Advanced SQL

06 — Recursion

Summer 2020

Torsten Grust Universität Tübingen, Germany SQL has grown to be an **expressive data-oriented language.** Intentionally, it has *not* been designed as a general-purpose programming language:

- 1. SQL does not loop forever: Any SQL query is expected to terminate, regardless of the size/contents of the input tables.
- 2. SQL can be evaluated efficiently:
 A SQL query over table T of c columns and r rows can be evaluated in $O(r^c)$ space and time.

¹ SQL cannot compute the set of all subsets of rows in T which requires $O(2^r)$ space, for example.

A Giant Step for SQL

The addition of recursion to SQL changes everything:

Expressiveness SQL becomes a Turing-complete language and thus a general-purpose PL (albeit with a particular flavor).

Efficiency

No longer are queries guaranteed to terminate or to be evaluated with polynomial effort.

Like a pact with the 🔀 — but the payoff is magnificient...

Recursion in SQL: WITH RECURSIVE

Recursive common table expresssion (CTE):

- In particular, any q_j may refer to itself (\circ)! Mutual references are OK, too. (Think letrec in FP.)
- \bullet Typically, final query q performs post-processing only.

Shape of a Self-Referential Query

```
WITH RECURSIVE \langle T \rangle (\langle c_1 \rangle, ..., \langle c_k \rangle) AS ( -- common schema of q_0 and q_0(\cdot) -- base case query, evaluated once UNION [ ALL ] -- either UNION or UNION ALL \langle q_0(T) \rangle -- recursive query refers to T -- itself, evaluated repeatedly \langle q_1(T) \rangle -- final post-processing query
```

• Semantics in a nutshell:

```
q(q\theta(\cdots q\theta(q\theta(q_0))\cdots) \cup \cdots \cup q\theta(q\theta(q_0)) \cup q\theta(q_0) \cup q_0) repeated evaluation of q\theta (when to stop?)
```

Semantics of a Self-Referential Query (UNION Variant)

Iterative and recursive semantics—both are equivalent:

```
iterate(q\theta, q_0):
r \leftarrow q_0
t \leftarrow r
while t \neq \phi
t \leftarrow q\theta(t) \setminus r
return r
t \leftarrow q\theta(t) \setminus r
return r
return r
return r
return r
```

- Invoke the recursive variant with recurse $(q\theta, q_0)$.
- ⊌ denotes disjoint set union, \ denotes set difference.
- $q\theta(\cdot)$ is evaluated over the new rows found in the last iteration/recursive call. Exit if there were no new rows.

A Home-Made generate_series()

Generate a single-column table series of integers $i \in \{from, from+1, ..., to\}$:

```
WITH RECURSIVE

series(i) AS (

A VALUES (<from>) -- q₀

UNION

SELECT s.i + 1 AS i -- 
FROM → series AS s
WHERE s.i < <to> -- } q⊕(series)

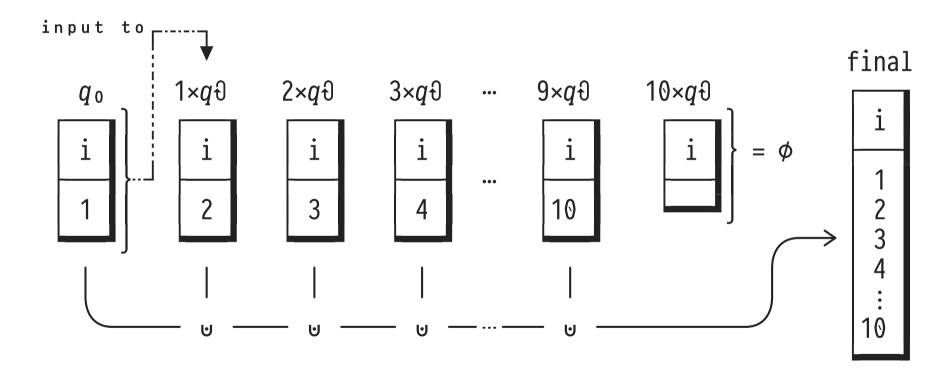
TABLE series
```

• Q: Given the predicate s.i < <to>, will to indeed be in the final table?

A Home-Made generate_series()

• Assume from = 1, to = 10:

New rows in table series after evaluation of...



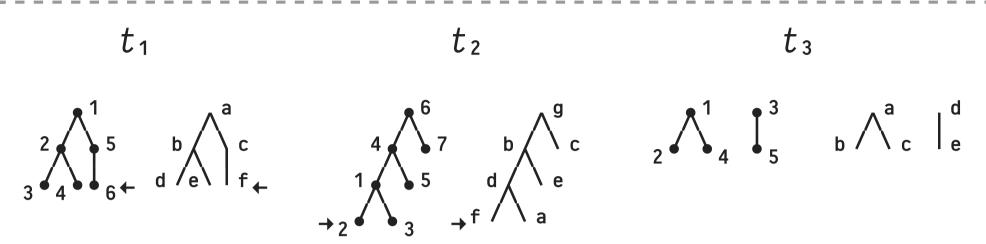
Semantics of a Self-Referential Query (UNION ALL Variant)

With UNION ALL, recursive query $q \cdot 0$ sees **all rows added in** the last iteration/recursive call:

```
iterate<sup>all</sup>(q\theta, q_0):
r \leftarrow q_0
t \leftarrow r
while t \neq \phi
t \leftarrow q\theta(t)
return\ r
t \leftarrow r \ t
return r
```

- Invoke the recursive variant via recurse^{all} $(q\theta, q_0)$.
- t denotes bag (multiset) union.
- Note: Could immediately emit t—no need to build r. \triangle

1 Traverse the Paths from Nodes 'f' to their Root



Array-based tree encoding (parent of node n = parents[n]):

<u>tree</u>	parents (□ ≡ NULL)	labels				
t_1	$\{\Box, 1, 2, 2, 1, 5\}$	{'a','b','d','e','c','f'}				
t_2		{'d','f','a','b','e','g','c'}				
t_3		{'a','b','d','c','e'}				
	1 2 3 4 5 6 7	1 2 3 4 5 6 7 ← node				
Troop						

Irees

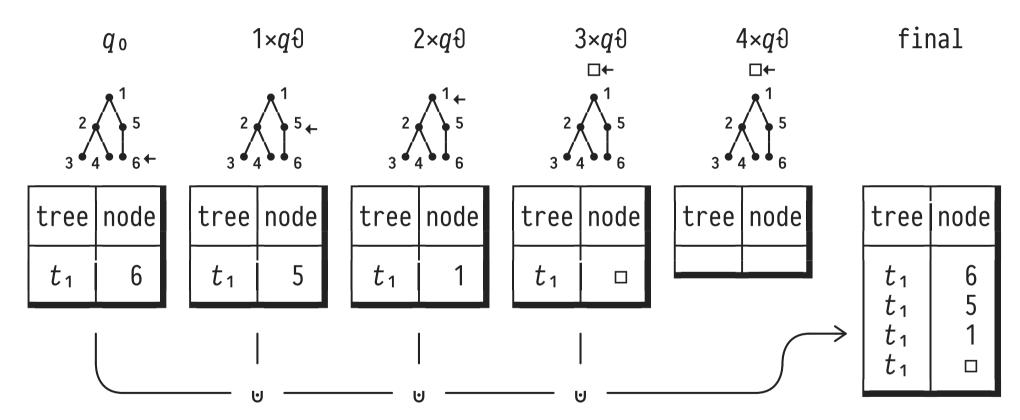
Traverse the Paths from Nodes 'f' to their Root

```
WITH RECURSIVE
 paths(tree, node) AS (
   SELECT t.tree, array_position(t.labels, 'f') AS node
    FROM Trees AS t
     UNION
   SELECT t.tree, t.parents[p.node] AS node
   FROM paths AS p,
          Trees AS t
   WHERE p.tree = t.tree
TABLE paths
```

