

# Ohua-Powered, Semi-Transparent UDF's in the Noria Database

By Justus Adam

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# A query to start with

**Query:** How many clicks, on average, does it take for a user to get from the **start page** to a **purchase**

uid	Category	Timestamp
1	1	001
1	0	005
1	2	010

Table layout

```
SELECT avg(pageview_count)
FROM
( SELECT
    c.user_id, matching_paths.ts1,
    count(*) - 2 as pageview_count
  FROM
    clicks c,
    ( SELECT user_id, max(ts1) as ts1, ts2
      FROM
        ( SELECT DISTINCT ON (c1.user_id, ts1)
            c1.user_id,
            c1.ts as ts1,
            c2.ts as ts2
          FROM clicks c1, clicks c2
          WHERE
            c1.user_id = c2.user_id AND
            c1.ts < c2.ts AND
            c1.category = 1 AND
            c2.category = 2
          ORDER BY
            c1.user_id, c1.ts, c2.ts
        ) candidate_paths
      GROUP BY user_id, ts2
    ) matching_paths
  WHERE
    c.user_id = matching_paths.user_id AND
    c.ts >= matching_paths.ts1 AND
    c.ts <= matching_paths.ts2
  GROUP BY
    c.user_id, matching_paths.ts1
) pageview_counts;
```

5. Average of the count, per user

1. The table, but more than once

3. Only the non-overlapping ones

2. Delimiters for an ordered sequence, if user is the same

4. The actual clicks in between the sequence, if user is the same

- Eric Friedman, Peter Pawlowski, and John Cieslewicz. 2009. SQL/MapReduce: a practical approach to self-describing, polymorphic, and parallelizable user-defined functions. *Proc. VLDB Endow.* 2, 2 (August 2009), 1402-1413.

# Coding it up

**Query:** How many clicks, on average, does it take for a user to get from the start page to a purchase

uid	Category	Timestamp
1	1	001
1	0	005
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Table layout

```
SELECT avg(pageview_count)
FROM
( SELECT
    c.user_id, matching_paths.ts1,
    count(*) - 2 as pageview_count
FROM
    clicks c,
    ( SELECT user_id, max(ts1) as ts1, ts2
      FROM
        ( SELECT DISTINCT ON (cl.user_id, cl.ts)
          cl.user_id,
          cl.ts as ts1,
          c2.ts as ts2
        FROM clicks c1, clicks c2
        WHERE
          c1.user_id = c2.user_id AND
          c1.ts < c2.ts AND
          c1.category = 0 AND
          c2.category = 1
        ORDER BY
          c1.user_id,
          candidate_paths
        GROUP BY user_id
      ) matching_paths
    WHERE
      c.user_id = matching_paths.user_id AND
      c.ts >= matching_paths.ts1 AND
      c.ts <= matching_paths.ts2
    GROUP BY
      c.user_id, matching_paths.ts1
  ) pageview_counts;
```

Easier<sup>[1,2]</sup>:

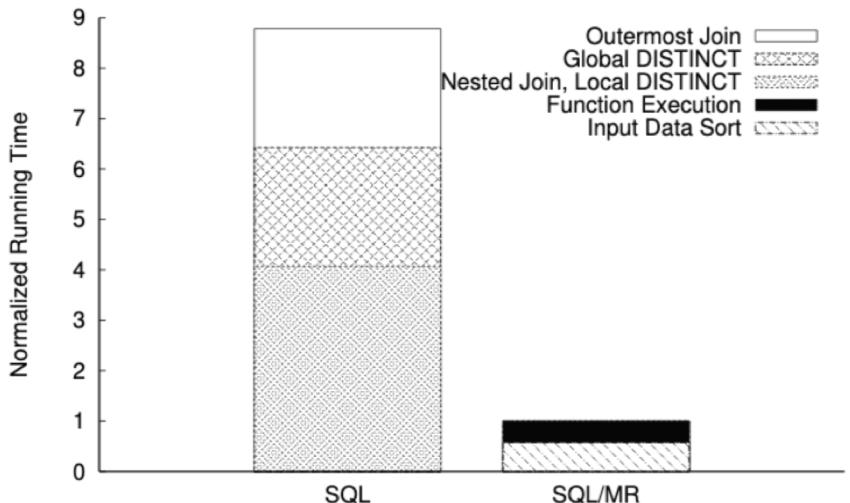
```
fn click_ana(clicks: RowStream<i32, i32, i64>)
  -> GroupedRows<i32, i32> {
  for (uid, group_stream) in group_by(0, clicks) {
    let sequences = IntervalSequence::new();
    for (_, cat, time)
      in sort_on(2, group_stream) {
      if *cat == 1 {
        sequences.open(*time)
      } else if *cat == 2 {
        sequences.close(*time)
      } else {
        sequences.insert(*time)
      }
    };
    (uid,
     sequences.iter()
       .filter(Interval::is_bounded)
       .map(Interval::len)
       .average())
  }
}
```

1. Rakesh Agrawal et al. 2008. The Claremont Report on Database Research. In: SIGMOD Rec. 37.3, 9–19.
2. Charles Welty and David W. Stemple. 1981. Human Factors Comparison of a Procedural and a Nonprocedural Query Language. In: ACM Trans. Database Syst. 626– 649

# Coding it up

Qu  
click  
it t  
fr  
a P

uid  
1  
1  
1



**Figure 12: A comparison of the runtime breakdown of SQL and SQL/MR clickstream analysis queries.**

```
icks: RowStream<i32, i32, i64>
GroupedRows<i32, i32> {
    group_stream) in group_by(0, clicks) {
        sequences = IntervalSequence::new();
        cat, time)
```

Imperative is more efficient because of the many joins in SQL

```
ces.iter()
.filter(Interval::is_bounded)
.map(Interval::len)
.average()
```

- Eric Friedman, Peter Pawlowski, and John Cieslewicz. 2009. SQL/MapReduce: a practical approach to self-describing, polymorphic, and parallelizable user-defined functions. *Proc. VLDB Endow.* 2, 2 (August 2009), 1402-1413.

