cars have few motions, e.g. forward, reverse
 - Similar most DBMS have few building blocks:
 - Select (point query, range query), join, sort, ...
 - A SQL query is decomposed in building blocks

Point Query	Scan all
	Ordered file structure
	Spatial index
	Scan all
Range Query	Ordered file structure
	Spatial index
Join	Nested loop
	Sort merge
	Hybrid join
	Hash join
	Star join
Spatial Join	Nested loop
	Nested loop with index
	Spatial partitioning-based
	Tree matching

Three Key Concepts in QPO

- 1. Building blocks
- 2. Query processing strategies for building blocks
 - Cars have a few gears for forward motion: 1st, 2nd, 3rd, overdrive
 - DBMS keeps a few processing strategies for each building block
 - e.g. a point query can be answer via an index or via scanning data file
- 3. Query optimization

Three Key Concepts in QPO

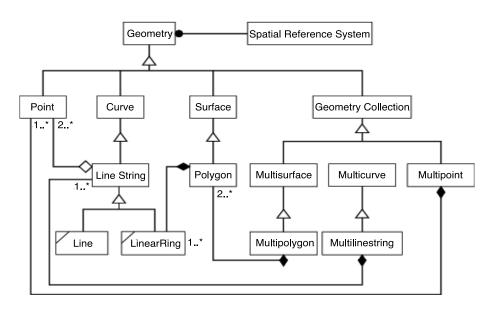
- 1. Building blocks
- 2. Query processing strategies for building blocks
- 3. Query optimization
 - Automatic transmission tries to picks best gear given motion parameters
 - For each building block of a given query, DBMS QPO tries to choose
 - "Most efficient" strategy given database parameters
 - Parameter examples: Table size, available indices, ...

QPO Challenges

- Choice of building blocks
 - SQL Queries are based on relational algebra (RA)
 - Building blocks of RA are select, project, join
 - SQL3 adds new building blocks like transitive closure
- Choice of processing strategies for building blocks
 - Constraints: Too many strategies => higher complexity
 - Commercial DBMS have a total of 10 to 30 strategies
 - 2 to 4 strategies for each building block
- How to choose the "best" strategy from among the applicable ones?
 - May use a fixed priority scheme
 - May use a simple cost model based on DBMS parameters

QPO Challenges in SDBMS

- Building Blocks for spatial queries
 - Rich set of spatial data types, operations
 - A consensus on "building blocks" is lacking
 - Current choices include spatial select, spatial join, nearest neighbor



Spatial select:

Return the boundary of Minnesota

Spatial join:

List the countries sharing boundary with Germany

Nearest neighbor:

Find the nearest library from a given apartment

QPO Challenges in SDBMS

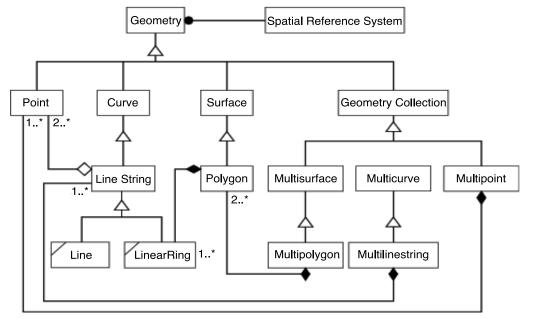
- Choice of strategies
 - Limited choice for some building blocks, e.g. nearest neighbor
- Choosing best strategies
 - Cost models are more complex since
 - Spatial Queries are both CPU and I/O intensive
 - While traditional queries are I/O intensive
 - Cost models of spatial strategies are in not mature

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Building Blocks for Spatial Queries

- Challenges in choosing building blocks
 - Rich set of data types point, linestring, polygon, ...
 - Rich set of operators topological, euclidean, set-based, ...
 - Large collection of computation geometric algorithms
 - Different spatial operations on different spatial data types
 - Desire to limit complexity of SDBMS



Basic Functions	SpatialReference ()
	Envelop ()
	Export ()
	IsEmpty ()
	IsSimple ()
	Boundary ()
Topological / Set	Equal
Operators	Disjoint
	Intersect
	Touch
	Cross
	Within
	Contains
	Overlap

Simplifying Choices for Building

- Reusing a Geographic Information System (GIS)
 - Which already implements spatial data types and operations
 - However may have difficulties processing large data set on disk
- SDBMS is used as a filter to reduce set of objects to a GIS
- This is filter and refinement approach

Simplifying Choices for Building

Filter and refinement approach



Which countries are crossed by Nile River?

Brute-force approach: Traverse all countries in the world and identify the crossed countries. Filter-and-refine approach

Filter:

Rule out all the countries outside Africa

Refine:

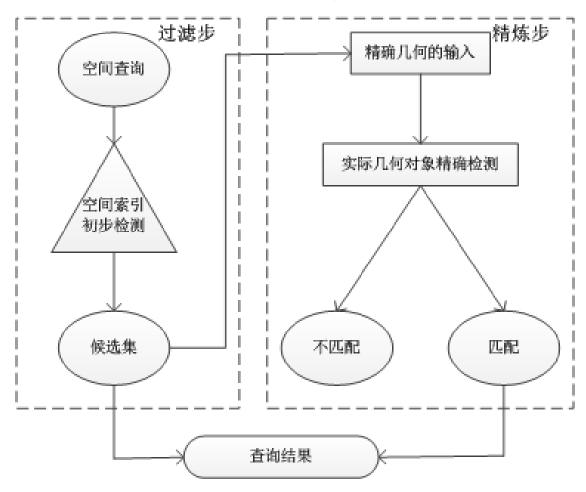
Traverse all countries in Africa and identify the crossed countries

空间查询

- 空间数据库中空间查询操作一般分为过滤和精炼两步
 - 一过滤步是利用空间对象索引信息以及空间对象的近似形状 ,检索出可能满足该空间查询条件的对象候选集
 - 精炼步是对候选集中的空间对象按查询要求进行精确的处理计算,以获得满足查询条件的最终结果
- 空间索引主要用于空间查询执行的过滤步

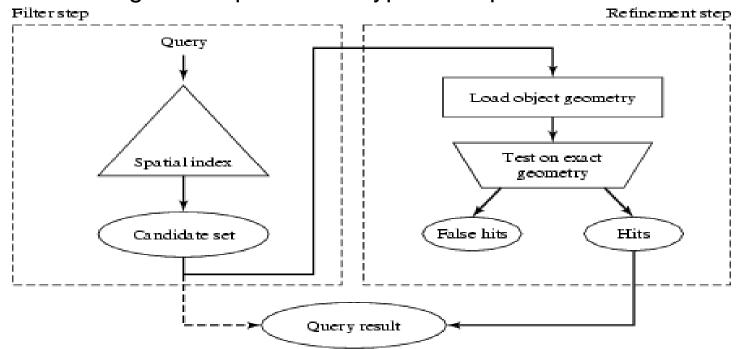
空间查询

• 空间查询一般分为过滤和精炼两步



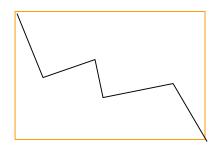
The Filter-Refine Paradigm

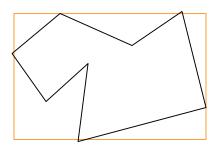
- Processing a spatial query Q
 - Filter step: find a superset S of object in answer to Q
 - Using approximate of spatial data type and operator
 - Refinement step: find exact answer to Q reusing a GIS to process S
 - Using exact spatial data type and operation



Approximate Spatial Data types

- Approximating spatial data types
 - Minimum orthogonal bounding rectangle (MOBR or MBR)
 - Approximates linestring, polygon, ...
 - See Examples below (red rectangle are MBRs for black objects)
 - MBRs are used by spatial indexes, e.g. R-tree
 - Algorithms for spatial operations MBRs are simple
- Question: Which OGIS operation returns MBRs?

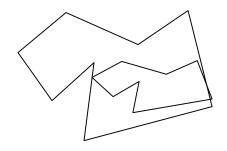


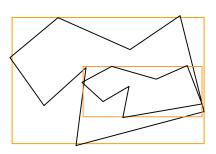


思考: Touches等拓扑操作的cost为100, MBRs获取的cost为1, MBRs之间拓扑操作的cost?

