SPATIAL QUERY ON VANILLA DATABASES

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Spatial Query on Vanilla Databases

Spatial and GIS applications have traditionally required specialized databases, or at least specialized data structures like r-trees. Unfortunately this means that hybrid applications such as spatial analytics are not well served, and many people are unaware of the power of spatial queries because their favorite database does not support them.

In this talk, we describe how Apache Calcite enables efficient spatial queries using generic data structures such as HBase’s key-sorted tables, using techniques like Hilbert space-filling curves and materialized views. Calcite implements much of the OpenGIS function set and recognizes query patterns that can be rewritten to use particular spatial indexes. Calcite is bringing spatial query to the masses!
@julianhyde

SQL
Query planning
Query federation
BI & OLAP
Streaming
Hadoop

ASF member
Original author of Apache Calcite
PMC Apache Arrow, Calcite, Drill, Eagle, Kylin
Architect at Looker
Apache Calcite

Apache top-level project since 2015

Query planning framework used in many projects and products

Also works standalone: embedded federated query engine with SQL / JDBC front end

Apache community development model

https://calcite.apache.org
https://github.com/apache/calcite
Relational algebra

Based on set theory, plus operators: Project, Filter, Aggregate, Union, Join, Sort

Requires: declarative language (SQL), query planner

Original goal: data independence

Enables: query optimization, new algorithms and data structures
Algebraic rewrite

Optimize by applying rewrite rules that preserve semantics

Hopefully the result is less expensive; but it’s OK if it’s not (planner keeps “before” and “after”)

Planner uses dynamic programming, seeking the lowest total cost
A spatial query

Find all restaurants within 1.5 distance units of my location (6, 7)

<table>
<thead>
<tr>
<th>restaurant</th>
<th>x</th>
<th>y</th>
</tr>
</thead>
<tbody>
<tr>
<td>Zachary’s pizza</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>King Yen</td>
<td>7</td>
<td>7</td>
</tr>
<tr>
<td>Filippo’s</td>
<td>7</td>
<td>4</td>
</tr>
<tr>
<td>Station burger</td>
<td>5</td>
<td>6</td>
</tr>
</tbody>
</table>
A spatial query

Find all restaurants within 1.5 distance units of my location (6, 7)

Using OpenGIS SQL extensions:

```
SELECT * 
FROM Restaurants AS r
WHERE ST_Distance(
    ST_MakePoint(r.x, r.y),
    ST_MakePoint(6, 7)) < 1.5
```
Using ESRI’s geometry-api-java library, almost all ST_ functions were easy to implement in Calcite.

Slow – one row at a time.

```
package org.apache.calcite.runtime;

import com.esri.core.geometry.*;

/** Simple implementations of built-in geospatial functions. */
public class GeoFunctions {
    /** Returns the distance between g1 and g2. */
    public static double ST_Distance(Geom g1, Geom g2) {
        return GeometryEngine.distance(g1.g(), g2.g(), g1.sr());
    }

    /** Constructs a 2D point from coordinates. */
    public static Geom ST_MakePoint(double x, double y) {
        final Geometry g = new Point(x, y);
        return new SimpleGeom(g);
    }

    /** Geometry. It may or may not have a spatial reference associated with it. */
    public interface Geom {
        Geometry g();
        SpatialReference sr();
        Geom transform(int srid);
        Geom wrap(Geometry g);
    }

    static class SimpleGeom implements Geom { … }
}

SELECT *
FROM Restaurants AS r
WHERE ST_Distance(
    ST_MakePoint(r.x, r.y),
    ST_MakePoint(6, 7)) < 1.5
```
Traditional DB indexing techniques don’t work

**Sort**

```sql
CREATE /* b-tree */ INDEX I_Restaurants
ON Restaurants(x, y);
```

**Hash**

```sql
CREATE TABLE Restaurants(
    restaurant VARCHAR(20),
    x INTEGER,
    y INTEGER)
PARTITION BY (MOD(x + 5279 * y, 1024));
```

A scan over a two-dimensional index only has locality in one dimension
A “vanilla database”

Master

Region server [A - Gh]
Region server [Gi - Ts]
Region server [Tr - Z]
Spatial data structures and algorithms

The challenge: Reduce dimensionality while preserving locality

- **Reduce dimensionality** – We want to warp the information space so that we can access on one composite attribute rather than several
- **Preserve locality** – If two items are close in 2D, we want them to be close in the information space (and in the same cache line or disk block)

Two main approaches to spatial data structures:

- Data-oriented
- Space-oriented
R-tree (a data-oriented structure)
R-tree (split vertically into 2)
R-tree (split horizontally into 4)
R-tree (split vertically into 8)
R-tree (split horizontally into 16)
R-tree (split vertically into 32)
Grid (a space-oriented structure)
Grid (a space-oriented structure)
Spatial query

