

▼ Overview

▼ Stream Processing

▼ Applications

- Stock Markets
- Internet of Things
- Intrusion Detection

▼ Central Idea

- **Classical Queries:** Queries Change, Data Fixed
- **View Maintenance:** Data Changes, Queries Fixed, Slow Response
- **Here:** Data Changes, Queries Fixed, Fast Response

▼ Language Models

- Classical SQL w/ Windows
- Stream-specific query langs

▼ Challenges & Advantages

- Limited Compute Time: Want to deal with large numbers of records as they come in quickly.
- All compute requirements (structurally, at least) are given upfront.
- Typically specialized for bounded data sizes

▼ Cayuga

▼ Stream Definition Operators

▼ SELECT x, y, z FROM [stream]

- Classical Projection. Optionally defines a new stream
- Optional PUBLISH clause names the stream

▼ FILTER { condition } [stream]

- Classical Selection. Pass only tuples that pass a condition

▼ [stream] NEXT { condition } [stream]

- “JOIN”-like operation
- ▼ For each tuple on the LHS
 - Find (and emit) the next tuple from the RHS that matches the condition

▼ [stream] FOLD { group_condition, done_condition, aggregate } [stream]

- “JOIN+AGGREGATE”-like operation
- ▼ For each tuple on the LHS
 - Start a group
 - Attach each tuple from the RHS that matches group_condition
 - Update the group with the aggregate expression
 - If the RHS tuple matches done_condition, close out the group and emit the aggregate

▼ Discussion

▼ Why not use regular joins

▼ Regular Joins are Non-Streaming

- ▼ Unclear when a tuple stops being relevant
 - Unbounded memory use
 - Steadily growing compute

- ▼ Language chosen to ensure finite state per tuple being joined

- ▼ NEXT: State = unmatched tuples from LHS
 - One-One join

- ▼ FOLD: State = unfinished groups: Constant per LHS tuple
 - One-Many join

- **What about many/many?**

- ▼ Hard to express temporal relationships w/ joins

- WHERE t2 > t1 and/or some sort of nested subquery trickery to get LIMIT

▼ Autometa

▼ DFA

▼ Data Model

- Nodes represent states
- Edges represent transitions
- One node designated as the “start” state
- One or more nodes designated as “terminal” or “output” states

▼ Language

- Start with an alphabet $[\Sigma]$
- Edges labeled with letters in the alphabet
- ▼ Every node has an out-edge for every letter in the alphabet
 - Implicit ‘error’ state if no edge for a letter given explicitly

▼ Evaluation

- Given a string in $[\Sigma]$
- For each letter in the string travel the edge with the same label.
- “Success” if you end in one of the terminal states.

▼ NDFA

▼ Data Model

- Same as DFA, but allowed to have >1 edges with the same label.

▼ Evaluation

- At any given point in time, you can be “present” at multiple nodes/states
- If at a state with multiple out-edges labeled with the same letter as the next letter in the string, travel to all of them in parallel

▼ Reduction to DFA

- Given an NDFA with N states (e.g., {A, B, C}), create a new graph with 2^N states, call them hyperstates ({ }, {A}, {B}, {C}, {AB}, {AC}, {BC}, {ABC})
- Each state represents the state of the NDFA where you are in some subset of the N states (there are 2^N such states)
- ▼ For each hyperstate (e.g., {AB})...
 - ▼ For each letter in the alphabet
 - ▼ For each state in the hyperstate (e.g., A and B)
 - Compute the set of states that the state would transition to for that letter
 - Compute the union of these states
 - This is the hyperstate that you transition to

▼ Cayuga-Autometta

▼ Data Model

- ▼ Same as NDFA, but extended in one additional dimension: Every state has a set of associated instances
 - Like a generalization from Zeroth- to First-order logic
 - $\text{AliceIsASStudent} \rightarrow \text{AliceIsInClass}$ vs $\text{IsStudent}(x) \rightarrow \text{IsInClass}(x)$
 - Strictly more powerful (infinite number of states)
- In short, every state behaves like a relation
- Edges represent opportunities for tuples to travel from one relation to another.
- ▼ Edges are labeled with
 - Condition (for the tuple to travel)
 - Projection rule (for generating the new tuple)