Approximate Spatial Operations

- Approximating spatial operations
 - SDBMS processes MBRs for refinement step
 - Overlap predicate used to approximate topological operations
 - Example: inside(A, B) replaced by
 - overlap(MBR(A), MBR(B)) in the filter step
 - Let A be outer polygon and B be the inner one
 - inside(A, B) is true only if overlap(MBR(A), MBR(B))
 - However overlap is only a filter for inside predicate needing refinement later

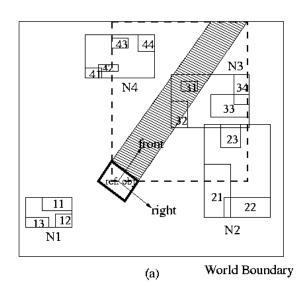


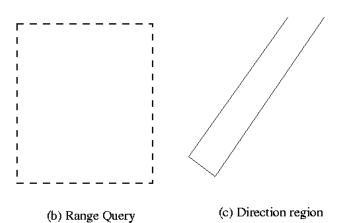




Filter Step Example 1

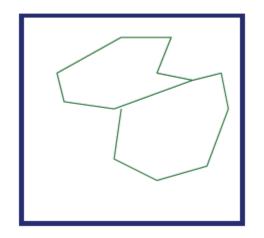
- Query:
 - List objects in front of a viewer V
- Equivalent overlap query
 - Direction region is a polygon
 - List objects overlapping with
 - polygon(front(V))
- Approximate query
 - List objects overlapping with
 - MBR(polygon (front (V)))

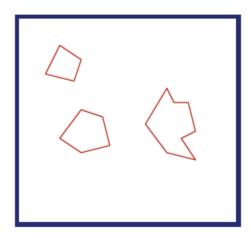


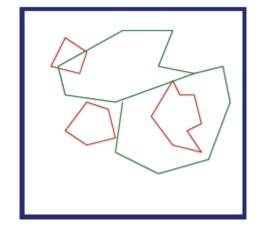


Filter Step Example 2

Spatial joins: find (quickly) all counties intersecting lakes

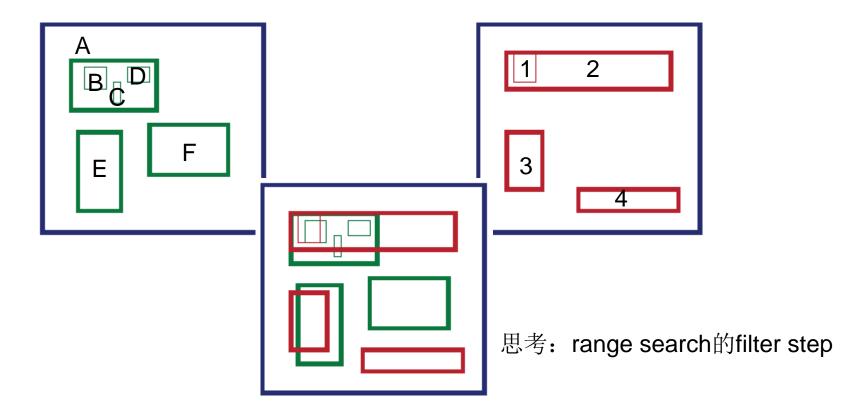






Filter Step Example 2

- Assume that they are both organized in R-trees:
 - 不利用R-trees: 6*4 = 24次复杂几何的空间关系判断
 - 利用R-trees进行filter: (A, 1), (A, 2), (B, 1), (B, 2), (C, 2),
 (D, 2), (E, 3), (F, 4)



Choice of Building Blocks

- Choice of building blocks
 - Varies across software vendors and products
 - Representative building blocks are listed here
- List of building blocks
 - Point Query Name a highlighted city on a digital map
 - Return one spatial object out of a table
 - Range Query List all countries crossed by of the river Amazon
 - Returns several objects within a spatial region from a table
 - Nearest Neighbor Find the city closest to Mount Everest
 - Return one spatial object from a collection
 - Spatial Join List all pairs of overlapping rivers and countries
 - Return pairs from 2 tables satisfying a spatial predicate

Choice of Building Blocks

Name the highlighted city



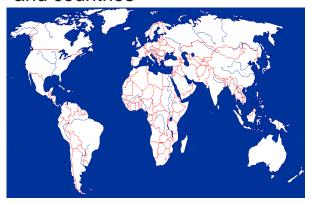
Find the city closest to Chicago



List countries crossed by Amazon River



List all pairs of overlapping rivers and countries

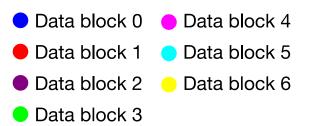


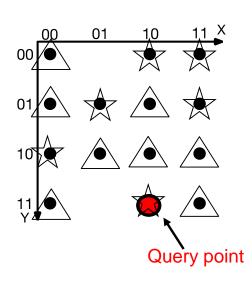
Strategies for Point Queries

- Recall Point Query Example
 - Given a location
 - Return a property (e.g., place name) of the location
- List of strategies
 - Scan all B disk sectors of the data file
 - If records are ordered using space filling curve (say Z-Curve)
 - Then use binary search on the Z-Curve of search point to examine about log(B, base = 2) disk sectors
 - If an index is available on spatial location of data objects,
 - Then use find() operation on the index
 - Number of disk sector examined = depth of index (typically 4 to 5)

Point Query Example

- Data: 14 points, each with a type triangle or star
- Query: Return type at location (x, y) = (10, 11)
- Candidate Storage Methods
 - 7 data blocks, each with 2 points

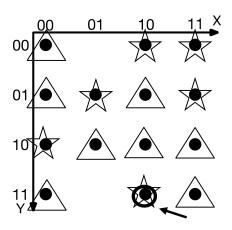




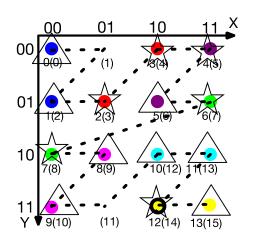
Candidate Storage & Indexing Methods

- Data block 0 Data block 4
 Data block 1 Data block 5
 Data block 2 Data block 6
- Data block 3

A. Unordered

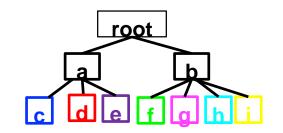


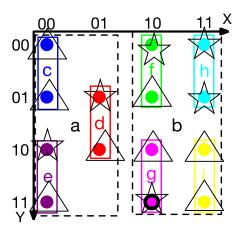
B. Z-order (Y-major)



Sorting number (Z-order index)

C. R-tree (primary index)

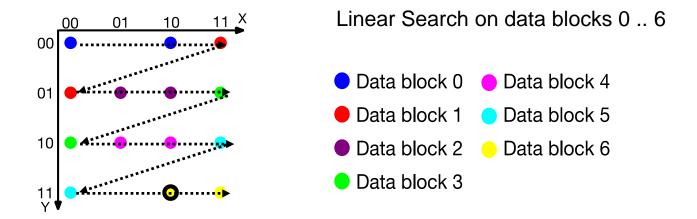




Linear Search for Point Query

- Data: 14 points, each with a type triangle or star
- (Marked) Query: Return the type of crime in the location (x, y) = (2, 3)
- Storage Methods: Unordered

Cost for linear search on this dataset: 7



Binary Search for Point Query

- Data: 14 points, each with a type triangle or star
- (Marked) Query: Return the type of crime in the location (x, y) = (2, 3)
- Storage Methods: Z-order (Y-major)

Cost for linear search on this dataset: 3

Y-major

00 01 10 11 X

00 00 01 10 11 X

00 00 01 10 11 X

00 01 10 11 X

00 Data block 0 Data block 4

01 Data block 1 Data block 5

10 2(3) 5(6) 6(7) Data block 2 Data block 6

10 7(8) 8(9) 10(12) 11(13)

Block序号和Z-Value 存在非线性关系,对 Z-value构建索引,如 B+树,需要访问哪些 Blocks?