 $(t,n) \in \text{paths} \iff \text{node } n \text{ lies on path from } 'f' \text{ to } t' \text{s root}$

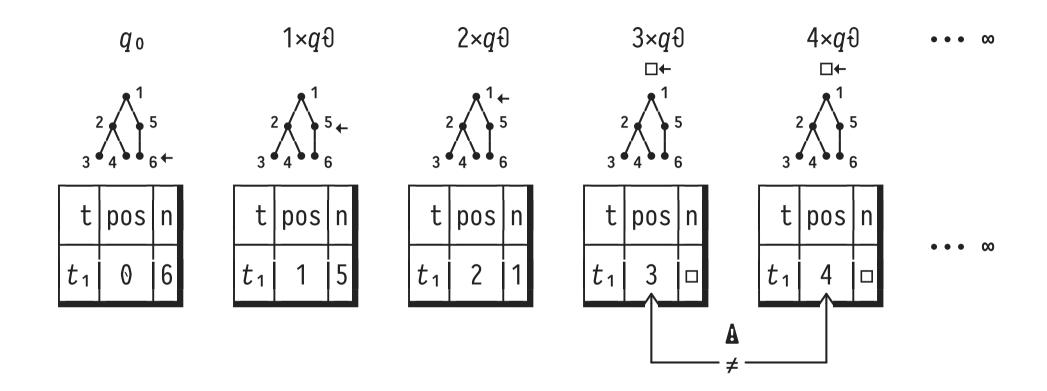
\sim A Trace of the Path in Tree t_1

New rows produced by...



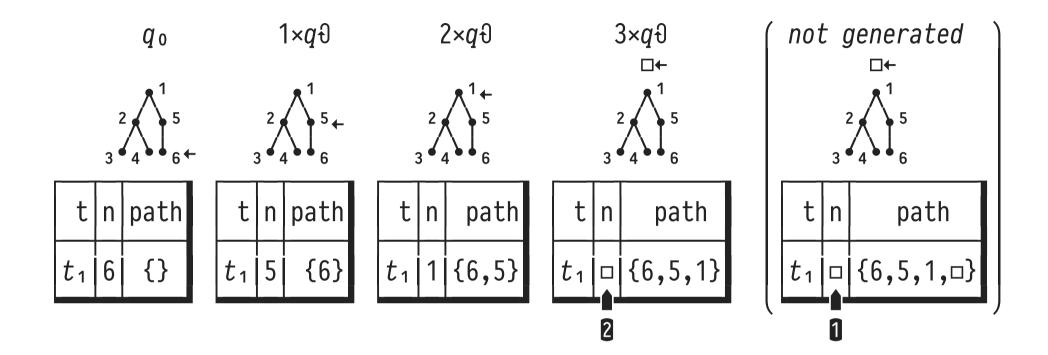
• 4×q0 yields no new rows (recall: t.parents[NULL] ≡ NULL).

\sim Ordered Path in Tree t_1 (New Rows Trace)



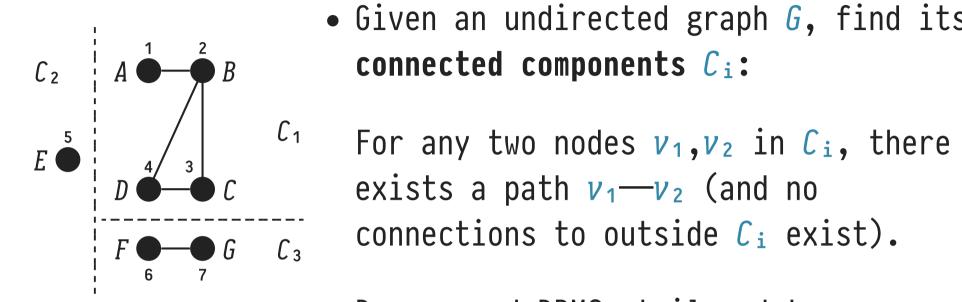
The (non-)generation of new rows to ensure termination is the user's responsibility—a common source of \Re .

\mathcal{F} Path as Array in Tree t_1 (New Rows Trace)



- **1** Ensure termination (enforce ϕ): filter on $n \neq \square$ in $q\theta$.
- 2 Post-process: keep rows of last iteration $(n = \Box)$ only.

2 Connected Components in a Graph

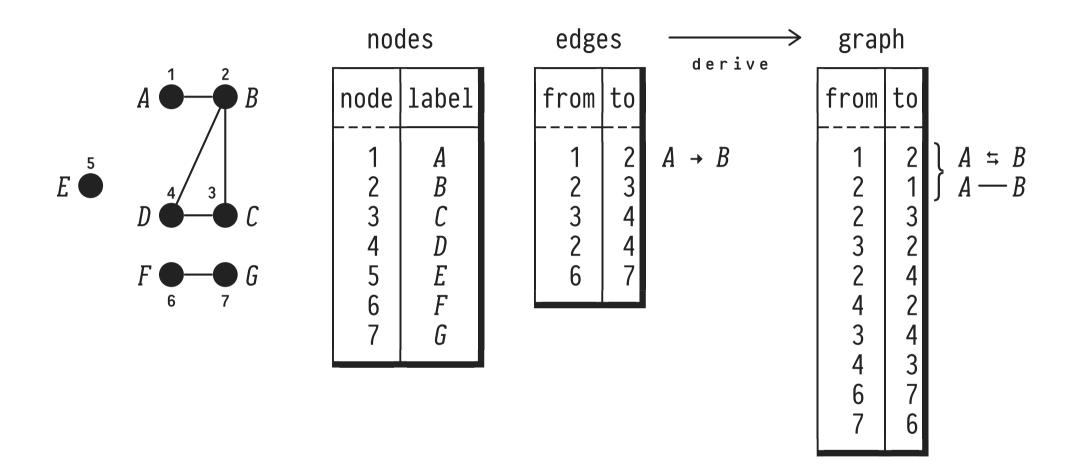


• Given an undirected graph G, find its

• Do we need DBMSs tailored to process graph data and queries?

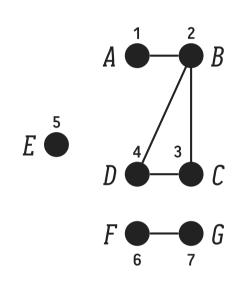
Graphs are (edge) relations. Connected components are the equivalence classes of the reachability relation on G.

Representing (Un-)Directed Graphs



• Use tables nodes and graph to formulate the algorithm.

Computing Connected Components (Query Plan)

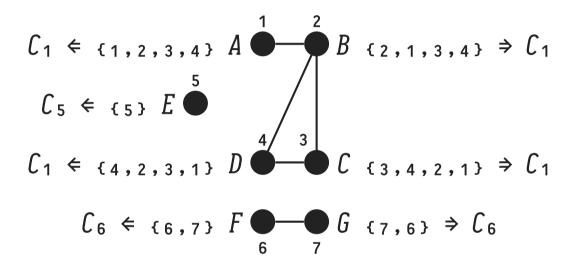


- 1. For each node n, start a **walk** through the graph. Record each node f ("front") that we can **reach** from n.
- 2. For each n, use the **minimum ID** i of all front nodes as n's component C_i .
 - ⇒ Nodes that can reach each other will use the same component ID.

In Step 1, take care to not walk into endless cycles.

Computing Connected Components (Query Plan)

• {...}: Reachable front nodes, C_i derived component ID:



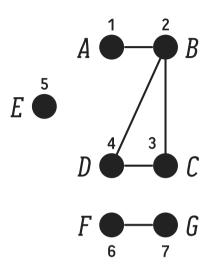
- Tasks for further post-processing:
 - \circ Assign sane component IDs $(\mathcal{C}_1, \mathcal{C}_2, \mathcal{C}_3)$.
 - Extract subgraphs based on components' node sets.