▼ Reducing CEL to Cayuga

▼ SELECT

- (True, Projection Targets) \rightarrow Next State

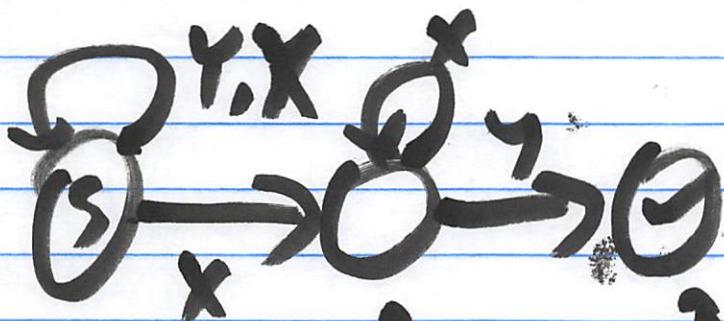
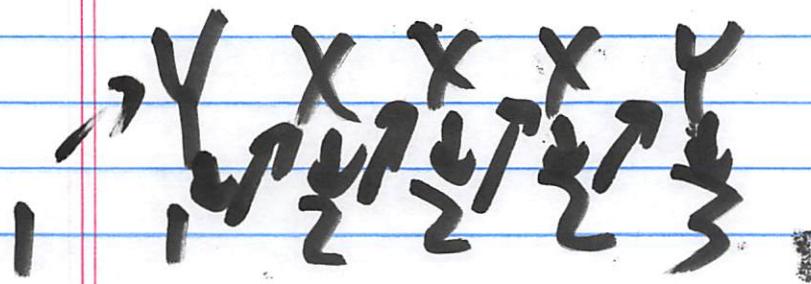
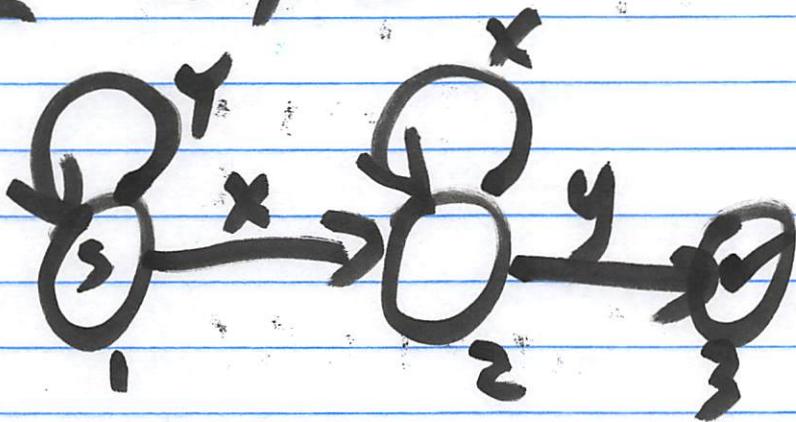
▼ NEXT

- $(\neg \text{condition}, \text{ID}) \rightarrow \text{Same State}$
- $(\text{condition}, \text{ID}) \rightarrow \text{Next State}$

▼ FOLD

- $(\text{group_condition}, \text{aggregate}) \rightarrow \text{Same State}$
- $(\neg \text{group_condition}, \text{ID}) \rightarrow \text{Same State}$
- $(\text{done_condition}, \text{ID}) \rightarrow \text{Next State}$

q(PM) $\sum \in [x, y]^n$



13 13 1,23 1,23 1,23 1,33

stream
::=

SELECT x,y,z FROM [stream]

↳ RA proj

FILTER { φ } [stream]

↳ RA select in on φ

[stream] ^A NEXT { φ } ^B [stream]