Binary search on data blocks 0 .. 6

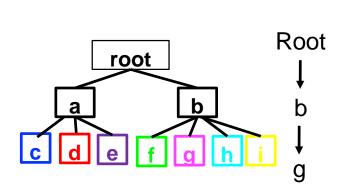
$$(0+6) / 2 = 3$$
3
Ceil((3+6) / 2) = 5
5
Ceil((5+6) / 2) = 6
6

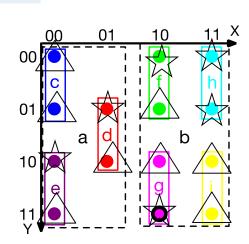
3 blocks (i.e., green, cyan, yellow) accessed

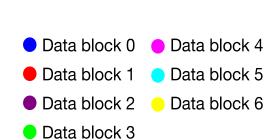
Search for Point Query Using R-Tree

- Data: 14 points, each with a type triangle or star
- (Marked) Query: Return the type of crime in the location (x, y) = (2, 3)
- Storage Methods: R-Tree (Primary Index), root cached in main meory

In this example		
Index block	1	
Data block	1	

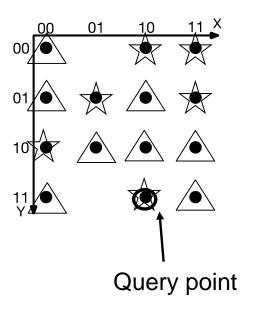






Comparing 3 Strategies for Point Query

- Data: 14 points, each with a type triangle or star
- (Marked) Query: Return the type of crime in the location (x, y) = (2, 3)



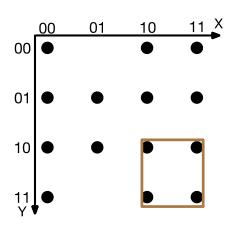
Storage Method	In this example	
Linear Search	7	
Binary Search	3	
Inday Coarab	Index blocks	1
Index Search	Data Blocks	1

Strategies for Range Queries

- Recall Range Query Example
 - List all countries crossed by of the river Amazon
 - Returns several objects within a spatial region from a table
- List of strategies
 - Scan all B disk sectors of the data file
 - If records are ordered using space filling curve (say Zorder)
 - Then determine range of Z-order values satisfying range query
 - Use binary search to get lowest Z-order within query answer
 - Scan forward in the data file till the highest z-order satisfying query
 - If an index is available on spatial location of data objects
 - then use range-query operation on the index

Range Query Example

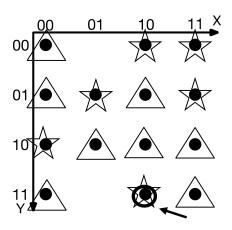
- Data: 14 points, each with a type triangle or star
- (Brown Box) Query: (2 <= x <= 3) and (2 <= y <= 3)
- Storage Methods:
 - 7 data block with 2 points each
 - Unordered, Z-ordered, R-tree



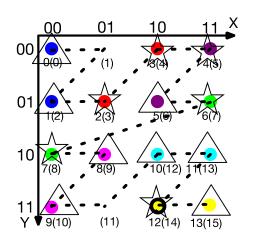
Candidate Storage & Indexing Methods

- Data block 0Data block 4Data block 1Data block 5Data block 6
- Data block 3

A. Unordered

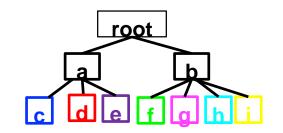


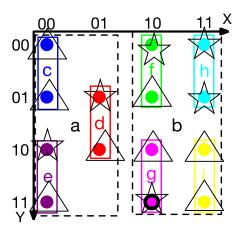
B. Z-order (Y-major)



Sorting number (Z-order index)

C. R-tree (primary index)

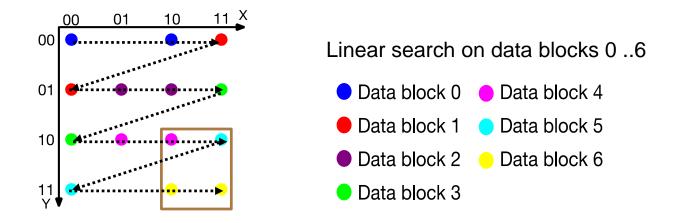




Linear Search for Range Query

- Data: 14 points, each with a type triangle or star
- (Brown Box) Query: (2 <= x <= 3) and (2 <= y <= 3)
- Storage Method: Unordered

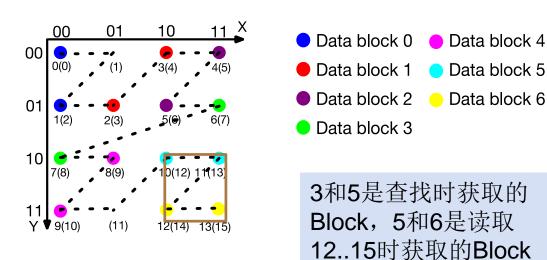
Cost for linear search on this dataset: 7



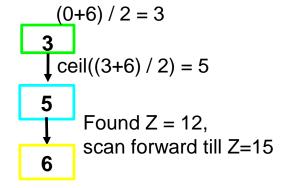
Binary Search for Range Query

- Data: 14 points, each with a type triangle or star
- (Brown Box) Query: (2 <= x <= 3) and (2 <= y <= 3)
- Storage Method: Z-order
 - One Z-interval 12..15 → search for 12 then scan forward

3 blocks (i.e., green, cyan, yellow) accessed

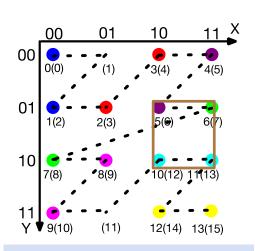


Binary search on data blocks 0..6



Range Query with Two Z-intervals

- Data: 14 points, each with a type triangle or star
- (Brown Box) Query: (2 <= x <= 3) and (1 <= y <= 2)
 - Two Z-intervals: [5..6] and [12..13]
 - One binary search (followed by scan) for each Z-interval



这里假设内存或 Cache比较大,如果 内存较小,Block 3可 能需要重复读取 Data block 0
Data block 4
Data block 1
Data block 5
Data block 2
Data block 6

Binary search to find [(6), (7)]

3 blocks (i.e., green, purple, cyan) accessed

Binary search to find [(12), (13)]

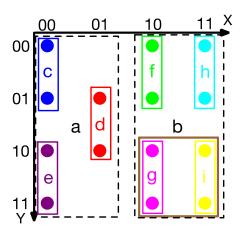
是否需要读Block 6?

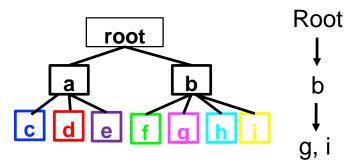
Search for Range Query Using R-Tree

- Data: 14 points, each with a type triangle or star
- (Brown Box) Query: (2 <= x <= 3) and (2 <= y <= 3)
- Storage Method: R-Tree (Primary Index)

Cost in this example			
Index block 1			
Data block 2			

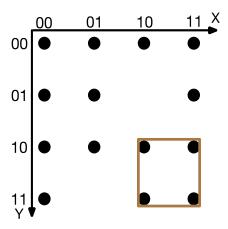






Comparing Algorithms for Range Query

- Data: 14 points, each with a type triangle or star
- (Brown Box) Query: (2 <= x <= 3) and (2 <= y <= 3)



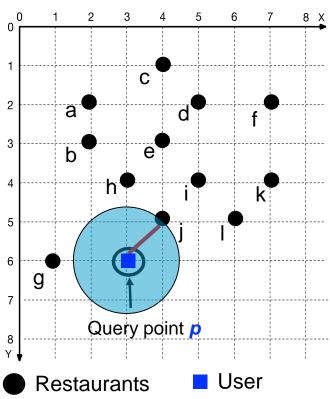
Storage Method	In this example	
Linear Search	7	
Binary Searches	3	
Inday Coarab	Index blocks	1
Index Search	Data Blocks	2

Strategies for Nearest Neighbor Queries

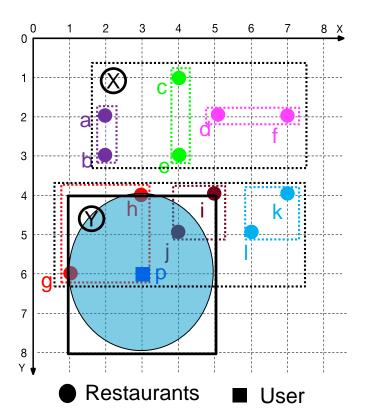
- Recall Nearest Neighbor Example
 - Find the city closest to Mount Everest
 - Return one spatial object from city data file C
- List of strategies
 - Two phase approach
 - Fetch C's disk sector(s) containing the location of Mt. Everest
 - M = minimum distance(Mt. Everest, cities in fetched sectors)
 - Test all cities within distance M of Mt. Everest (Range Query)
 - Single phase approach
 - Recursive algorithm for R-tree
 - Eliminate children dominated by some other children
 - Check the remaining data blocks for nearest neighbor

Nearest Neighbor Example

- Each point represents location of a restaurant
- Query: Given the location of a user p, find the nearest restaurant (if more than one nearest neighbors, return all results)
- Result
 - Nearest neighbor of p is j



Two Phase Strategy (with a R-Tree)



Find the index leaf containing the query point p: block red In red leaf, Point g, h are the closest points to p, $d_B = 2$

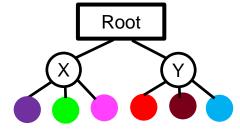
Create a circle Circle_p whose center is p, and radius = d_B