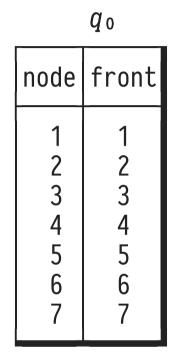
Recursive Graph Walks, From All Nodes at the Same Time

```
WITH RECURSIVE
walks(node, front) AS (
 SELECT n.node, n.node AS front -- (n,n) \in \text{walks:} we can
                          -- reach ourselves
  FROM nodes AS n
   UNION -- only new front nodes will be recorded ✓
 SELECT w.node, g."to" AS front -- record front node
 FROM walks AS w, graph AS g -- \ finds all incident
 WHERE w.front = g."from" -- \int graph edges
```

Invariant: If $(n,f) \in \text{walks}$, node f is reachable from n.

Recursive Graph Walks, From All Nodes at the Same Time





1 2		_
	node	front
2 4 3 2 3 4 4 2 4 3 6 7 7 6	2 2 2 3 3 4 4 6	1 3 4 2 4 2 3 7

 $1 \times q \theta$

node	front
1 1 3 4	3 4 1 1

 $2 \times q\theta$

3× q €					
node front					

3 Recursive Text Processing

- Tree path finding and connected component search used node adjacency information to explore graph structure, iteration by iteration.
- In a variant of this theme, let us view text as lists of adjacent characters that we recursively explore.
- We particularly use the observation (let s :: text,
 n ≥ 1):

```
s = \text{left}(s, n) \mid | \text{right}(s, -n)
prefix of s of length n all but the first n chars of s
```

Set-Oriented (Bulk) Regular Expression Matching

Goal: Given a—potentially large—table of input strings,
validate all strings against a regular expression:2

input		input	parse?
S ₁	\longrightarrow	S ₁	✓
S ₂		S ₂	×
:		:	:
S _n		S _n	✓

• Plan: Parse all s_i in parallel (run n matchers at once).

 $^{^{2}}$ We consider parsing given a context-free grammar in the sequel.

Breaking Bad (Season 2)

Match the **formulæ of chemical compounds** against the regular expression:

compound	formula
citrate	$C_6H_5O_7^{3}$
glucose	C ₆ H ₁₂ O ₆ H ₃ O ⁺
hydronium	H ₃ O+
	•

Table compounds

• Generally: support regular expressions re of the forms c, $[c_1c_2...c_n]$, re_1re_2 , re*, re*, re*, re*, re*, $re_1|re_2$.

From Regular Expression to Finite State Machine (FSM)

Represent re in terms of a deterministic FSM:

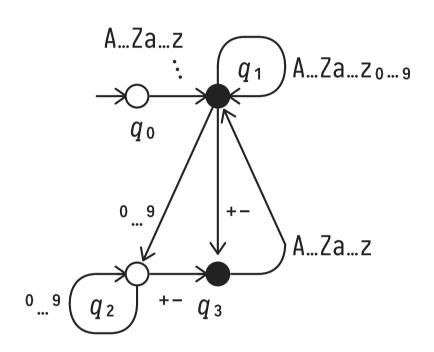


Table **fsm**

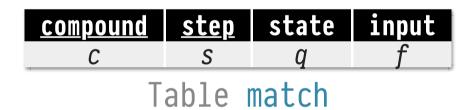
source	labels	target	final?
Q 0 Q 1 Q 1 Q 1 Q 2 Q 2 Q 3	AZaz AZaz ₀ 9 09 +- 09 +- AZaz	Q1 Q1 Q2 Q3 Q2 Q3	false true true false false true

We tolerate the non-key-FD source→final? for simplicity.

Driving the Finite State Machines (Query Plan)

- 1. For *n* entries in table compounds, operate *n* instances of the FSM "in parallel":
 - Each FSM instance maintains its current state and the residual input still to match.

2. Invariant:



• After $s \ge 0$ transitions, FSM for compound c has reached state q. Residual input is f (a suffix of c's formula).

Driving the Finite State Machines (SQL Code)

```
WITH RECURSIVE
match(compound, step, state, input) AS (
SELECT c.compound, 0 AS step, 0 AS state,
       c.formula AS input -- --
FROM compounds AS c -- \equiv q_0
  UNION ALL -- bag semantics (see below)
SELECT m.compound, m.step + 1 AS step, f.target AS state,
       right(m.input, -1) AS input
       match AS m, fsm AS f
FROM
WHERE length(m.input) > 0
AND m.state = f.source
AND strpos(f.labels, left(m.input, 1)) > 0
```

Matching Progess (by Compound / by Step)

1 Focus on indivdiual compound

compound	step	state	input	
citrate	0	0	C ₆ H ₅ O ₇ ³⁻	
citrate	1	1	6H5O73-	
citrate	2	1	H ₅ O ₇ 3-	
citrate	3	1	5073-	
citrate	4	1	073-	
citrate	5	1	73-	
citrate	6	1	3 –	
citrate	7	2	-	empty ¦
citrate	8	3	ε ←	─ string ¦
			:	,
hydronium	0	0	H ₃ O+	l ¦
hydronium	1	1	₃ 0+	
hydronium	2	1	0+	
hydronium	3	1	+	final
hydronium	4	3 ←		state

2 Focus on parallel progress

step	compound	state	input
0	citrate	0	C ₆ H ₅ O ₇ 3-
0	hydronium	0	H ₃ O+
1	citrate	1	6H5O7 ³⁻
1	hydronium	1	₃ 0+
2	citrate	1	$H_5O_7^{3}-$
2	hydronium	1	0+
3	citrate	1	50 ₇ 3-
3	hydronium	1	+
4	citrate	1	073-
4 5	hydronium	3	3
5	citrate	1	73-
6	citrate	1	3 —
7	citrate	2	-
8	citrate	3	3
:	:	:	:

Termination and Bag Semantics (UNION ALL)

The recursive CTE in regular expression matching uses **bag semantics** (UNION ALL). Will matching always **terminate**?

- Column step is increased in each iteration, thus...
 - 1. $q\theta$ will never produce duplicate rows and
 - 2. there is no point in computing the difference $q\theta(t) \setminus r$ in iterate($q\theta, q_0$): $q\theta(t) \cap r = \phi$.
- $q\theta$ is guaranteed to evaluate to ϕ at one point, since...
 - 1. one character is chopped off in each iteration and
 length(m.input) > 0 will yield false eventually, or
 - 2. the FSM gets stuck due to an invalid input character (strpos(f.labels, left(m.input, 1)) yields 0).

4 Recursive Array Processing: Solving Sudoku³ Puzzles

			6				7	5
4				5		8		1
	3			7			2	
		6			1			
			7			5	8	
	9			3				6
	4				9			
		1	8			2		
							3	

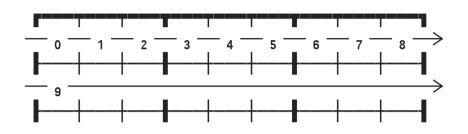
- Fill in the blanks with digits \in {1,...,9} such that
 - 1. no 3×3 box and
 - 2. no row or column

carries the same digit twice.

• Here: encode board as digit array.

³ Japanese: $s\bar{u}(ji)+doku(shin)$, "number with single status." (Yes, this board has a unique solution.)

Row-Major Array-Encoding of a 2D Grid



- Build row-wise int[] array
 of 81 cells ∈ {0,...,9},
 with 0 = blank.
- Derive row/column/box index from cell $c \in \{0, ..., 80\}$:
 - \circ Row of c: $(c / 9) * 9 \in \{0, 9, 18, 27, 36, 45, 54, 63, 72\}$
 - \circ Column of $c: c \% 9 \in \{0,1,2,3,4,5,6,7,8\}$
 - ∘ Box of c: ((c / 3) % 3) * 3 + (c / 27) * 27 ϵ {0,3,6,27,30,33,54,57,60}
- (Clunky—But: relational encodings of grids upcoming.)