↳ JOIN lite

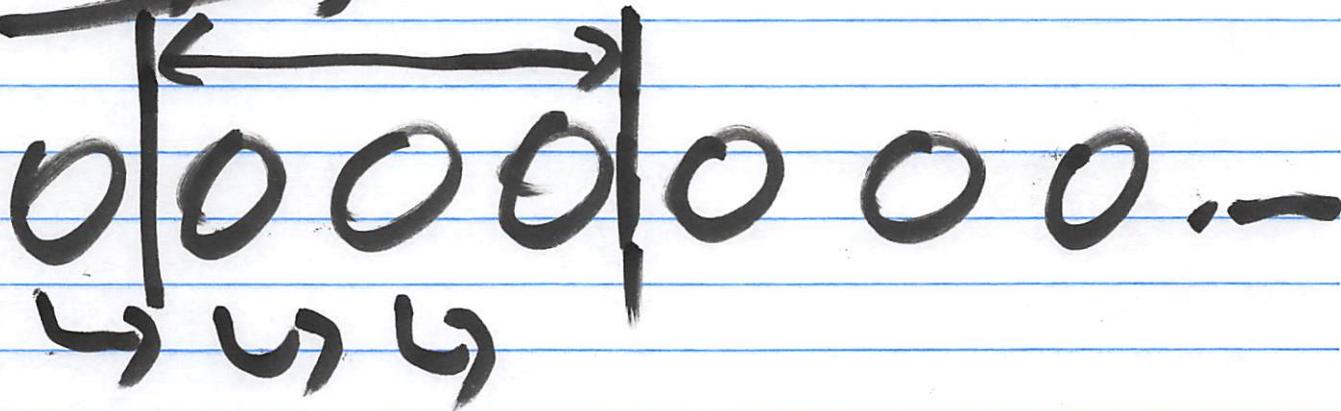
↳ for each A
find next B matching φ

~~output~~ →
~~Stream~~ Filter == ~~SAC/RA~~ SELECT
Aggregate
(Group-By)



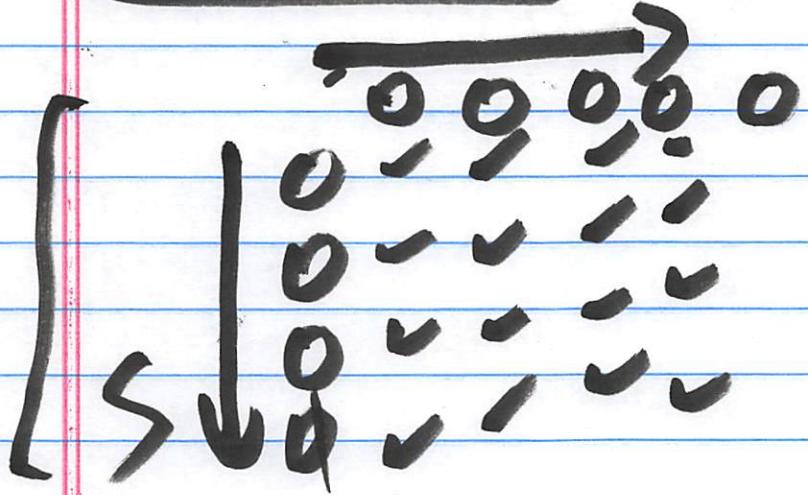
PROJECT
JOIN?

Aggregate



- Cumulative
 - ↳ per tuple
 - ↳ per 'breakpoints'
 - ↳ externally triggered
- GROUP BY
- WINDOW

JOIN R



- WINDOW

NEXT - MERGE JOIN (sorted Data)

- 1-1 Join

↳ But need to make sure matches show up fast

L

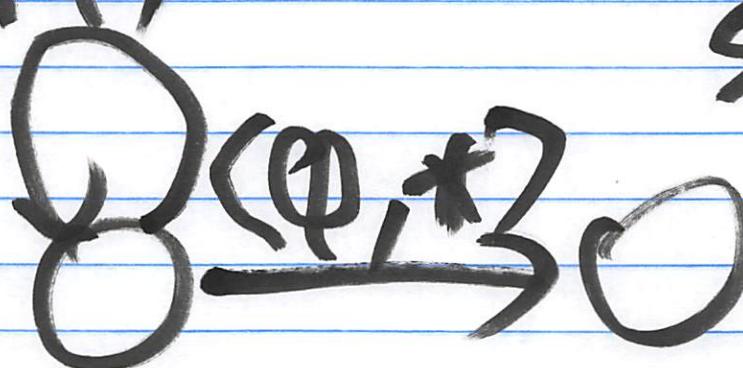
= 1-ManyJoin + Aggregate

OLD

Stocks (Ticker, Price)

Stocks NEXT { \$1.Ticker
= \$2.Ticker
AND \$1. Price
> \$2.Price }
Stocks \$2

(1φ, #7) > (2φ, #3)



IBM \$22
MSFT \$23

A [stream] FOLD {

Group,
done,

agg register

3 [stream]

B ↳ 1-many join + agg

↳ Every tuple in A
starts a group

↳ Agg over tuples in B
emit join on group

↳ Emit when done