# Coding it up

**Query:** How many clicks, on average, does it take for a user to get from the **start page** to a **purchase**

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    count(*) - 2 as pageview_count
FROM
    clicks c,
    ( SELECT user_id, max(ts1) as ts1, ts2
      FROM
        ( SELECT DISTINCT ON (c1.user_id, ts1)
```

Noria[2]  
Dataflow system. Uses materialization (state) to improve read performance

Easier:

```
fn click_ana(clicks: RowStream<i32, i32, i64>)
  -> GroupedRows<i32, i32> {
    let stream = clicks
      .group_by(0, clicks) {
```

Dataflow!

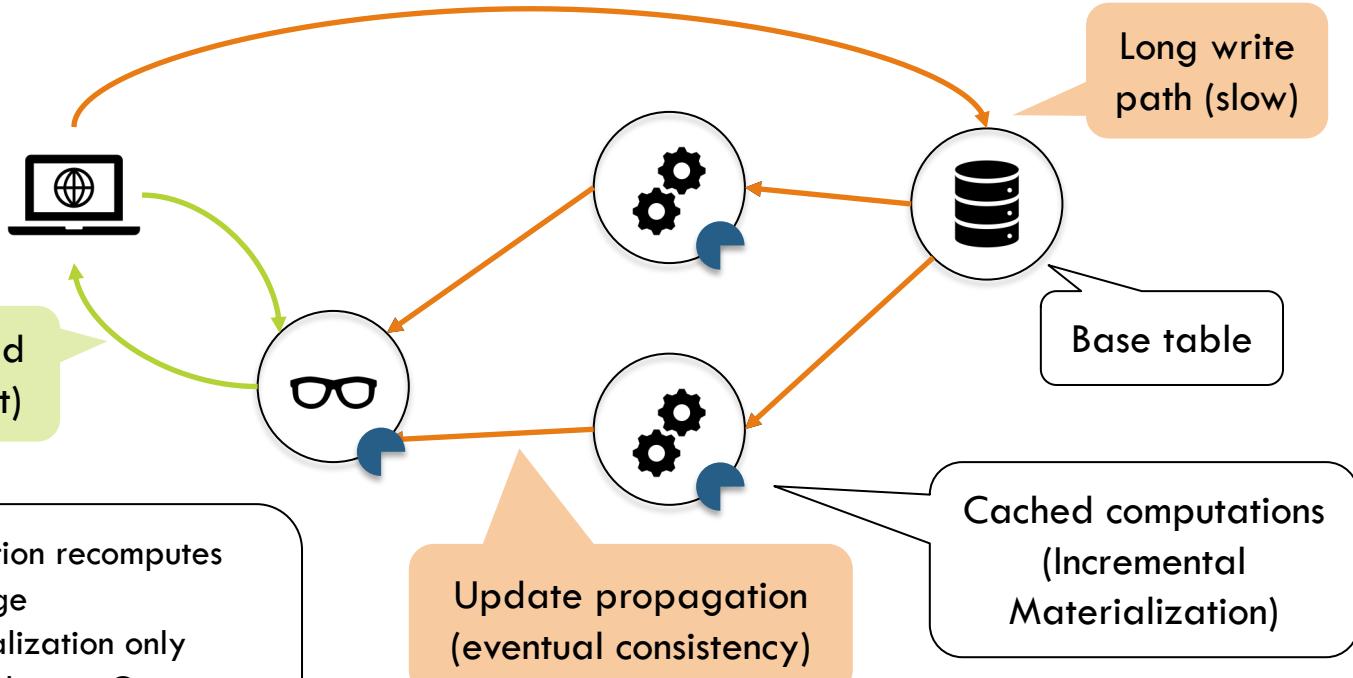
```
    }
```

Ohua[1]

Parallelizable language with a stateful dataflow backend

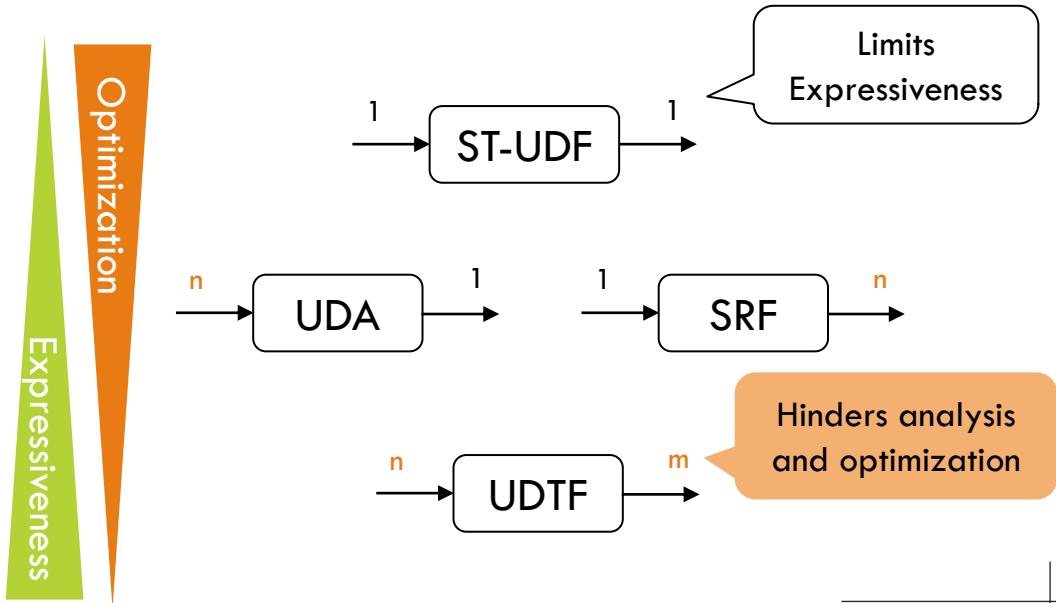
1. Sebastian Ertel, Christof Fetzer, and Pascal Felber. Ohua: Implicit Dataflow Programming for Concurrent Systems. 2015. PPPJ '15. 51–64
2. Jon Gjengset et al. 2018. Noria: dynamic, partially-stateful data-flow for high-performance web applications. In *Proceedings of the 12th USENIX conference on Operating Systems Design and Implementation (OSDI'18)*. USENIX Association, Berkeley, CA, USA, 213–231.

- Multicore
- Distributed
- No UDF Support