Find all restaurants within 1.5 distance units of where I am:

```
SELECT *
FROM Restaurants AS r
WHERE ST_Distance(
    ST_MakePoint(r.x, r.y),
    ST_MakePoint(6, 7)) < 1.5
```
Hilbert space-filling curve

- A space-filling curve invented by mathematician David Hilbert
- Every (x, y) point has a unique position on the curve
- Points near to each other typically have Hilbert indexes close together
Using Hilbert index

Add restriction based on h, a restaurant’s distance along the Hilbert curve

Must keep original restriction due to false positives

```
SELECT *
FROM Restaurants AS r
WHERE (r.h BETWEEN 35 AND 42
  OR r.h BETWEEN 46 AND 46)
AND ST_Distance(
  ST_MakePoint(r.x, r.y),
  ST_MakePoint(6, 7)) < 1.5
```
Telling the optimizer

1. Declare h as a generated column
2. Sort table by h

Planner can now convert spatial range queries into a range scan

Does not require specialized spatial index such as r-tree

Very efficient on a sorted table such as HBase

```
CREATE TABLE Restaurants (
  restaurant VARCHAR(20),
  x DOUBLE,
  y DOUBLE,
  h DOUBLE GENERATED ALWAYS AS ST_Hilbert(x, y) STORED)
SORT KEY (h);
```
Constraint: Table T has a column H such that: 
H = Hilbert(X, Y)
Variations on a theme

Several ways to say the same thing using OpenGIS functions:

- \( \text{ST\_Distance(ST\_Point(X, Y), ST\_Point(x, y))} < d \)
- \( \text{ST\_Distance(ST\_Point(x, y), ST\_Point(X, Y))} < d \)
- \( \text{ST\_DWithin(ST\_Point(x, y), ST\_Point(X, Y), d)} \)
- \( \text{ST\_Contains(ST\_Buffer(ST\_Point(x, y), d), ST\_Point(X, Y))} \)

Other patterns can use Hilbert functions:

- \( \text{ST\_DWithin(ST\_MakeLine(ST\_Point(x1, y1), ST\_Point(x2, y2)), ST\_Point(X, Y), d)} \)
- \( \text{ST\_Contains(ST\_PolyFromText('POLYGON((0 0,20 0,20 20,0 20,0 0))'), ST\_Point(X, Y), d)} \)
More spatial queries

What state am I in? (1-point-to-1-polygon)

Which states does Yellowstone NP intersect? (1-polygon-to-many-polygons)

Which US national park intersects with the most states? (many-polygons-to-many-polygons, followed by sort/limit)
More spatial queries

What state am I in? (point-to-polygon)

Which states does Yellowstone NP intersect? (polygon-to-polygon)

```sql
SELECT *
FROM States AS s
WHERE ST_Intersects(s.geometry, ST_MakePoint(6, 7))
```

```sql
SELECT *
FROM States AS s
WHERE ST_Intersects(s.geometry, ST_GeomFromText('LINESTRING(...)'))
```
Tile index

We cannot use space-filling curves, because each region (state or park) is a set of points and not known as planning time.

Divide regions into a (coarse) set of tiles. They intersect only if some of their tiles intersect.
## Tile index

<table>
<thead>
<tr>
<th>tileId</th>
<th>state</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>Nevada</td>
</tr>
<tr>
<td>0</td>
<td>Utah</td>
</tr>
<tr>
<td>1</td>
<td>Utah</td>
</tr>
<tr>
<td>2</td>
<td>Colorado</td>
</tr>
<tr>
<td>2</td>
<td>Utah</td>
</tr>
<tr>
<td>3</td>
<td>Idaho</td>
</tr>
<tr>
<td>3</td>
<td>Nevada</td>
</tr>
<tr>
<td>4</td>
<td>Idaho</td>
</tr>
<tr>
<td>4</td>
<td>Utah</td>
</tr>
<tr>
<td>4</td>
<td>Wyoming</td>
</tr>
<tr>
<td>5</td>
<td>Wyoming</td>
</tr>
<tr>
<td>6</td>
<td>Idaho</td>
</tr>
<tr>
<td>6</td>
<td>Montana</td>
</tr>
<tr>
<td>7</td>
<td>Montana</td>
</tr>
<tr>
<td>7</td>
<td>Wyoming</td>
</tr>
<tr>
<td>8</td>
<td>Montana</td>
</tr>
<tr>
<td>8</td>
<td>Wyoming</td>
</tr>
</tbody>
</table>
Aside: Materialized views

A materialized view is a table that is defined by a query. The planner knows about the mapping and can transparently rewrite queries to use it.

```
CREATE MATERIALIZED VIEW EmpSummary AS
SELECT deptno, COUNT(*) AS c
FROM Emp
GROUP BY deptno;
```
Building the tile index

Use the `ST_MakeGrid` function to decompose each state into a series of tiles.