Finding All Puzzle Solutions (Query Plan)

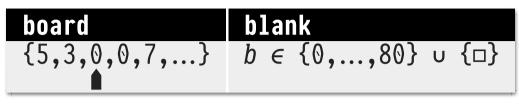


Table sudoku

1. Invariant:

- Column board encodes a valid (but partial) Sudoku board in which the first blank (≡ 0) occurs at index b. If the board is complete, b = □.
- 2. In each iteration, fill in all digits $\in \{1,...,9\}$ at b and keep all boards that turn out valid.

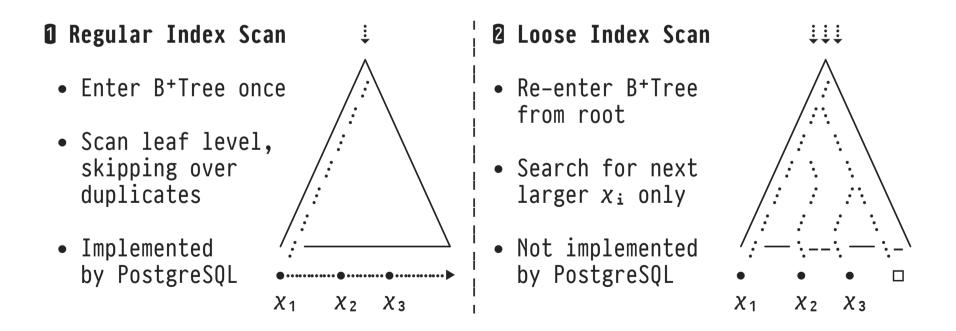
Finding All Puzzle Solutions (SQL Code)

```
WITH RECURSIVE
sudoku(board, blank) AS (
SELECT i.board AS board, array_position(i.board, 0)-1 AS blank
     input AS i
FROM
                                          -- encodes blank
  UNION ALL
         s.board[1:s.b] || fill_in || s.board[s.b+2:81] AS board,
        array_position(
          s.board[1:s.b] || fill_in || s.board[s.b+2:81], 0)-1 AS blank
       sudoku AS s(board, b), generate_series(1,9) AS fill_in
FROM
                               try to fill in all 9 digits
WHERE s.b IS NOT NULL AND NOT EXISTS (
  SFIFCT NULL
  FROM generate_series(1,9) AS i
        9 cells in row/column/box
  WHERE fill_in IN (<digits in row/column/box of s.b at offset i>))
```

5 Emulating Physical Operator Behavior: Loose Index Scans

Implement SELECT DISTINCT t.dup FROM t efficiently, given

- column dup contains a sizable number of duplicates, and
- B+Tree index support on column dup.



Emulating Physical Operator Behavior: Loose Index Scans

```
WITH RECURSIVE
loose(x_i) AS (
 SELECT MIN(t.dup) AS x_i -- } find smallest value x_1 FROM t -- } in column dup
    UNION ALL
  SELECT (SELECT MIN(t.dup) -- find next larger rank t -- value x_i (\equiv NULL
           WHERE t.dup > 1.x_i) AS x_i -- \mid if no such value)
  FROM loose AS 1
  WHERE 1.x; IS NOT NULL -- last search successful?
SELECT 1.xi
FROM loose AS 1
WHERE 1.xi IS NOT NULL
```

Loose Index Scans: Does Recursion Pay Off?

Micro benchmark: |t| = 10° rows, number of duplicates in column dup :: int varies:4

<pre># of distinct values in dup</pre>	index scan [ms]	loose index scan [ms]
10	428	< 1 .
100	440	2 🔔
1000	442	31
10000	454	194
100000	672	1778

Performance comparison

 Recursion beats the built-in index scan if the number of B+Tree root-to-leaf traversals is not excessive.

⁴ PostgreSQL 12.1 on macOS Mojave (10.14.6), 3.3GHz Intel Core i7, 16GB RAM @ 2133 MHz, SATA SSD. Each query run multiple times, average reported here.

6 How SQL Can Tackle Problems in Machine Learning⁵

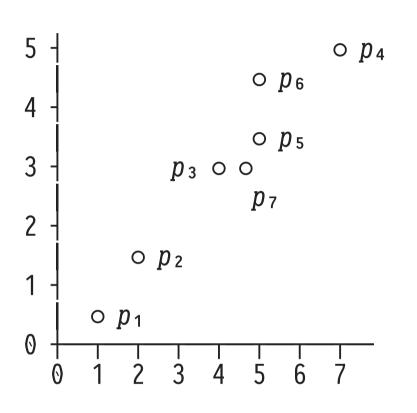
Most sizable *source data* for Machine Learning (ML) problems reside **inside** database systems. Actual *ML algorithms* are predominantly implemented **outside** the DBMS—Python, R, MatLab—however:

- Involves data serialization, transfer, and parsing. ∇
- The main-memory based ML libraries and programming frameworks are challenged by the data volume. 🗣

Demonstrate how ML algorithms (here: K-Means clustering) may be expressed in SQL and thus executed close to the data.

⁵ I apologize for the hype vocabulary.

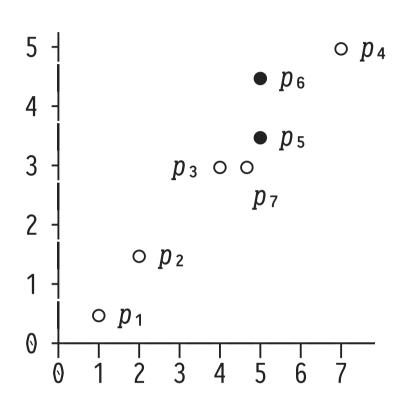
K-Means Clustering



- Goal: Assign each n-dimensional point p_i to one of k clusters (k given).
- Once done, each p_i shall belong to the cluster with the nearest mean (a point that serves as "the prototype of the cluster").

K-Means is computationally difficult (NP-hard) but good approximations/heuristics exist.

K-Means: Lloyd's Algorithm with Forgy Initialization



- Pick k random points (here: p_5 , p_6 for k = 2) as initial means.
- 1. Assignment:

Assign each p_i to nearest mean.

2. Update:

Determine *k* new means to be the **centroids** of the points assigned to each cluster.

Iterate 1. + 2. until assignments no longer change.

K-Means: Forgy Initialization (Query Plan)

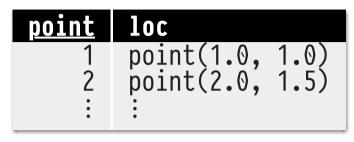


Table points

Picking random rows from table T:

```
TABLE <T>
ORDER BY random()

LIMIT <k> -- pick (at most) k rows

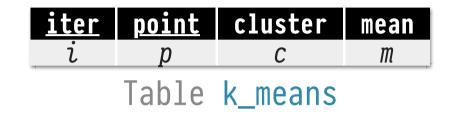
SELECT t.*

FROM <T> AS t

TABLESAMPLE BERNOULLI(s) -- pick ≈ s% random rows
```

K-Means: Lloyd's Algorithm (Query Plan)

Invariant:



- In iteration i, point p has been assigned to cluster c. The mean of cluster c is at location m :: point.
 - o After iteration 0 (initialization), k_means will have k
 rows; later on we have |k_means| = |points|.
- Again: we tolerate the embedded FD cluster → mean.