- Simple Materialization recomputes everything on change
- Incremental Materialization only recomputes affected rows. Operators must work with changes (Deltas)

# Hierarchy of UDF's



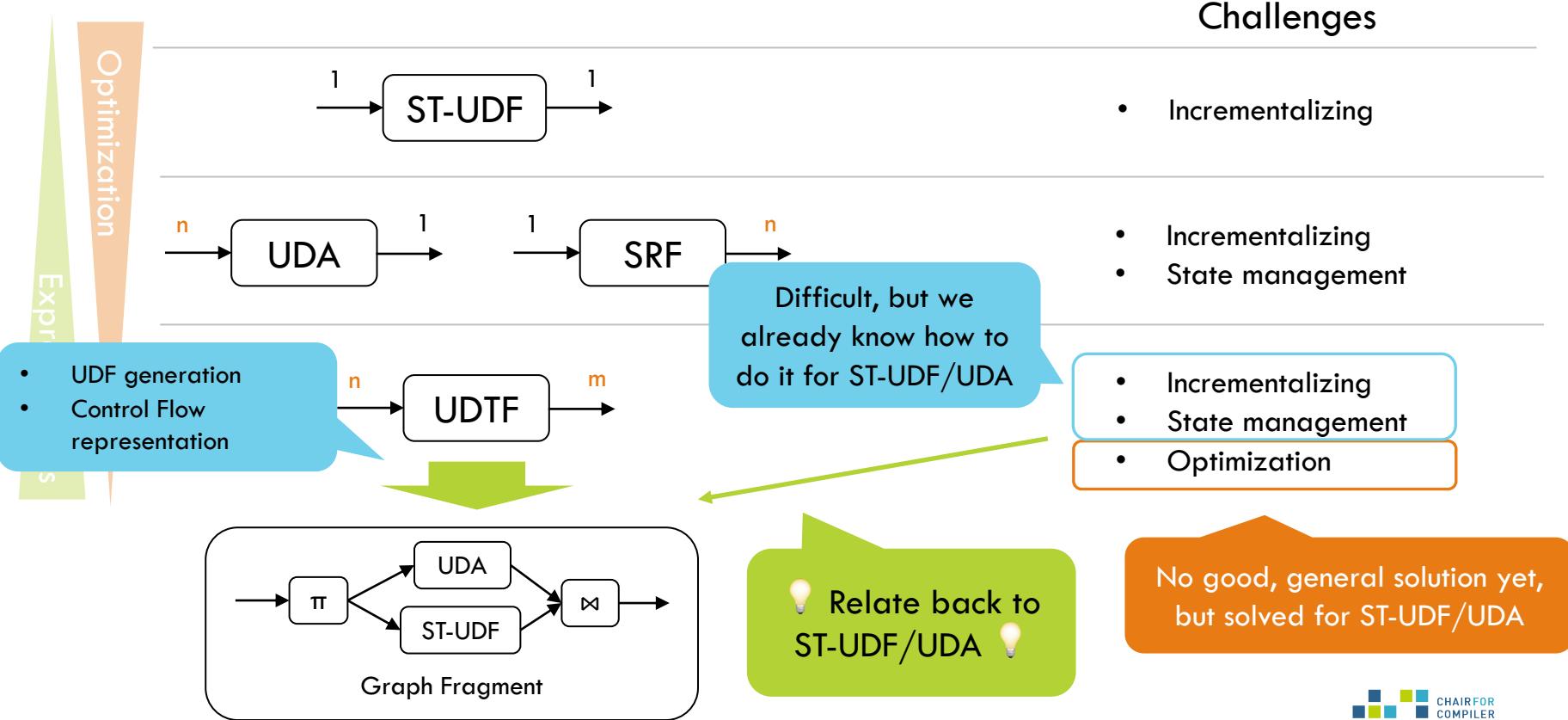
Partial order: Any UDA can be expressed as a UDTF but not vice versa

	UDF	SRF	UDA	UDTF
Postgres	×	×	×	
Apache Hive	×	×	×	×
SQLite	×	×		
MySQL	×		×	
Noria	○	○	○	○

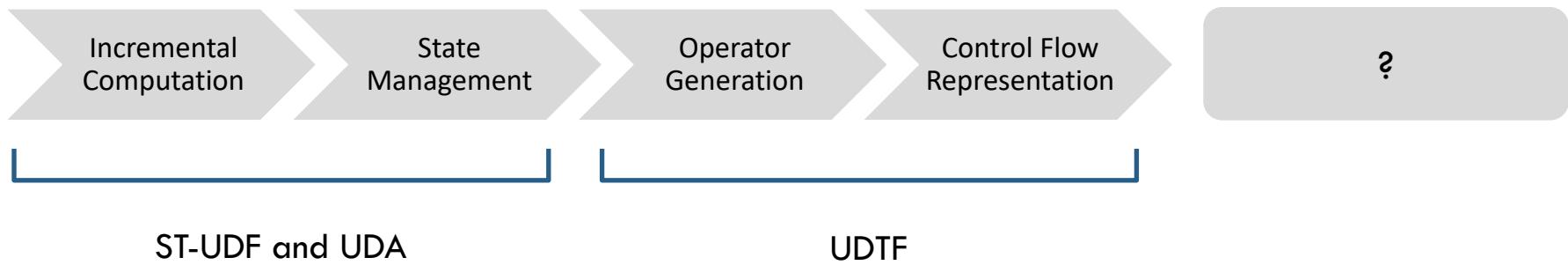
Our target

UDF support in different Databases

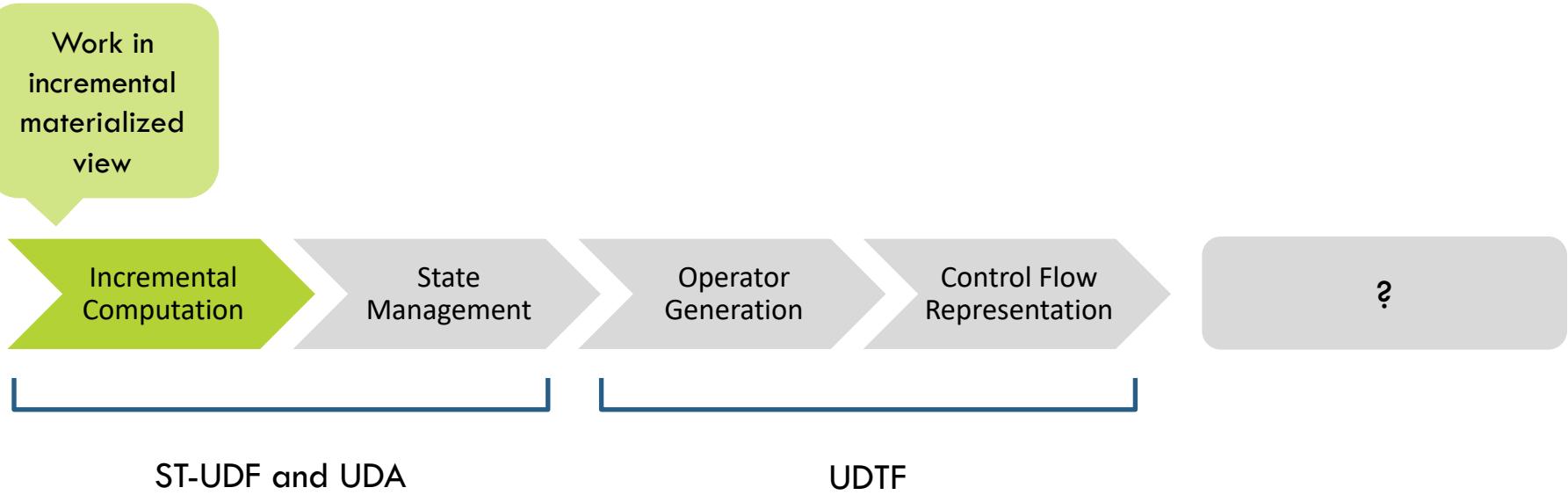
# Hierarchy of UDF's



# Roadmap



# Roadmap



# Incremental computation

## Simple Mat.

- Complete Recompute
- Easy to build
- Inefficient

## Incremental Mat.

- Changes recompute
- Efficient
- Difficult to build
- Represented with inserts and deletes

Operators must recompute all affected previous results (requires tracking state) and issue updates downstream.

Only state needs to be incremental

## ST-UDF

Relatively easy, propagate whether input was update or delete to the output.

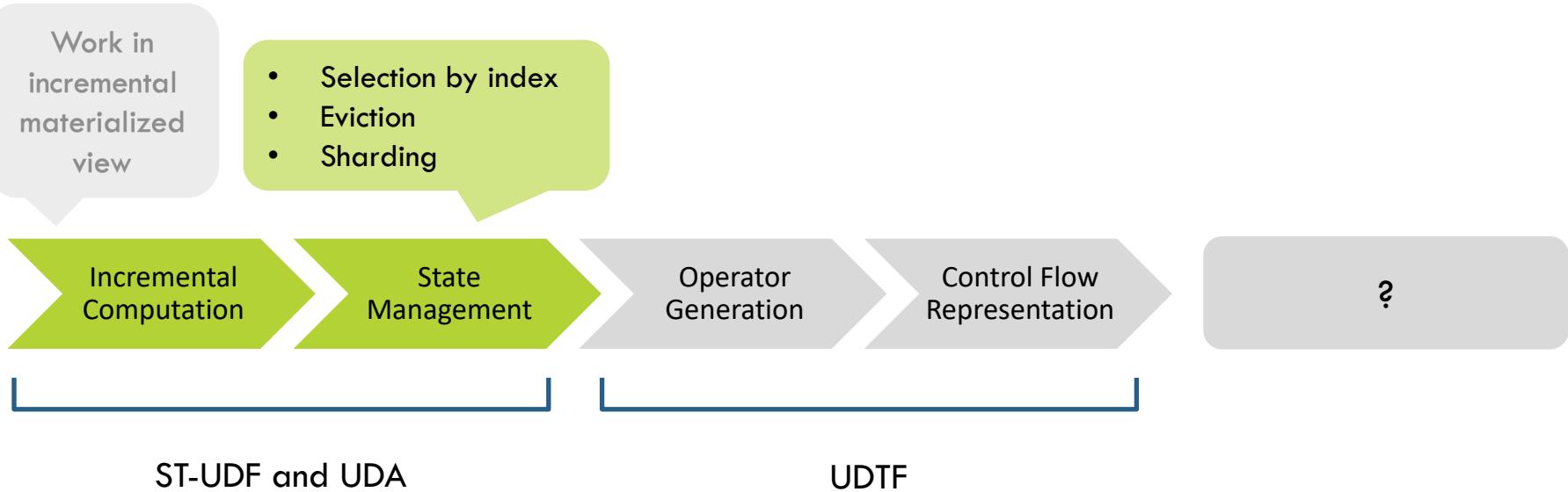
(Same for SRF)

- Only one, known affected previous result
- State determines new value
- Must reverse changes to state

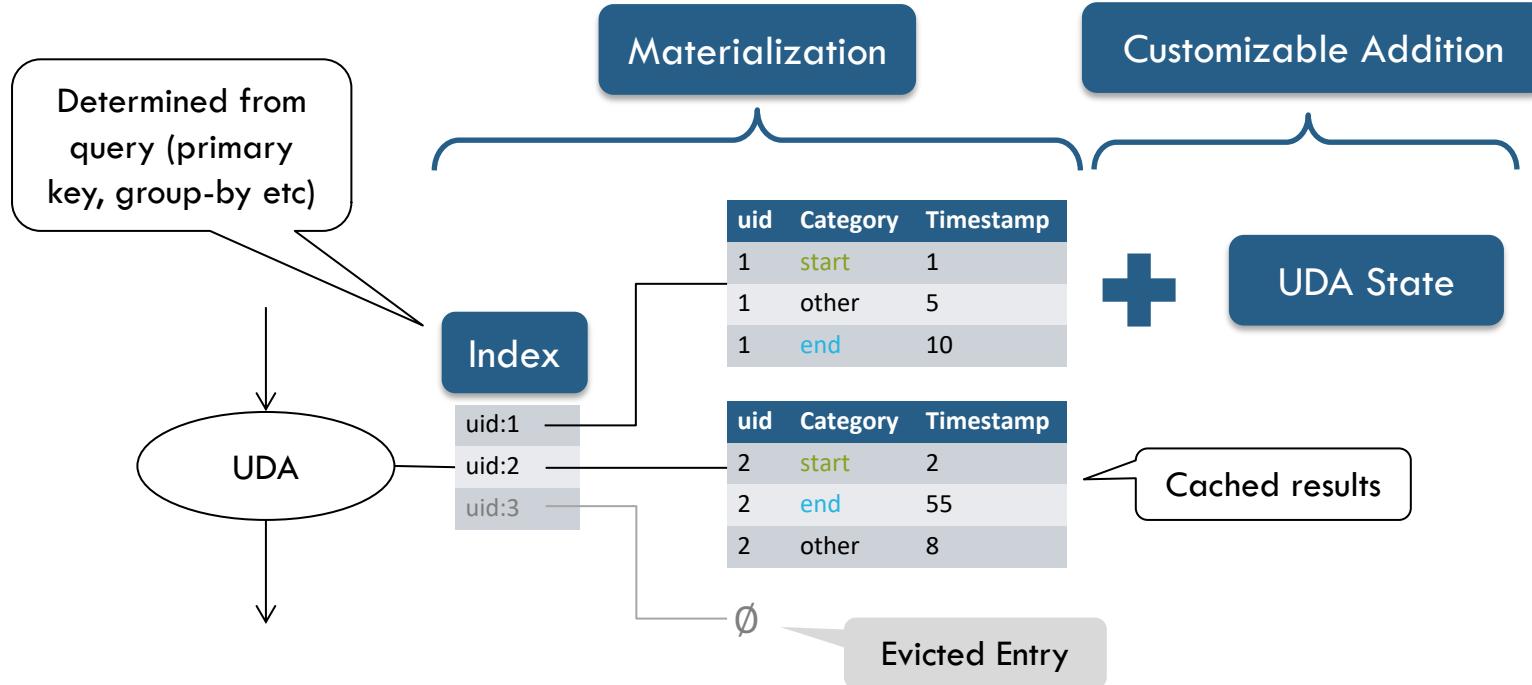
## UDA