Store the results in a table, sorted by tile id.

A materialized view is a table that remembers how it was computed, so the planner can rewrite queries to use it.

```
CREATE MATERIALIZED VIEW StateTiles AS
SELECT s.stateId, t.tileId
FROM States AS s,
     LATERAL TABLE(ST_MakeGrid(s.geometry, 4, 4)) AS t
```
Point-to-polygon query

What state am I in? (point-to-polygon)

1. Divide the plane into tiles, and pre-compute the state-tile intersections
2. Use this ‘tile index’ to narrow list of states

```
SELECT s.*
FROM States AS s
WHERE s.stateId IN (SELECT stateId
                     FROM StateTiles AS t
                     WHERE t.tileId = 8)
AND ST_Intersects(s.geometry, ST_MakePoint(6, 7))
```
Constraint #1: There is a table “Tiles” defined by
SELECT s.stateld, t.tileId FROM States AS s,
LATERAL TABLE(ST_MakeGrid(s.geometry, x, y)) AS t
Constraint #2: stateld is primary key of S

Venn diagram: Scan [S] → SemiJoin [S.stateld = T.stateld] → Filter [T.tileId = 8] → Filter [ST_Intersects(S.geometry, ST_Point(x, y))] → Scan [T]
Streaming + spatial

Example query: Every minute, emit the number of journeys that have intersected each city. (Some journeys intersect multiple cities.)

(Efficient implementation is left as an exercise to the reader. Probably involves splitting journeys into tiles, partitioning by tile hash-code, intersecting with cities in those tiles, then rolling up cities.)

```sql
SELECT STREAM c.name, COUNT(*)
FROM Journeys AS j
CROSS JOIN Cities AS c
  ON ST_Intersects(c.geometry, j.geometry)
GROUP BY c.name, FLOOR(j.rowtime TO HOUR)
```
Summary

Traditional DB techniques (sort, hash) don’t work for 2-dimensional data.

Spatial presents tough design choices:
- Space-oriented vs data-oriented algorithms
- General-purpose vs specialized data structures

Relational algebra unifies traditional and spatial:
- Use general-purpose structures
- Compose techniques (transactions, analytics, spatial, streaming)
- Must use space-oriented algorithms, because their dimensionality-reducing mapping is known at planning time
Thank you! Questions?

@ApacheCalcite | @julianhyde | https://calcite.apache.org

Resources & credits

- [CALCITE-1616] Data profiler
- [CALCITE-1870] Lattice suggester
- [CALCITE-1861] Spatial indexes
- [CALCITE-1968] OpenGIS
- [CALCITE-1991] Generated columns
- Talk: “Data profiling with Apache Calcite” (Hadoop Summit, 2017)
- Talk: “SQL on everything, in memory” (Strata, 2014)
- Zhang, Qi, Stradling, Huang (2014). “Towards a Painless Index for Spatial Objects”
Extra slides
Architecture

Conventional database

Calcite
select p.productName, count(*) as c
from splunk.splunk as s
join mysql.products as p
  on s.productId = p.productId
where s.action = 'purchase'
group by p.productName
order by c desc
select p.productName, count(*) as c
from splunk.splunk as s
join mysql.products as p
on s.productId = p.productId
where s.action = 'purchase'
group by p.productName
order by c desc
## Calcite framework

### Relational algebra
- **RelNode (operator)**
  - TableScan
  - Filter
  - Project
  - Union
  - Aggregate
  - ...
- **RelDataType (type)**
- **RexNode (expression)**
- **RelTrait (physical property)**
  - RelConvention (calling-convention)
  - RelCollation (sortedness)
  - RelDistribution (partitioning)
- **RelBuilder**

### SQL parser
- **SqlNode**
- **SqlParser**
- **SqlValidator**

### Transformation rules
- **RelOptRule**
  - FilterMergeRule
  - AggregateUnionTransposeRule
  - 100+ more
- **Global transformations**
  - Unification (materialized view)
  - Column trimming
  - De-correlation

### Cost, statistics
- **RelOptCost**
- **RelOptCostFactory**
- **RelMetadataProvider**
  - RelMdColumnUniqueness
  - RelMdDistinctRowCount
  - RelMdSelectivity

### Metadata
- **Schema**
- **Table**
- **Function**
  - TableFunction
  - TableMacro
- **Lattice**

### JDBC driver
- **RelBuilder**