Create the MOBR of Circle_p: M_p

Range query: M_p, and test all points in M_p

Root -> Y -> Block brown

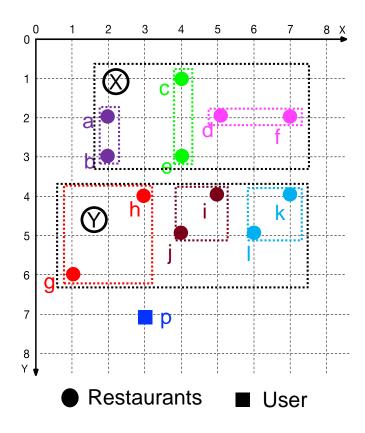
Since dist(p, j) = 1.41 < d_B , point j is nearest neighbor of p

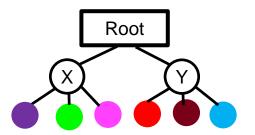


Cost:	Index blocks	Data blocks
Phase 1	1	1
Phase 2	1	1

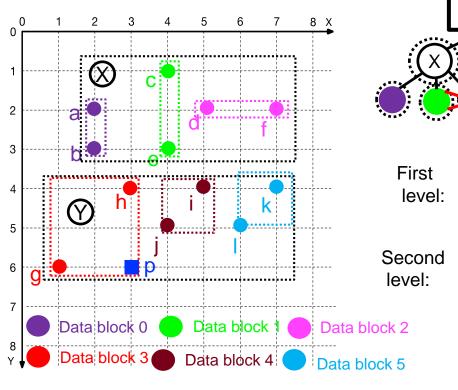
Two Phase Strategy (with a R-Tree)

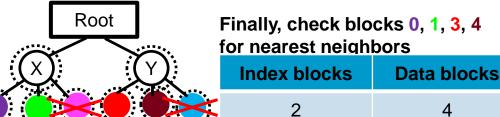
 Exercise: Generalize the algorithm to the case when query point is outside bounding box of root of the R-Tree?





One Phase Strategy (with a R-Tree)





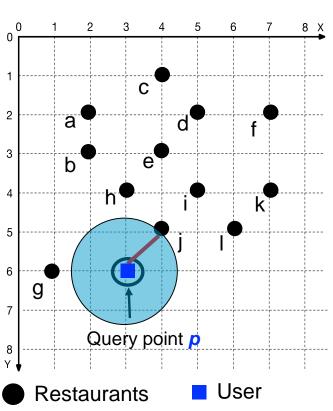
	MaxDist	MinDist	Node
	7.47	3	X
Nothing eliminated	4.47	0	Υ
	4.12	3.16	0
	5.10	3.16	1
Node 2 eliminated		4.47	2
	2.83	0	3
	2.83	1.41	4
Node 5 eliminated		3.16	5

4

Comparing Algorithms for Nearest Neighbor Queries

- Each point represents location of a restaurant
- Query: Given the location of a user p, find the nearest restaurant (if more than one nearest neighbors, return all results)
- Result
 - Nearest neighbor of p is j

Storage Method	In this example	
Two phose approach	Index blocks	2
Two phase approach	Data Blocks	2
One phase approach	Index blocks	2
One phase approach	Data Blocks	4



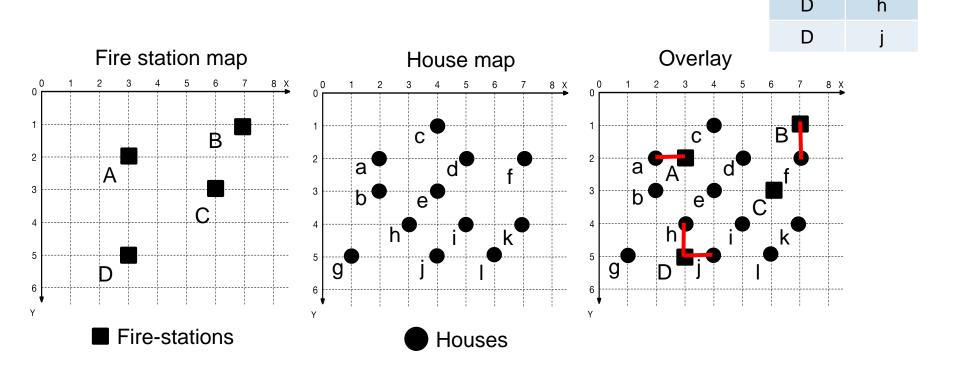
Strategies for Spatial Joins

- Recall Spatial Join Example
 - List all pairs of overlapping rivers and countries
 - Return pairs from 2 tables satisfying a spatial predicate
- List of strategies
 - Nested loop
 - Test all possible pairs for spatial predicate
 - All rivers are paired with all countries
 - Space Partitioning
 - Test pairs of objects from common spatial regions only
 - Rivers in Africa are tested with countries in Africa only
 - Tree Matching
 - Hierarchical pairing of object groups from each table
 - Other, e.g. spatial-join-index based, external planesweep, ...

Sptial Join Example

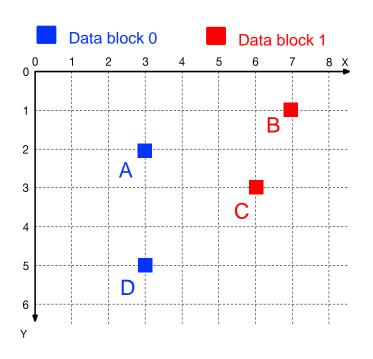
Query: For each fire station, find all the houses within a distance <= 1

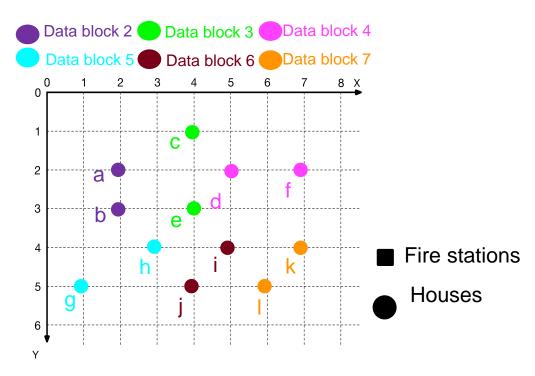
a



Storage Structure

- 2 blocks for fire stations
- 6 blocks for houses





Nested Loop

- Nested loop
 - Test all possible pairs for spatial predicate
 - Outer loop: bring data blocks of first table in memory
 - Inner loop: scan the second table

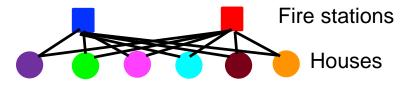
Data block 0 Data block 1

Algorithm: For each block B_{fs} of fire stations For each block B_h of houses

Data block 2 Data block 3 Data block 4

Data block 5 Data block 6 Data block 7

Scan all pairs of fire stations in B_{fs} and houses in B_h



Cost: For Blue block, inner loop fetches all 6 (circle) blocks For Red block, inner loop fetches all 6 (circle) blocks # blocks for fire stations * # blocks for houses = 2*6 = 12

总的Block = 2 + 12 Outer loop数据有关

Assume: 3 memory buffers (i.e., 1 for fire-stations, 1 for houses, 1 for results)

Nested Loop with Spatial Index

- Outer loop: For each data blocks D of first table
- Inner loop: Range Query second table for overlapping block

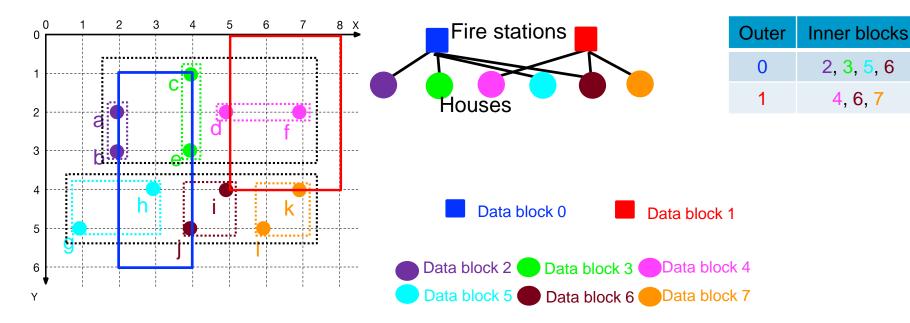
E.g., Houses within a distance <= 1</p> Root_h

Nested Loop with Spatial Index

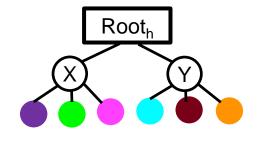
Outer loop: For each data blocks D of first table

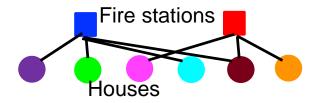
Root_h

- Inner loop: Range Query second table for overlapping block
 - E.g., Houses within a distance <= 1</p>