K-Means: Core of the SQL Code

```
WITH RECURSIVE
k_means(iter, point, cluster, mean) AS (
-- 2. Update
SELECT assign.iter+1 AS iter, assign.point, assign.cluster,
       point(AVG(assign.loc[0]) OVER cluster,
              AVG(assign.loc[1]) OVER cluster) AS mean
        -- 1. Assignment
FROM (SELECT DISTINCT ON (p.point)
                            k.iter, p.point, k.cluster, p.loc
             points AS p, k_means AS k
        ORDER BY p.point, p.loc <-> k.mean) AS assign
WHERE assign.iter < <iterations>
WINDOW cluster AS (PARTITION BY assign.cluster)
```

SQL Notes and Grievance (1)

 We first deconstruct and later reconstruct the points for centroid computation:

```
point(AVG(assign.loc[0]) OVER cluster,
        AVG(assign.loc[1]) OVER cluster) AS mean
```

- Wanted: aggregate AVG() :: bag(point) → point.
 - ♀ In PostgreSQL, we can build user-defined aggregates. 6

⁶ See CREATE AGGREGATE at https://www.postgresql.org/docs/current/xaggr.html.

SQL Notes and Grievance (2)

- K-Means is the prototype of an algorithm that searches for a **fixpoint**. Still, we were using UNION ALL semantics and manually maintain column iter \mathfrak{P} . Why?
 - o There is no equality operator = :: point × point → bool
 in PostgreSQL, a requirement to implement set semantics
 and \ (recall functions iterate(•,•) and recurse(•,•)).

 ② User-defined equality or split point (•[0],•[1]).
 - A strictly increasing iter counter will never lead to a fixpoint anyway ⇒ endless recursion.
 - ♀ User-defined type that admits counting but whose values are all considered equal.

SQL Notes and Grievance (3)

• Is the subquery (1. Assignment) in the recursive query q0 of Lloyd's algorithm the nicest solution? Can't we write:

○ A: No. References to recursive table k_means inside a subquery in SELECT or WHERE clauses are forbidden.

7 | Table-Driven Query Logic (Control Flow → Data Flow)

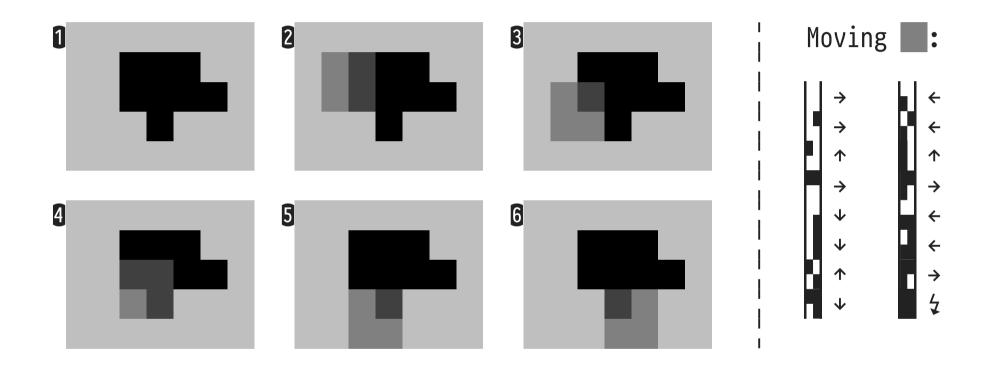
SQL provides a family of constructs to encode the **logic** (in the sense of **control flow**) of algorithms:

- 1. Obviously: WHERE $\langle p \rangle$, HAVING $\langle p \rangle$,
- 2. $\langle q_1 \rangle$ UNION ALL $\langle q_2 \rangle$ UNION ALL \cdots UNION ALL $\langle q_n \rangle$ in which the $\langle q_i \rangle$ contain guards (predicates) that control their contribution,
- 3. CASE WHEN ... THEN ... ELSE ... END.

SQL being a data-oriented language additionally suggests the option to turn control flow into data flow. Encoding query logic in tables can lead to compact, self-describing, and extensible query variants.

Find Isobaric or Contour Lines: Marching Squares

Goal: Trace the boundary of the object — in **①:**



• 15 cases define the movement of the 2×2 pixel mask.

Marching Squares (Query Plan)

- 1. **Encode mask movement** in table directions that maps 2×2 pixel patterns to $(\Delta x, \Delta y) \in \{-1,0,1\} \times \{-1,0,1\}$. Examples: \blacksquare maps to $(1,0) \rightarrow$, \blacksquare maps to $(0,-1) \uparrow$.
- 2. For each 2D-pixel p_0 , read pixels at $p_0+(1,0)$, $p_0+(0,1)$, $p_0+(1,1)$, to form a 2×2 squares map [table squares].
- 3. Iteratively fill table march(x,y):
 - ∘ $[q_0]$: Start with (1,1) ∈ march.
 - \circ [q0]: Find 2×2 pixel pattern at (x,y) in squares, lookup pattern in directions to move mask to (x,y) + $(\Delta x, \Delta y)$.

Marching Squares (SQL Code)

```
WITH RECURSIVE
march(x,y) AS (
 SELECT 1 AS x, 1 AS y
    UNION
 SELECT new.x AS x, new.y AS y
       march AS m, squares AS s,
 FROM
        directions AS d,
        LATERAL (VALUES (m.x + (d.dir).\Delta x)
                           m.y + (d.dir).\Delta y) AS new(x,y)
WHERE (s.11, s.1r, s.u1, s.ur) = (d.11, d.1r, d.u1, d.ur)
AND \qquad (m.x,m.y) = (s.x,s.y)
```

* Table lookup replaces a 15-fold case distinction.

Cellular automata (CA)⁷ are discrete state-transition systems that can model a variety of phenomena in physics, biology, chemistry, maths, or the social sciences:

- **Cells** populate a regular *n*-dimensional **grid**, each cell being in one of a finite number of **states**.
- A cell can interact with the cells of its neighborhood.
- State of cell c changes from generation to generation by a fixed set of rules, dependent on c's state and those of its neighbors.

 $^{^7}$ Discovered by Stanislaw Ulam and John von Neumann in the 1940s at Los Alamos National Laboratory.

Cell State Change in Cellular Automata

Here, we will distinguish two flavors of CA:

1 Cell c is influenced by its neighborhood (c's next state is a function of the cell states in the neighborhood)

[Conway's Game of Life]

2 Cell c influences cells in its neighborhood (c contributes to state changes to be made in the neighborhood)

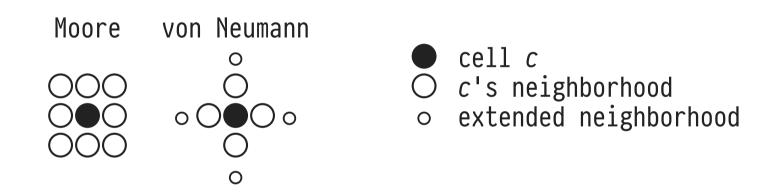
[Fluid simulation]

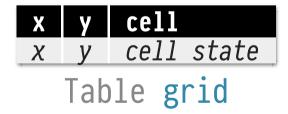
Both flavors lead to quite different SQL implementations.

1 is (almost) straightforward, 2 is more involved. Let us discuss both.

Cell **neighborhood** is flexibly defined, typically referring to (a subset of) a cell's *adjacent* cells:

• Types of neighborhoods, for n = 2 (2D grid):





Accessing the Cell Neighborhood — Non-Solution! =

• Excerpt of code in $q\theta$ (computes next generation of grid), access the Moore neighbors n of cell c:

- Looks like a suitable CA core (f, agg encode CA rules).
- BUT refers to recursive table more than once: 4 in SQL.

Interlude: WITH RECURSIVE — Syntactic Restrictions

WITH RECURSIVE syntactically restricts query forms, in particular the references to the recursive table T:

- 1. No references to T in q_0 .
- 2. A single reference to T in $q\theta$ only (linear recursion).
- 3. No reference to T in subqueries outside the FROM clause.
- 4. No reference to T in INTERSECT or EXCEPT.
- 5. No reference to T in the null-able side of an outer join.
- 6. No aggregate functions in $q\theta$ (window functions do work).
- 7. No ORDER BY, OFFSET, or LIMIT in $q\theta$.