```
trait State {  
    type Action;  
    type Output;  
    fn apply(&mut self,  
            action: Self::Action);  
    fn reverse(&mut self,  
              action: Self::Action);  
    fn compute(&self) -> Self::Output;  
}
```

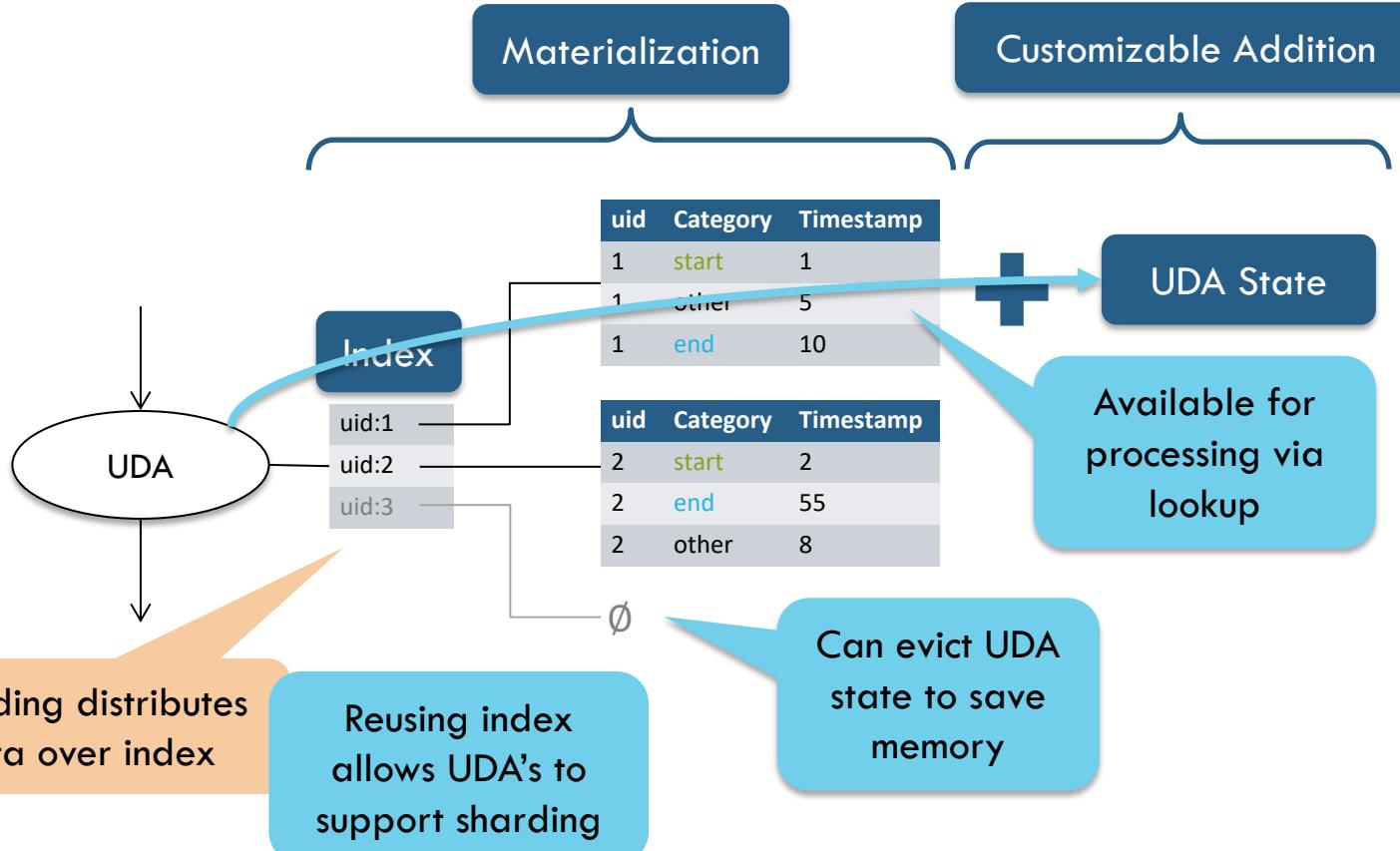
# Roadmap



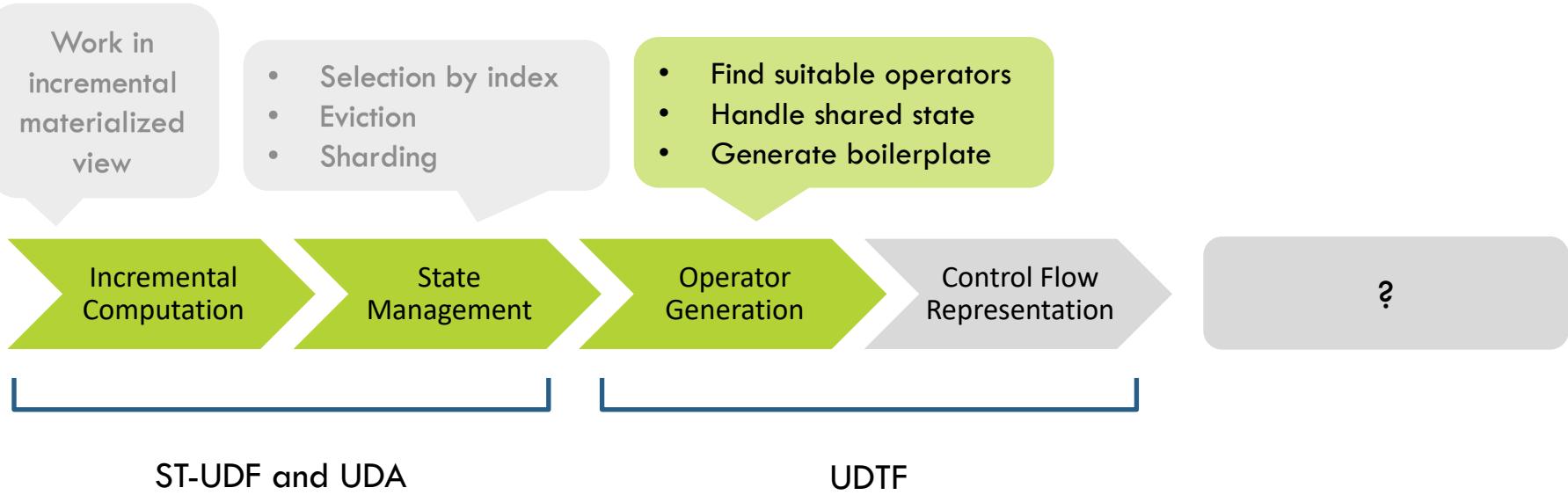
# UDA State Management



# UDA State Management



# Roadmap



# Operator Generation

- Shared state means synchronization
- Complicates or prevents parallelism
- Not supported in Noria

 Make minimal operator with local state