Nested Loop with Spatial Index





Index blocks	Data blocks		
2 + 2 = 4	FS	2	
	House	4+3=7	
	Total	9	

Block 0: Root -> X -> 2, 3 -> Y -> 5, 6

Block 1: Root -> X -> 4 -> Y -> 6, 7

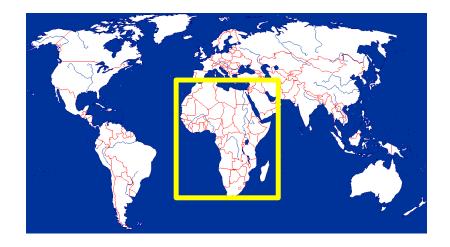


Data block 2 Data block 3 Data block 4

Data block 5 Data block 6 Data block 7

Space Partitioning Join

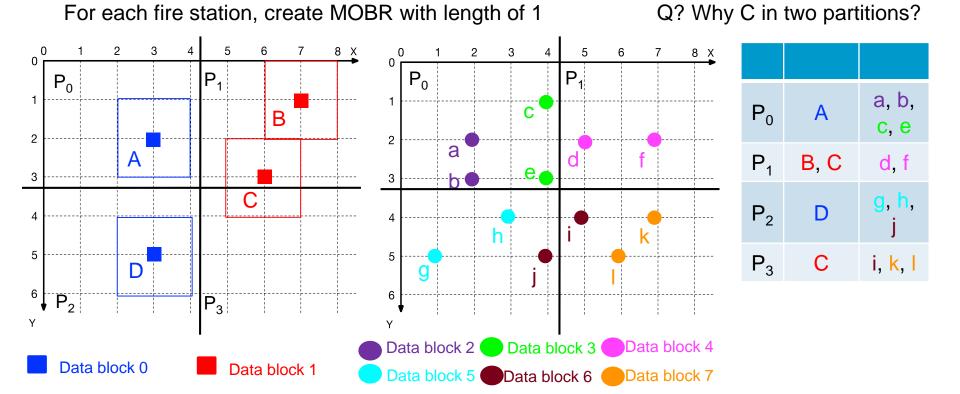
- Example Query: Pair rivers with countries they pass through.
 - Do we need to test Nile river with countries outside Africa?
- Space Partitioning Idea
 - Rivers in Africa are tested with countries in Africa only
 - Test pairs of objects within common spatial regions



Common Space Partitioning

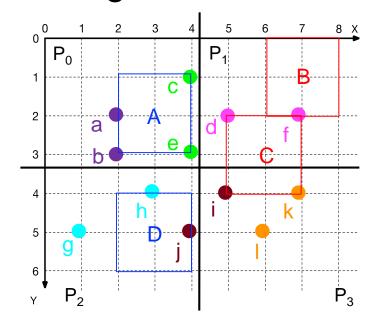
 Query: For each fire station, find houses within distance <= 1

Four Partition: P_0 , P_1 , P_2 , P_3



Space Partition Join Algorithm

- Filter: For each partition Pi
 - Bring Partition in main memory
 - Test all pairs of MOBR Mfs of fire-station in Pi and all houses in Pi
- Refine: Test remaining pair with exact geometry,
 e.g., distance <= 1



Partitions

P ₀	Α	a, b, c, e
P_1	B, C	d, f
P_2	D	g, h, j
P_3	С	i, k , l

Result after Filter Phase

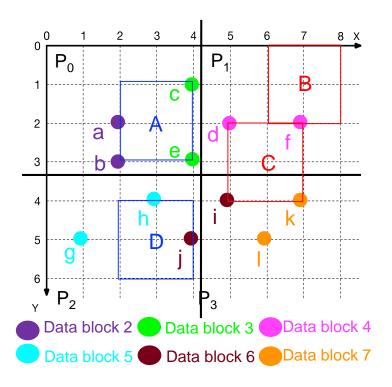
Result	MOBR	House
P_0	Α	a, b, c, e
·	В	f
P ₁	С	d, f
P_2	D	h, j
P_3	С	i, k

Space Partition Join Algorithm

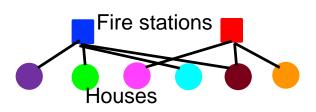
Total cost = 8+8+(3+2+3+3) = 27

Read all data blocks	Write partitioning back	Compute each part	
		P_0	3
0	0	P_1	2
8 8	P_2	3	
		P_3	3

About 3 "scans" of each table
If replication of objects across partitions is rare.



- Nested Loop with an Index
 - Inner loop range queries

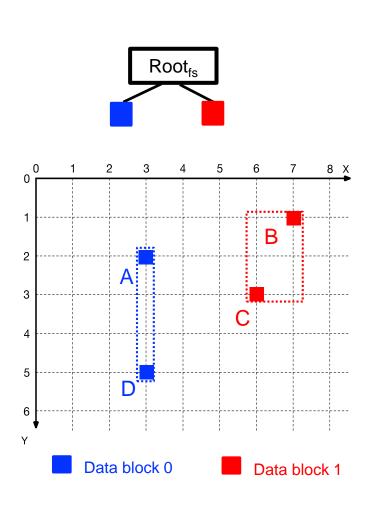


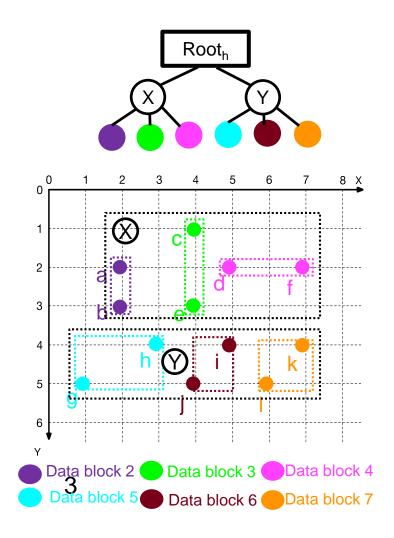
Eliminated pairs of data-blocks if disjoint MOBRs

- Space-partitioning join
 - Eliminated partition-pairs ((P0, P1), ...)
 since disjoint MOBRs

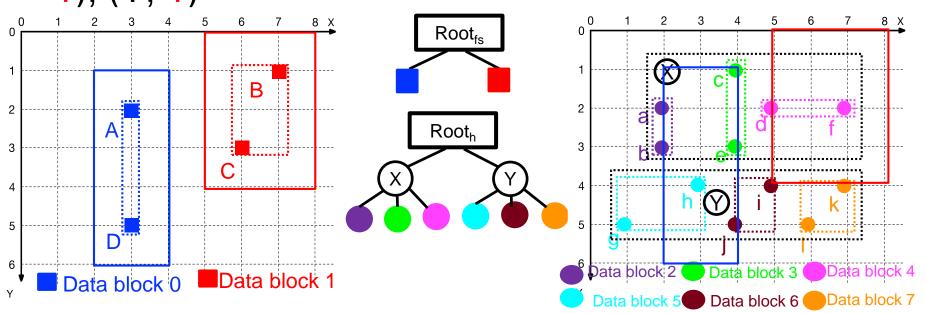
P_0	Α	a, b, c, e
P ₁	B, C	d, f
P_2	D	g, h, j
P_3	С	i, k, l

- Tree Matching, if both tables are indexed:
 - Eliminate pairs of index/data-blocks if disjoint MOBRs
 - Start at Root level Eliminate child-pair if irrelevant
 - Recursion on remaining pairs





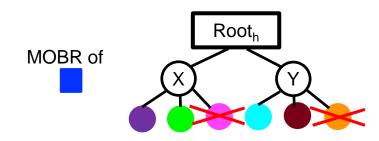
- For each fire station, find houses within distance <= 1
- MOBR buffer of size 1 to mimic spatial join predicate,
 i.e. distance <= 1
- Root level no child-pair is eliminated
- Recursion on remaining pairs, i.e., (X, 0), (Y, 0), (X, 1), (Y, 1)

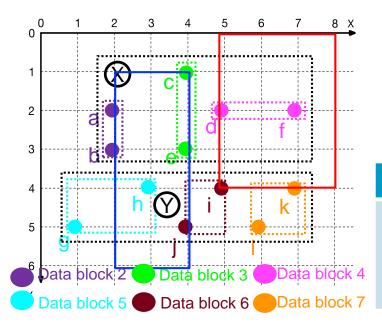


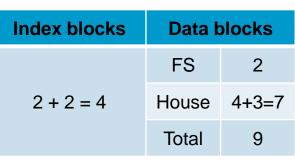
Next Iteration

- Data block 0 Data block 1
- Data block 2 Data block 3 Data block 4

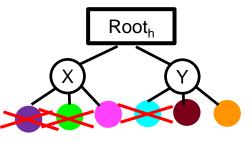
 Data block 5 Data block 6 Data block 7
- Recursion on (X, 0) => remaining pairs: (2, 0), (3, 0),
- Recursion on (Y, 0) = remaining pairs: (5, 0), (6, 0)
- Recursion on (X, 1) => remaining pairs: (4, 1),
- Recursion on (Y, 1) => remaining pairs: (6, 1), (7, 1)