Enforces **distributivity**: $q\theta(T \cup \{t\}) = q\theta(T) \cup q\theta(\{t\})$, allowing for incremental evaluation of WITH RECURSIVE.

Accessing the Cell Neighborhood — A Solution! 😅

Window functions admit access to rows in cell vicinity:

```
SELECT \cdots f(c.cell, agg(c.cell)) OVER (c.cell) (c.cell) (c.cell) (c.cell) (c.cell)
```

Conway's Game of Life

Life⁸ simulates the evolution of cells c (state: either alive or dead) based on the population count $0 \le p \le 8$ of c's Moore neighborhood:

- 1. If c is alive and p < 2, c dies (underpoulation).
- 2. If c is alive and $2 \le p \le 3$, c lives on.
- 3. If c is alive and 3 < p, c dies (overpopulation).
- 4. If c is dead and p = 3, c comes alive (reproduction).

Note: The next state of *c* is a function of the neighborhood states. *c* does *not* alter cell states in its neighborhood.

⁸ John H. Conway († April 2020), column *Mathematical Games* in *Scientific American* (October 1970).

Life — A Few Notable Cell Patterns

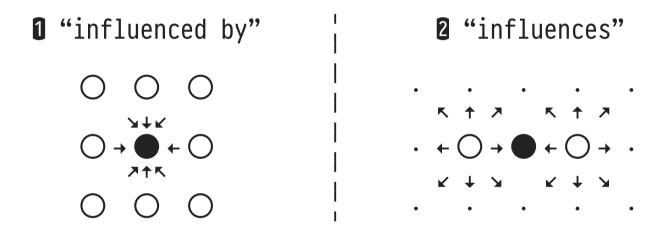
Still	Oscillators (period: 2)	Spaceships
· • • · •		
. ••• .	· •••	

\sim Life — SQL Encoding of Rules (f: below, agg = SUM)

```
WITH RECURSIVE
life(gen,x,y,cell) AS (
SELECT 1.gen + 1 AS gen, 1.x, 1.y,
        CASE (l.cell, ( SUM(l.cell) OVER <horizontal --->
                       + SUM(1.cell) OVER < vertical :>
                       + SUM(1.cell) OVER < diagonal :>
                       + SUM(1.cell) OVER < diagonal :>
          -- (c, p): c = \text{state of cell}, p = # of live neighbors}
          WHEN (1, 2) THEN 1 -- ]
          WHEN (1, 3) THEN 1 -- } alive
          WHEN (0, 3) THEN 1 --
          ELSE
                           0 -- dead
        END AS cell
 FROM life AS 1
```

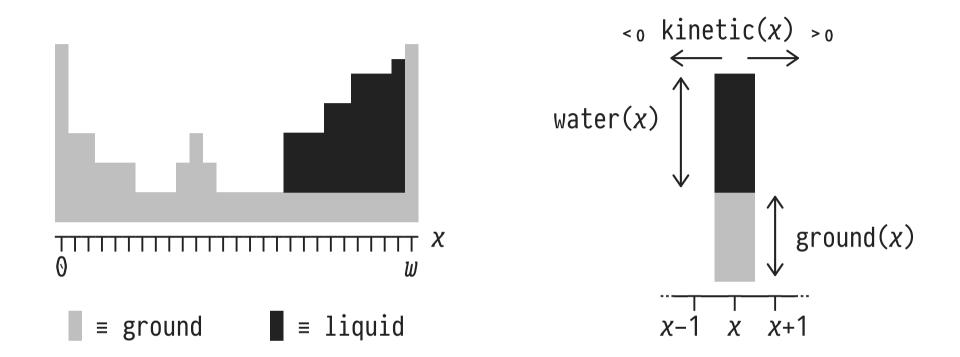
9 CA with Cells That Influence Their Neighborhood

If cells assume an **active role** in influencing the next generation, this suggests a different SQL implementation.



In type ②, cells ○ actively influence their neighbors.
 Affected cells ● need to accumulate these individual influences (up to 8 in this grid—only two shown here).

Simulate the Flow of Liquid (in a 1D Landscape)



Goal: Model two forms of energy in this system:

- potential energy at x (pot(x) = ground(x) + water(x))
- left/right kinetic energy at x (kinetic(x))

Liquid Flow: Cellular Automaton⁹

```
\Deltawater \leftarrow (0,0,...,0) -- changes to water and energy levels
\Delta kin \leftarrow (0,0,...,0) -- in next generation
for x in 1...w-1:
  -- liquid flow to the left?
  if pot(x)-kin(x) > pot(x-1)+kin(x-1):
                                                      -- force ← > force →
  | flow \leftarrow \frac{1}{4} \times \min(\text{water}(x), \text{pot}(x) - \text{kin}(x) - (\text{pot}(x-1) + \text{kin}(x-1)))
                                      -- aggregate the
-- influences on
   \Deltawater(x-1) + \Deltawater(x-1)+flow
 \Delta kin(x-1) \leftarrow \Delta kin(x-1) - \frac{1}{2} \times kin(x-1) - flow -- | cells @ x / x-1
  -- liquid flow to the right?
  if pot(x)+kin(x) > pot(x+1)-kin(x+1):
                                              -- force → > force ←
  -- "mirror" the above code
water ← water + ∆water —— \[ \] apply the aggregated influences
                                  -- | to all cells (ground is constant)
kin ← kin + ∆kin
```

⁹ CA rules adapted from those posted by user YankeeMinstrel on the Cellular Automata . $\frac{1}{4}$, $\frac{1}{2}$ are (arbitrary) dampening/friction factors. See https://www.reddit.com/r/cellular_automata/.

CA with Neighborhood Influencing Rules: SQL Template

```
WITH RECURSIVE
cells(iter,x,y,state) AS (
 SELECT co.iter + 1 AS iter, co.x, co.y,
        c0.state ⊕ COALESCE(agg.∆state, ⟨z⟩) AS state
        cells AS c0 LEFT OUTER JOIN
FROM
        -- find and aggregate influences on all cells @ x,y
                                         -- \ \mathrew{\text{m}} encodes rules
            ////////// AS agg(x,y,\Deltastate) -- \ of the CA
           extract all influences on cell c0 (□ if none)
        ON (c0.x, c0.y) = (agg.x, agg.y)
WHERE c0.iter < <iterations>
```

- Design: no $agg(x,y,_)$ if cell @ x,y doesn't change state.
- Assume that z is neutral element for \oplus : $s \oplus z = s$.

CA: From Individual to Aggregated Influences (SQL Template)

```
SELECT c0.iter + 1 AS iter, c0.x, c0.y,
      c0.state ⊕ COALESCE(agg.∆state, <z>) AS state
FROM
      cells AS c0 LEFT OUTER JOIN
      -- find and aggregate influences on all cells @ x,y
      (SELECT infs.x, infs.y, <agg>(infs.Δstate) AS Δstate
       FROM
               GROUP BY infs.x, infs.y
      ) AS agg(x,y,\Delta state)
      -- extract all influences on cell c0 (□ if none)
      ON (c0.x, c0.y) = (agg.x, agg.y)
```

- $(x,y,\Delta state) \in infs: individual influence on cell @ x,y.$
- Typically, we will have $agg = (\phi, z, \oplus)$.

CA: Individual Neighborhood Influences (SQL Template)

- For each cell c1, $\frac{1}{2}$ computes an array of influence influence with elements $(x,y,\Delta state)$: c1 changes the state of cell $(x,y,\Delta state)$.
- For each c1, influence may have 0, 1, or more elements.

CA: Encoding Neighborhood Influencing Rules (SQL Template)

```
(SELECT (CASE WHEN \langle p_1 \rangle THEN -- if \langle p_1 \rangle holds, then c1 has ...
           \frac{\text{array}[\text{ROW}(\text{c1.x-1, c1.y,}]),}{\text{ROW}(\text{c1.x, c1.y+1,}]} -- influence on ← cell
         END
      | | CASE WHEN < p_2 > THEN
           -- x y ∆state
        ) AS influence
     cells AS c1
 FROM
WINDOW horizontal AS — -- \ provide frames to access neighbors
WINDOW vertical AS \cdots -- of c1 in \langle p_i \rangle, ...., and
) AS inf(influence)
```

Admits straightforward transcription of rules into SQL.