```
fn click_ana(clicks: RowStream<i32, i32, i64>)
    -> GroupedRows<i32, i32> {
    for (uid, group_stream) in group_by(0, clicks) {
        let sequences = IntervalSequence::new();
        for (_, cat, time) in sort_on(2, group_stream) {
            if *cat == 1 {
                sequences.open(*time);
            } else if *cat == 2 {
                sequences.close(*time);
            } else {
                sequences.insert(*time);
            }
        };
        (uid,
         sequences.iter()
             .filter(Interval::is_bounded)
             .map(Interval::len)
             .average())
    }
}
```

2. Select all state uses

1. Select init expression

4. Bundle into operator

3. Recursively select dependencies

Rest of program

5. Add boilerplate appropriate for type of generated UDF (not shown)

Only operator local state left

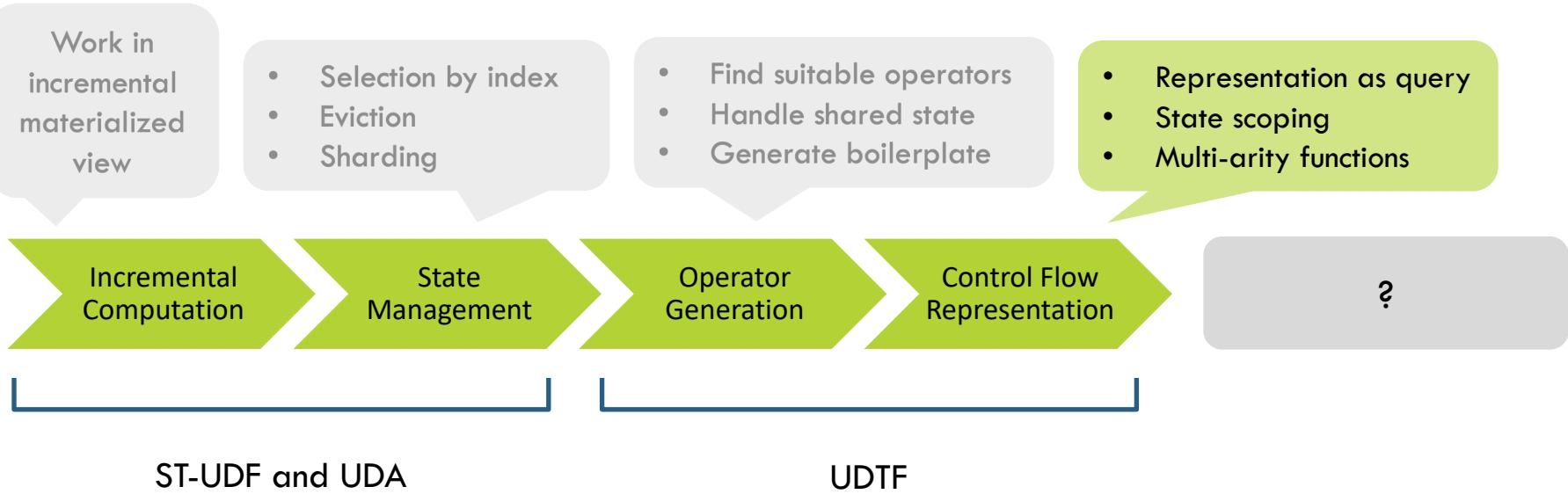
```
impl Op0 {
    fn init() -> Self {
        Op0(IntervalSequence::new())
    }
    fn run(&mut self, rows: Rows<(i32, i32, i64)>)
        -> f64 {
        for (_, cat, time) in rows { ... }
        self.0.iter()
            .filter(Interval::is_bounded)
            .map(Interval::len)
            .average()
    }
}
```

Operator Core (Rust)