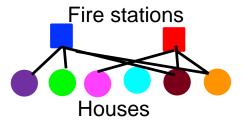






MOBR of



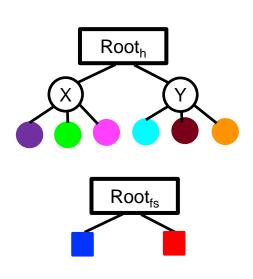


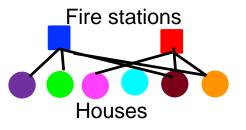
Cost



- Pairs examined:
 - (X, 0), (Y, 0), (X, 1), (Y, 1)
 - (2, 0), (3, 0), (5, 0), (6, 0), (4, 1), (6, 1), (7, 1)
- Blocks accessed
 - Index blocks besides roots: X, Y
 - Data blocks: all with 6 accessed twice

Index blocks	Data blocks	
	FS	2
2 + 2 = 4	House	4+3=7
	Total	9





Comparing Algorithms for Spatial Join

- Default choice is Nested loop
- Neither table has spatial index
 - Space partitioning if spatial-join predicate is selective

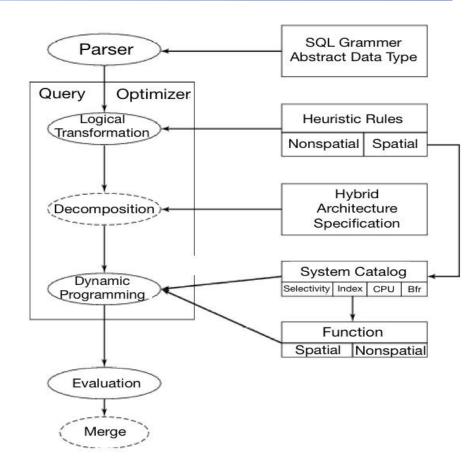
- One table has a spatial index
 - Nested loop with index
- Both table have spatial tree indexes & selective spatial join predicate
 - Tree matching

第六章 空间查询处理与优化

- 6.1 查询处理与优化
- 6.2 空间查询处理算法
- 6.3 查询优化
- 6.4 发展趋势 (自学)

Query Processing and Optimizer Process

- A site-seeing trip
 - Start : A SQL Query
 - End: An execution plan
 - Intermediate Stopovers
 - Query trees
 - Logical tree transforms
 - Strategy selection



- What happens after the journey?
 - Execution plan is executed
 - Query answer returned

Query Trees

- Nodes = Building blocks of (spatial) queries
- Children = Inputs to a building block
- Leafs = Tables
- Example SQL query and its query tree follows:

Select L.Name
From Lake L, Facilities Fa
Where ST_Area(L.Geometry) > 20 and
Fa.Name = 'compground' and
ST_Distance(Fa.Geometry, L.Geometr

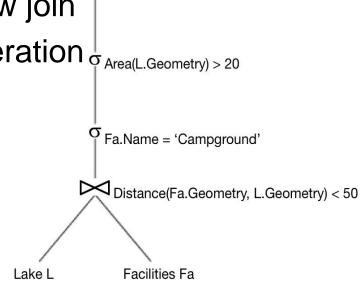
Distance(Fa.Geometry) < 50

 $\pi_{\text{L.Name}}$

Logical Transformation of Query Trees

- Motivation
 - Transformation do not change the answer of the query
 - But can reduce computational cost by
 - Reducing data produced by sub-queries
 - Reducing computation needs of parent node
- Example Transformation
 - Push down select operation below join
 - Reduces size of table for join operation | Area(L.Geometry) > 20
- Other common transformations
 - Push project down
 - Reorder join operations

_ ...



Logical Transformation and Spatial Queries

- Traditional logical transform rules
 - For relational queries with simple data types and operations
 - CPU costs are much smaller and I/O costs
 - Need to be reviewed for spatial queries
 - Complex data types, operations
 - CPU cost is hgher
- Example:
 - Push down spatial selection below join
 May not decrease cost if

 area() is costlier than distance()

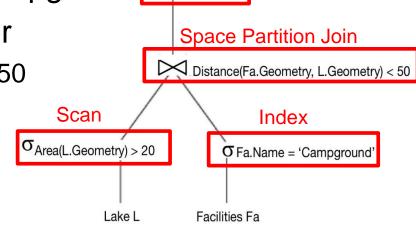
 \$\sigma_{\text{Area(L.Geometry)}} > 20\$
 \$\sigma_{\text{Fa.Name}} = 'Campground'}\$

Lake L

Facilities Fa

Execution Plans

- An execution plan has 3 components
 - A query tree
 - A strategy selected for each non-leaf node
 - An ordering of evaluation of non-leaf nodes
- Example
 - Strategies for Query tree
 - Use scan for Area(L.geom) > 20
 - Use index for Fa.Name = 'Campground' πL.Name
 - Use space-partitioning join for
 - Distance(Fa.geom, L.geom) < 50</p>
 - Use on-the-fly for projection
 - Ordering
 - As listed above



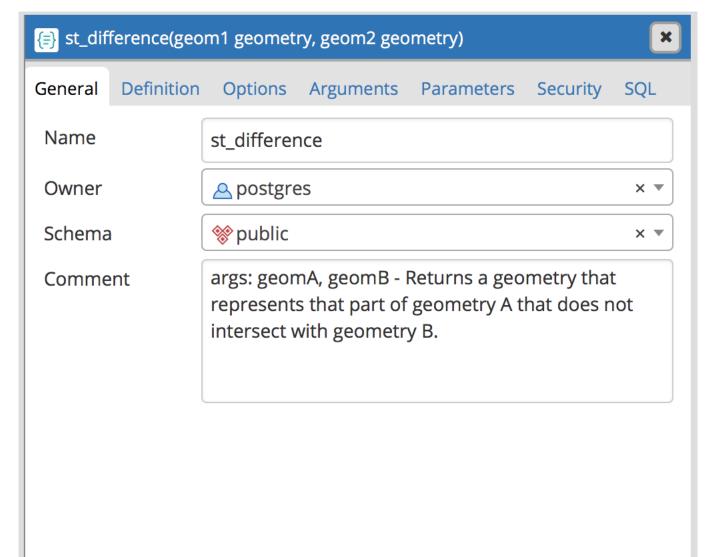
Choosing Strategies for Building Blocks

- A priority scheme
 - Check applicability of each strategies given filestructures and indices
 - Choose highest priority strategy
 - This procedure is fast, Used for complex queries
- Rule based approach
 - System has a set of rules mapping situations to strategy choices
 - Example: Use scan for range query if result size > 10 % of data file
- Cost based approach

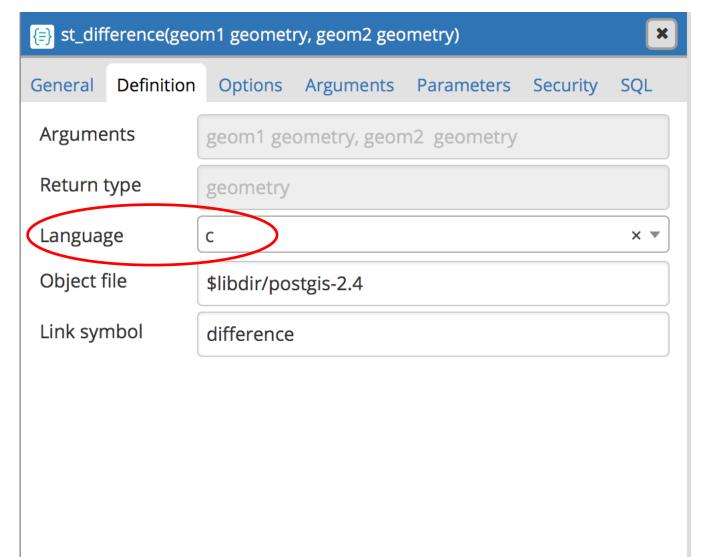
Choosing Strategies for Building Blocks

- Cost model based approach
 - Single building block
 - Use formulas to estimate cost of each strategy, given table size etc.
 - Choose the strategy with least cost
 - A query tree
 - Least cost combination of strategy choices for non-leaf nodes
 - Dynamic programming algorithm
- Commercial practice
 - RDBMS use cost based approach for relational building blocks
 - But cost models for spatial strategies are not mature
 - Rule based approach is often used for spatial strategies
 思考: PostgreSQL+PostGIS使用哪种策略?