CA: Summary of Influence Data Flow (Example)

• Assume \triangle state :: int, $agg \equiv SUM$ (i.e., $z \equiv 0$, $\oplus \equiv +$):

1 Table inf

	influence			
$\{(1, 1), (2,$	3,+4),(1,4,-2)} 3,-3),(1,3,+1)} 2,-5)} 4,+2)}			

neighborhood influence, computed based on current cell generation 2 Table infs

X	У	∆state
1 1 1	<u>ა</u> ა ა	+4 -3 +1
1	4	-2 +2
2	2	-5

3 Table agg

X	У	∆state
1 1 2	3 4 2	+2 0 -5

apply to current cell states using ⊕ to find next generation

Working Around the Linear Recursion Restriction

- Work around 10 linearity restriction for recursive table T:
 - Use non-recursive WITH to introduce new name \overline{T} for T,
 - $\mathbf{2}$ refer to \overline{T} as often as needed.

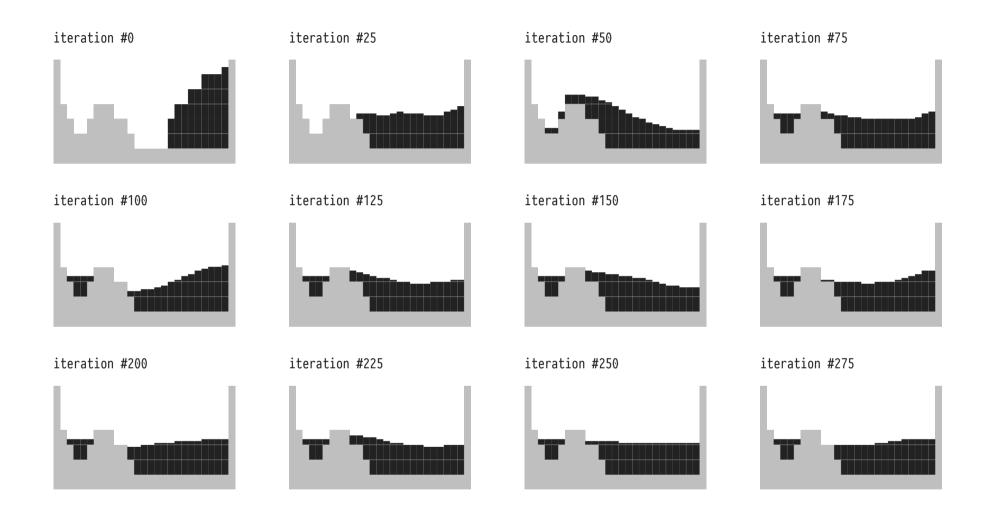
```
(WITH \overline{T}(\cdots) AS (TABLE T) -- original recursive query q\theta \\
SELECT \cdots
FROM \cdots, \overline{T} AS t1, \cdots, \overline{T} AS t2, \cdots
-- 2
```

This is closer to a hack than conceptual beauty. Due to SQL's scoping rules, however, we may choose $\overline{T} = T$ such that the original query may be left untouched.

Liquid Flow (SQL Code)

```
WITH RECURSIVE
sim(iter.x.ground.water.kinetic) AS (
 SELECT 0 AS iter, f.x, f.ground, f.water, 0.0 AS kinetic
  FROM fluid AS f
    UNION ALL
  (WITH sim(iter,x,ground,water,kinetic) AS (TABLE sim)
                                                                                     -- non-linearity "hack"
   SELECT so.iter + 1 AS iter, so.x, so.ground,
          s0.water + COALESCE(agg.∆water , 0) AS water,
          s0.kinetic + COALESCE(agg. Akinetic, 0) AS kinetic
   FROM sim AS s0
            LEFT OUTER JOIN
          LATERAL (SELECT infs.x, SUM(infs. Awater) AS Awater, SUM(infs. Akinetic) AS Akinetic
                   FROM (SELECT (-- flow to the left
                                   CASE WHEN <p1>
                                   THEN array[ROW(s1.x-1, <Δwater>, <Δkinetic>),
                                                                                                 Specific rules for the Liquid Flow CA,
                                              ROW(s1.x , <Δwater>, <Δkinetic>),
                                                                                                 the enclosing SQL code is generic.
                                              ROW(s1.x-1, <Δwater>, <Δkinetic>)
                                                                                                 • Use CASE --- WHEN --- THEN --- END to implement
                                   END
                                                                                                   conditional rules.
                                   -- flow to the right
                                                                                                 • Use windows to access cell neighborhood.
                                   CASE WHEN <p2>
                                   THEN array[ROW(s1.x+1, <Δwater>, <Δkinetic>),
                                                                                                 • Use array concatenation (||) to implement
                                               ROW(s1.x , <\Delta water>, <\Delta kinetic>),
                                                                                                   sequences of rules.
                                               ROW(s1.x+1, <\Delta water>, <\Delta kinetic>)
                                   END
                                  ) AS influence
                           FROM sim AS s1
                           WINDOW horizontal AS (ORDER BY s1.x)
                          ) AS inf(influence),
                          LATERAL unnest(inf.influence) AS infs(x int, Awater numeric, Akinetic numeric)
                   GROUP BY infs.x
                   ) AS agg(x, Δwater, Δkinetic)
                   ON (s0.x = agg.x)
   WHERE s0.iter < <iterations>
SELECT s.iter, s.x, s.ground, s.water
FROM sim AS's
ORDER BY s.iter, s.x;
```

Liquid Flow (First 275 Intermediate Simulation States)



10 Parsing with Context-Free Grammars

One of the classic problems in Computer Science: parsing.

• Given the productions of a **context-free grammar**, can the input string be parsed (≡ generated) by the grammar?

```
start symbol production rule (lhs→rhs)

Expr → Expr Plus Term | Term
Term → Term Mult Fact | Fact
Fact → '1'
Plus → '+'
Nult → '×'
Non-terminal terminal

Grammar for simple
arithmetic expressions:

operators +/×, literal 1,

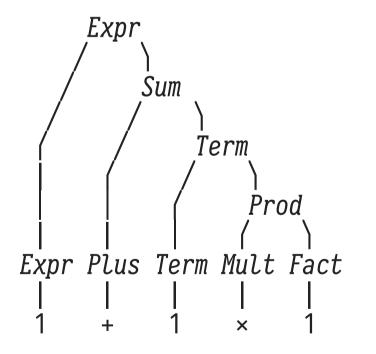
+/× left-associative,
op precedence: × > +.
```

Chomsky Normal Form and Parse Trees

We consider grammars in **Chomsky Normal Form** only: rules read $lhs \rightarrow terminal$ or $lhs \rightarrow non-terminal$ non-terminal.