```
fn click_ana(clicks: RowStream<i32, i32, i64>)
    -> GroupedRows<i32, i32> {
    for (uid, group_stream) in group_by(0, clicks) {
        let op0 = Op0::init();
        let op0_res = op0.run(sort_on(2, group_stream));
        (uid, op0_res)
    }
}
```

UDTF (Ohua)

# Roadmap



# Iteration Representation

Operators always work on batches for efficiency

- No special iteration operator needed

**State must respect scope**  
State value only valid for one iteration

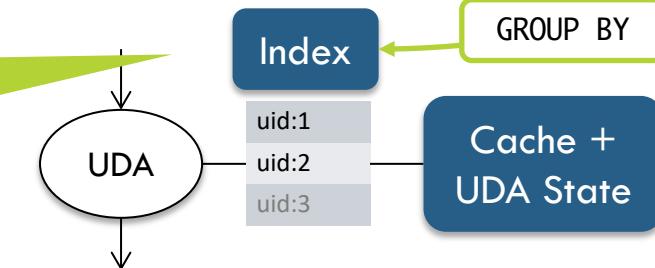
```
fn click_ana(clicks: RowStream<i32, i32, i64>)
    -> GroupedRows<i32, i32> {
    for (uid, group_stream) in group_by(0, clicks) {
        let op0 = Op0::init();
        let op0_res = op0.run(sort_on(2, group_stream));
        (uid, op0_res)
    }
}
```

Number of iterations not known.  
Incremental execution revisits state.  
→ Cannot just duplicate operator  
→ State index & dispatch needed

Sequence source provides index  
Found by analysing control flow context

- Sequence never created
- Source streams items
- Each row tagged with index

UDA State already indexed  
[12. Slide](#)



Nesting achieved via compound indices

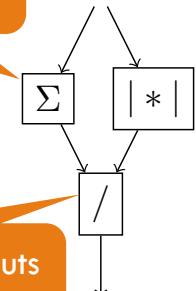
# Multi argument functions