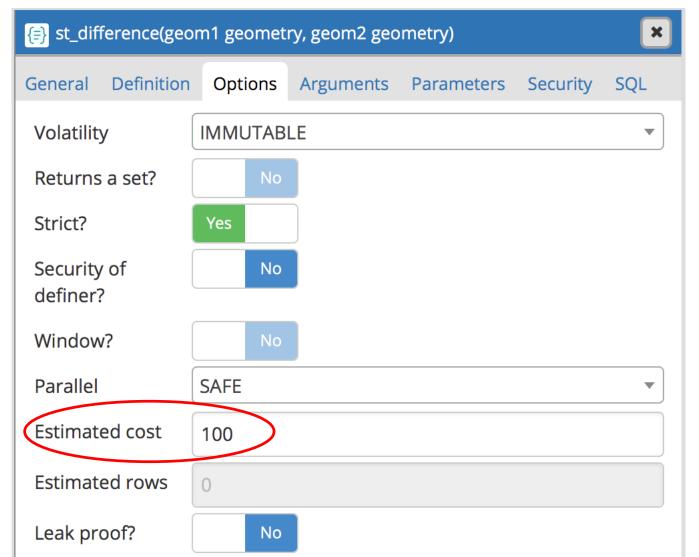
ST_Distance



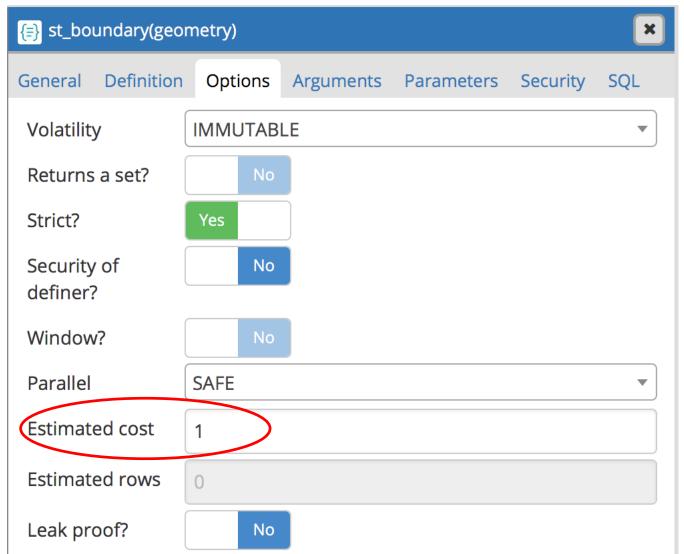
ST_Distance



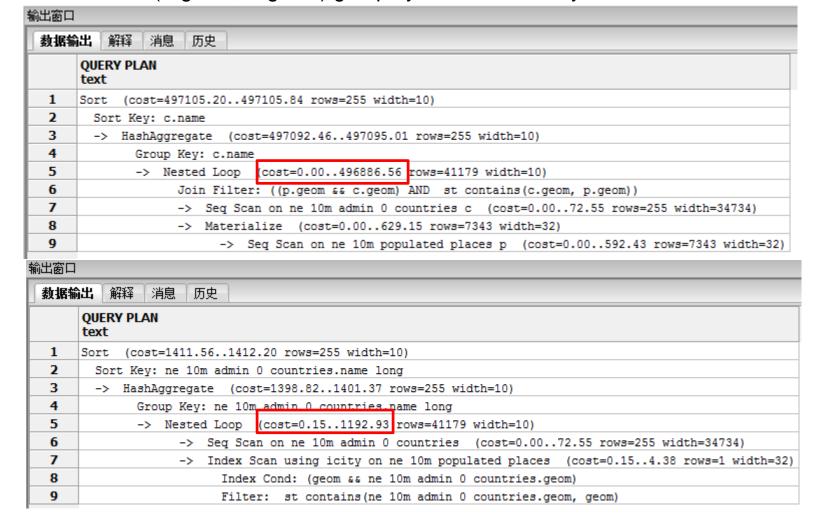
ST_Distance



ST_Boundary



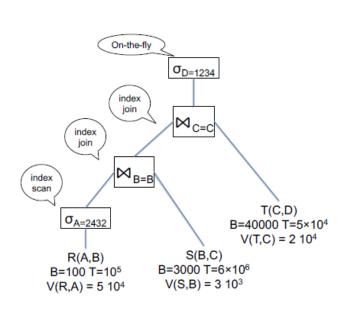
- explain select C.name, count(*)
- from ne_10m_admin_0_countries C, ne_10m_populated_places P
- where ST_Within(P.geom, C.geom) group by C.name order by C.name



- BigData '基于代价优化'究竟是怎么一回事?
 http://mp.weixin.qq.com/s/qP86_R6IPIOBr_TKi_Rk
 Ww
- 两个问题,SQL执行引擎如何知晓参与Join的两波数据集大小? 衡量两波数据集大小的是物理大小还是纪录多少抑或两者都有? 这关系到SQL解析器如何正确选择Join算法的问题。好了,这些就是这篇文章要为大家带来的议题一基于代价优化(Cost-Based Optimization,简称CBO)

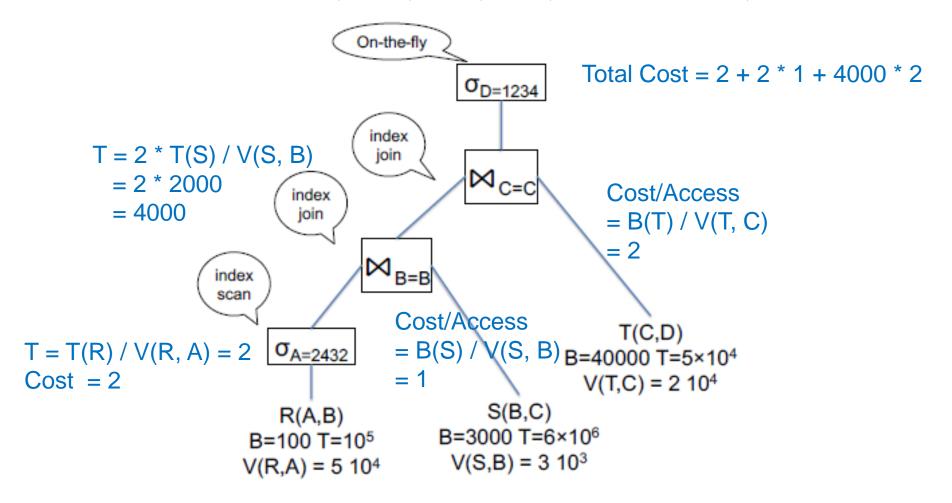
Cost Model Based Approach Example

- 关系R(A, B), S(B, C), T(C, D)统计信息如下
 - 数据量T(R) = 10⁵, T(S) = 6 * 10⁶, T(T) = 5 * 10⁴
 - 数据存储的Block数目B(R) = 100, B(S) = 3000, B(T) = 40000
 - 每个数值的取值数目 $V(R, A) = 5 * 10^4, V(R,B) = V(S, B)$ = 3 * 10³, $V(S, C) = V(T, C) = 2 * 10^4, V(T, D) = 10^4$
 - 非聚集索引R.A, R.B, S.C, T.D
 - 聚集索引S.B, T.C



Cost Model Based Approach Example

- 关系R(A, B), S(B, C), T(C, D)统计信息如下
 - 非聚集索引R.A, R.B, S.C, T.D, 聚集索引S.B, T.C

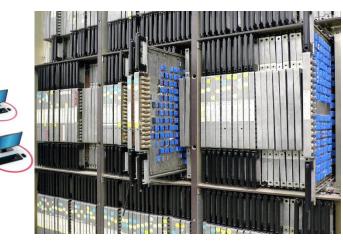


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Trends in Query Processing and Optimization

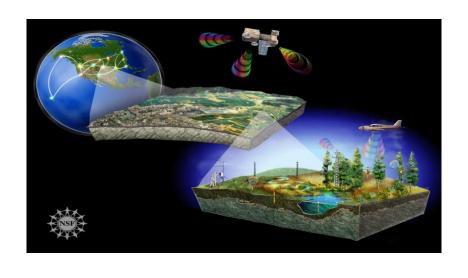
- Motivation
 - SDBMS and GIS are invaluable to many organizations
 - Price of success is to get new requests from customers
 - Support new computing hardware and environment
 - Support new applications
- New computing environments
 - Distributed computing
 - Internet and web
 - Parallel computers

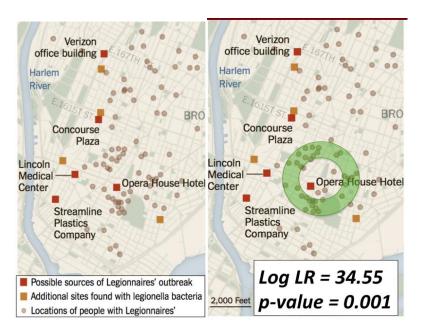


Trends in Query Processing and Optimization

- New applications
 - Location based services, transportation
 - Data Mining
 - Raster data