Expr → Expr Sum
Expr → Term Prod
Expr → '1'
Term → Term Prod
Term → '1'
Sum → Plus Term
Prod → Mult Fact
Fact → '1'
Plus → '+'
Mult → '×'

Parse tree for input 1+1×1:



A Tabular Encoding of Chomsky Grammars

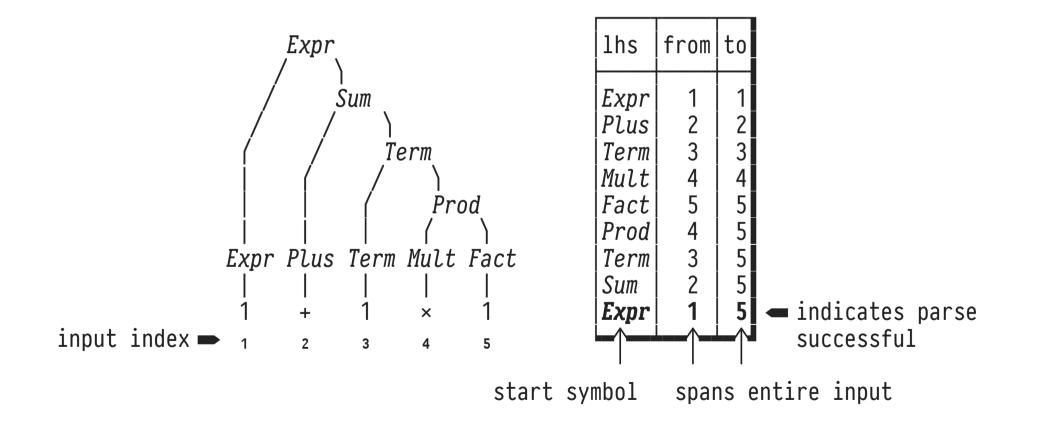
Simple encoding of the sample arithmetic expression grammar:

lhs	sym	rhs ₁	rhs ₂	start?
Expr		Expr	Sum	true
Expr		Term	Prod	true
Expr	1			true
Term		Term	Prod	false
Term	1			false
Sum		Plus	Term	false
Prod		Mult	Fact	false
Fact	1			false
Plus	+			false
Mult	×			false

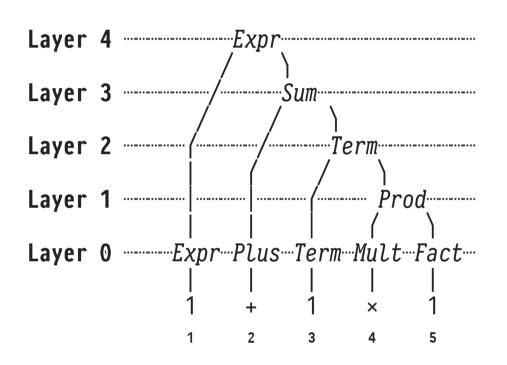
- Exploits that rules can have one of two forms only.
- Embedded FD lhs → start? identifies one non-terminal as the grammar's start symbol (here: Expr).

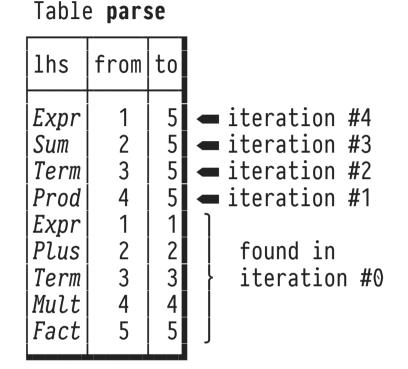
Building a Parse Tree, Bottom Up

Invariant: Keep track of which part of the input (index from
to to) can be generated by the lhs of a rule:



Building a Tree in Layers Requires Access to the Past

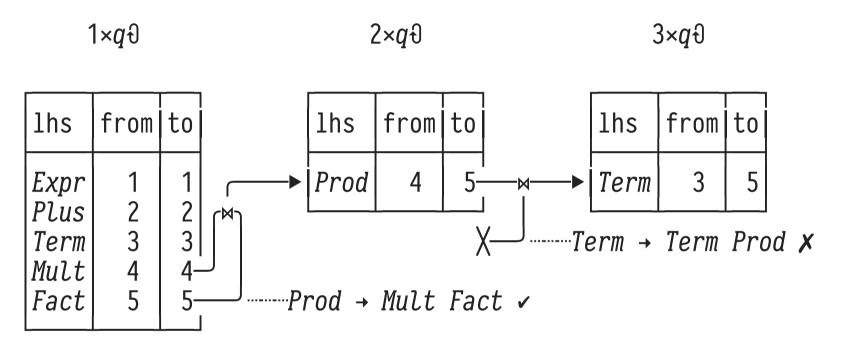




• To establish *Term* at **Layer 2** (iteration #2), we need *Prod* (**Layer 1**, iter #1 ✓) and *Term* (**Layer 0**, iter #0 鬟).

WITH RECURSIVE's Short-Term Memory

Rows seen in table parse by...



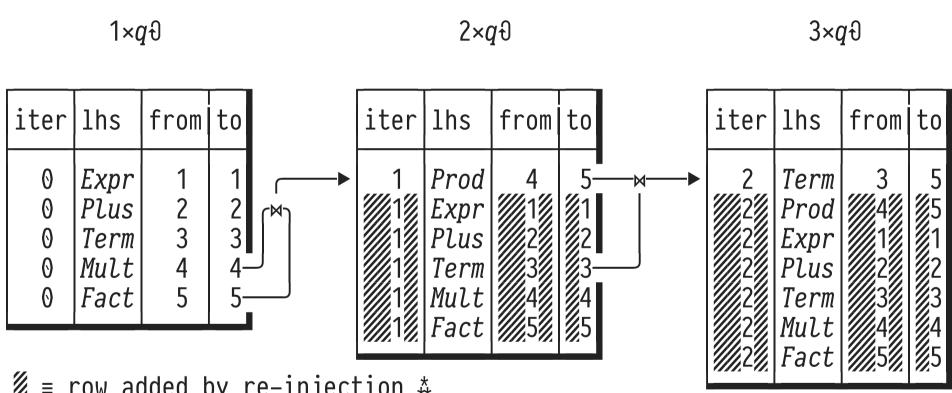
• Parsing fact (Term, 3,3) has been discovered by q_0 —more than one iteration ago—and is *not* available to $2 \times q_0$.

Re-Injecting Early Iteration Results (SQL Template)

```
WITH RECURSIVE
T(\text{iter}, c_1, ..., c_n) AS (
                                     -- \} add column iter (= 0) to -- \} result of q_0
  SELECT 0 AS iter, t.*
  FROM (q_0) AS t
    UNION ALL
  (WITH T(\text{iter}, c_1, ..., c_n) AS (TABLE T) -- non-linear recursion
   SELECT t.iter + 1 AS iter, t.*
   FROM (TABLE T
                                      -- (will be kept since iter advances)
             UNION
                                      -- original q\theta (refers to T)
           q\theta
          ) AS t
   WHERE p
                                      -- stop condition
```

WITH RECURSIVE With Long-Term Memory

Rows seen in table parse by...



Parsing: Cocke-Younger-Kasami Algorithm (CYK)

The **CYK algorithm** builds parse trees bottom up, relying on formerly discovered partial parses (dynamic programming):

- Iteratively populate table parse(lhs, from, to):
 - ∘ $[q_0]$: For each $lhs \rightarrow terminal$: if terminal is found at index from...to in input, add (lhs, from, to) to parse.
 - o [q0]: For each pair ($lhs_1, from_1, to_1$), ($lhs_2, from_2, to_2$) in parse × parse: 11 add ($lhs_3, from_1, to_2$) if
 - 1. $to_1 + 1 = from_2$ and
 - 2. $lhs_3 \rightarrow lhs_1 lhs_2$.

¹¹ Implies a self-join of parse, leading to non-linear recursion.

Parsing Using CYK (Core SQL Code)

```
WITH RECURSIVE
parse(..., lhs, "from", "to") AS (
 SELECT ..., g.lhs, i AS "from", i + length(g.sym) - 1 AS "to"
  FROM
       grammar AS g,
        generate_series(1, length(input)) AS i,
  WHERE g.sym IS NOT NULL
 AND substr(input, i, length(g.sym)) = g.sym
   UNION ALL
                                   -- A re-injection code omitted
  SELECT ..., g.lhs, l."from", r."to"
  FROM
       grammar AS g,
        parse AS 1, parse AS r -- A non-linear recursion
  WHERE 1."to" + 1 = r."from"
     (g.rhs1, g.rhs2) = (1.1hs, r.1hs)
  AND
```