```
fn average(      : RowStream<...>) {  
    div(sum(elems), count(elems))  
}
```

Order of output tuples  
cannot be guaranteed

Only interesting for  
multiple outputs i.e.  
iteration

Needs to line up inputs  
from same iteration



There already exists an  
operator that does this

⋈ (join)

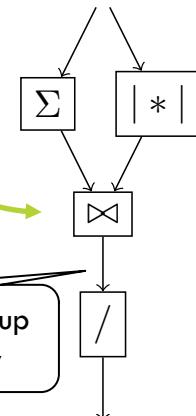
Needs an key to join  
iterations on

Scope key from  
before also  
associates iterations

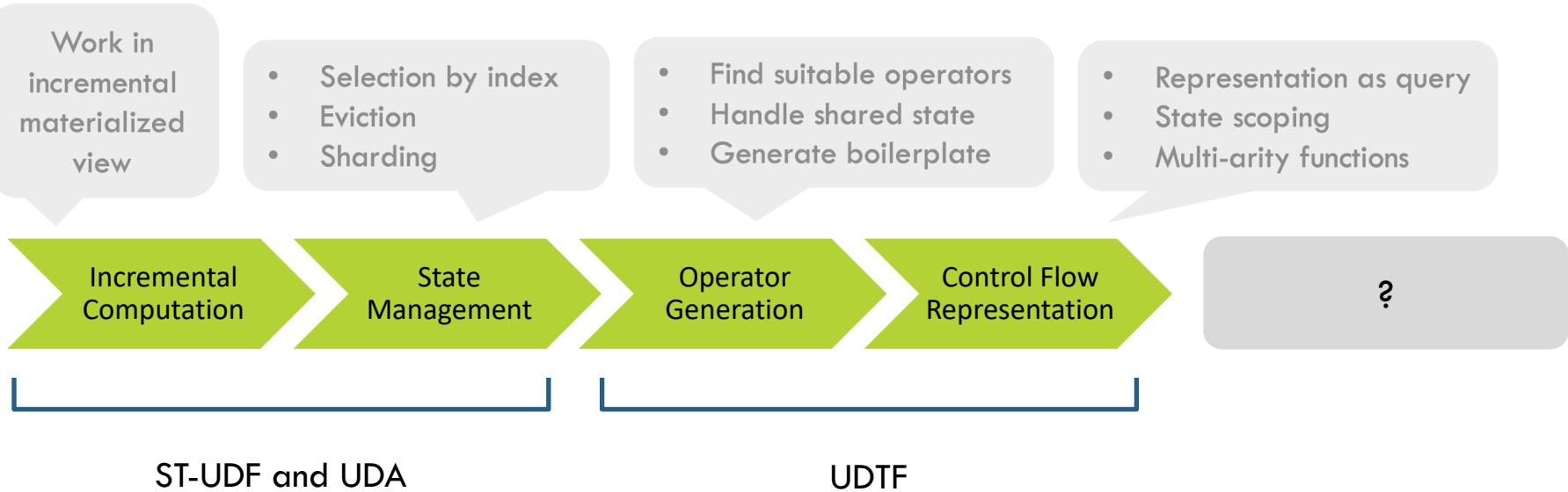
GROUP BY

All inputs packaged up  
nicely in single row

Also works correctly for variables  
from outside the for-loop



# Roadmap



ST-UDF and UDA

UDTF

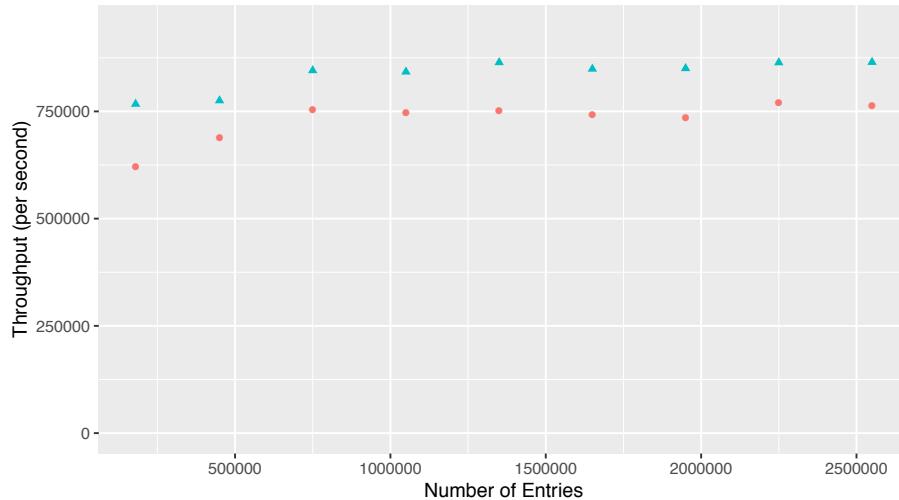
## Evaluation

- Interoperability with SQL
- Composition/Control Flow
- Optimization (Parallelization)

# Evaluation – Overhead & Expressiveness