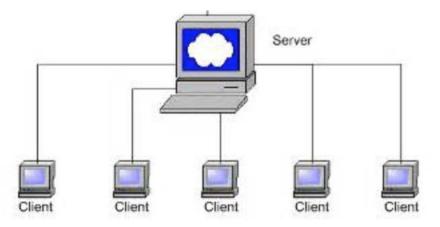






Distributed Spatial Databases

- Distributed Environments
 - Collection of autonomous heterogeneous computers
 - Connected by networks
 - Client-server architectures
 - Server computer provides well-defined services
 - Client computers use the services



Distributed Spatial Databases

- New issues for SDBMS
 - Conceptual data model
 - Translation between heterogeneous schemas
 - Logical data model
 - Naming and querying tables in other SDBMSs
 - Keeping copies of tables (in other SDBMs) consistent with original table
 - Query Processing and Optimization
 - Cost of data transfer over network may dominate CPU and I/O costs
 - New strategies to control data transfer costs

Internet and (World-wide-)web

- Internet and Web Environments
 - Very popular medium of information access in last few years
 - A distributed environment
 - Web servers, web clients
 - Common data formats (e.g. HTML, XML)
 - Common communication protocols (e.g. http)
 - Naming uniform resource locator (url), e.g. www.cs.umn.edu

Request

Response

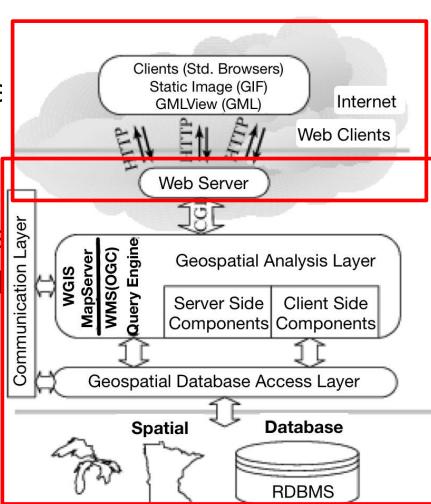
Internet

Response

- New issues for SDBMS
 - Offer SDBMS service on web
 - Use Web data formats, communication protocols etc.
 - Evaluate and improve web for SDBMS clients and servers

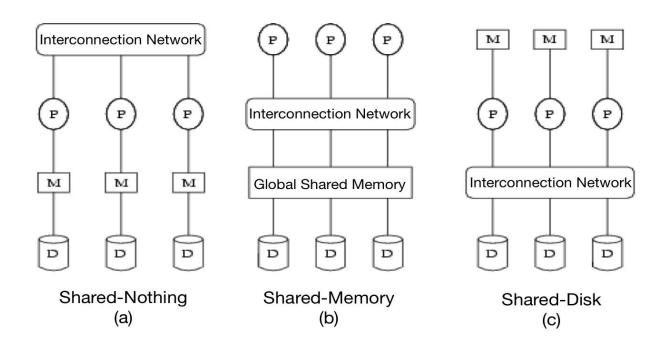
Web-based Spatial Database Systems

- SDBMS on web
 - MapServer case study
 - SDBMS talks to a web serve
 - Web server talks to web clie
- Commercial practice
 - Several web based products
 - Web data formats for spatial
 - GML
 - WMS



Parallel Spatial Databases

- Parallel Environments
 - Computer with multiple CPUs, Disk drives
 - All CPUs and disk available to a SDBMS
 - Can speed-up processing of spatial queries



Parallel Spatial Databases

- New issues for DBMS
 - Physical Data Model
 - Declustering: How to partition tables, indices across disk drives?
 - Query Processing and Optimization
 - Query partitioning: How to divide queries among CPUs?
 - Cost model of strategies on parallel computers
- Exmaple: Techniques for declustering
 - Simple technique: round robin based on an order (space filling curve)
 - Disk

Declustering for Data Partitioning

Exmaple

- A Simple Techniques for declustering
 - 1. Order the spatial objects using a space filling curve
 - 2. Allocate to disk drives in a round robin manner
- Effective for point objects, e.g. pixels in an image
- Many queries, e.g. large MBRs are parallelized well
 - Exercise: Consider a query to retrieve dat in bottom-left quarter of the space
 - Two data points retrieved fromeach disk drive for Z-curve

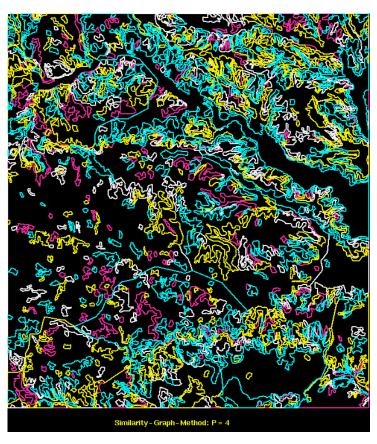
```
63 62 49 48 47 44 43 42
                                                                                        76107432
34567012
                         42 43 46 47 58 59 62 63
                                                  23672367
            70123456
                                                  01450145
                                                                60 61 50 51 46 45 40 41
                                                                                        45236501
67012345
            67012345
                         40 41 44 45 56 57 60 61
                                                                59 56 55 52 33 34 39 38
                                                  23672367
                                                                                        30741265
12345670
            56701234
                          34 35 38 39 50 51 54 55
                                                                58 57 54 53 32 35 36 37
                                                                                        21650312
                                                  01450145
            45670123
                          32 33 36 37 48 49 52 53
45670123
                                                                 5 6 9 10 31 28 27 26
                                                                                        56127432
                                                  23672367
70123456
            34567012
                          10 11 14 15 26 27 30 31
                                                                 4 7 8 11 30 29 24 25
                                                                                        47036501
                                                  01450145
            23456701
                          8 9 12 13 24 25 28 29
23456701
                                                                 3 2 13 12 17 18 23 22
                                                                                        32541276
                                                  23672367
56701234
            12345670
                          2 3 6 7 18 19 22 23
                                                                 0 1 14 15 16 19 20 21
                                                                                        01670345
                                                  01450145
                          0 1 4 5 16 17 21 21
01234567
            01234567
```

A Case Study: High Performance GIS

- Goal: Meet the response time constraint for real time battlefield terrain visualization in flight simulator
- Methodology:
 - Data-partitioning approach
 - Evaluation on parallel computers
 - e.g. Cray T3D, SGI Challenge
- Significance:
 - A major improvement in capability of geographic information systems for determining the subset of terrain polygons within the view point (Range Query) of a soldier in a flight simulator using real geographic terrain data set

A Case Study: High Performance GIS

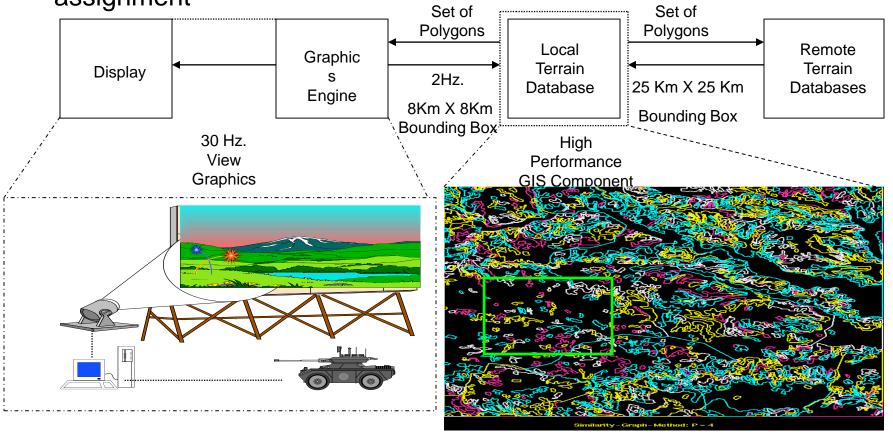
Dividing a Map among 4 processors. Polygons within a processor have common color



A Case Study: High Performance GIS

- (1/30) second Response time constraint on Range Query
- Parallel processing necessary since best sequential computer cannot meet requirement

Green rectangle = a range query, Polygon colors shows processor assignment



空间查询处理与优化总结

- Query processing and optimization (QPO)
 - Translates SQL Queries to execution plan
- QPO process steps include
 - Creation of a query tree for the SQL query
 - Choice of strategies to process each node in query tree
 - Ordering the nodes for execution
- Key ideas for SDBMS include
 - Filter-Refine paradigm to reduce complexity
 - New building blocks and strategies for spatial queries
 - CPU cost is higher
 - Push down spatial selection below join?
- 4类典型空间查询,3种执行规划选择策略