```
fn average(table: RowStream<...>) {  
    for (_, elems) in group_by(0, table) {  
        div(sum(elems), count(elems))  
    }  
}
```

Multi-argument functions and inner-joins naturally correspond



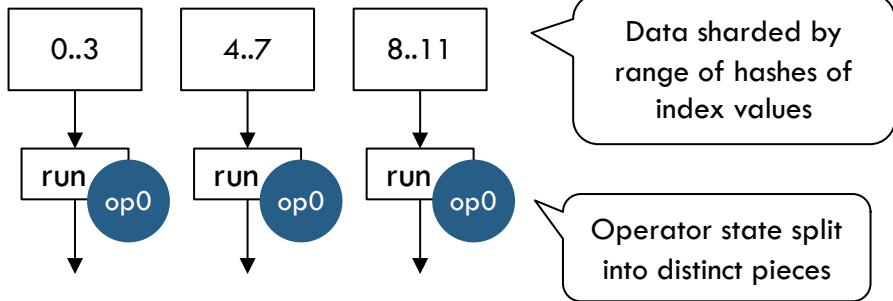
Performance difference in query due to extra operators inserted by compiler

Lang  
ohua  
sql

Separate performance comparison of generated sum and SQL sum operators shows no difference

Performance of Ohua-compiled average query in comparison to SQL

# Evaluation - Parallelism



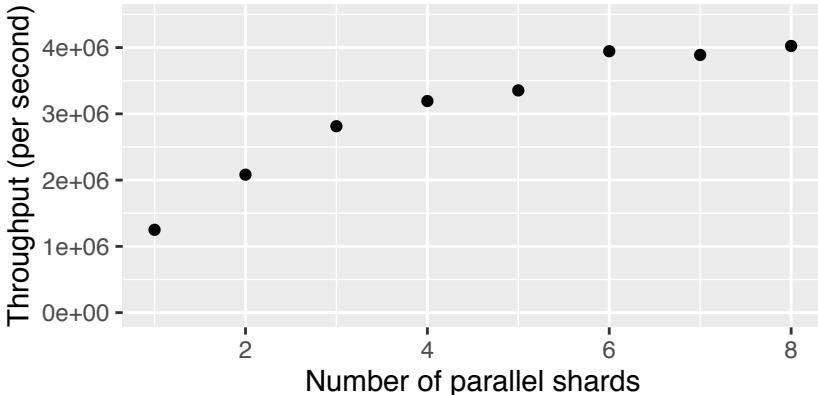
```
fn click_ana(clicks: RowStream<i32, i32, i64>)
    -> GroupedRows<i32, i32> {
    for (uid, group_stream) in group_by(0, clicks) {
        let op0 = Op0::init();
        let op0_res = op0.run(sort_on(2, group_stream));
        (uid, op0_res)
    }
}
```

Iteration local state  
allows splitting

Leveraging the parallelism is simply  
setting a runtime parameter

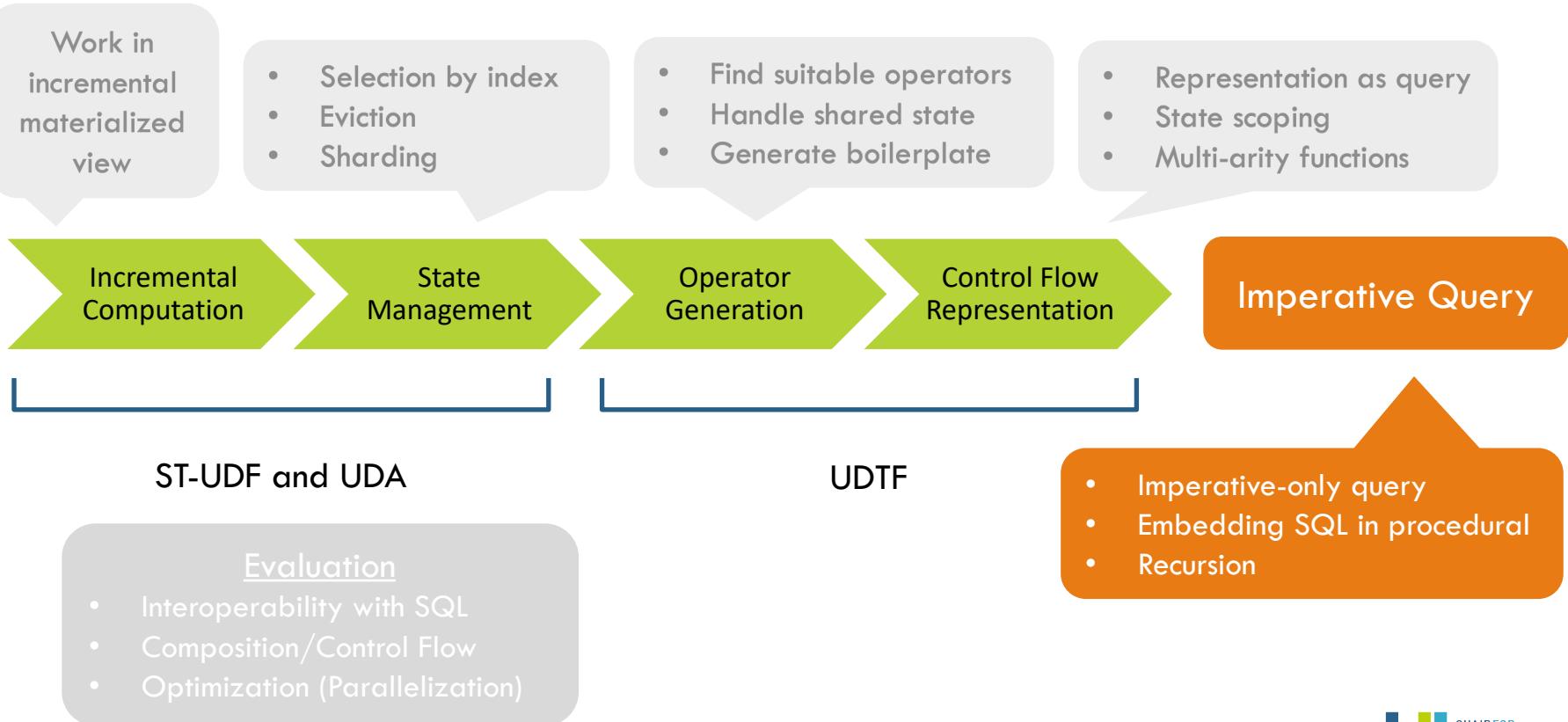
```
let sharding_factor = 8;
let mut b = Builder::default();
b.set_sharding(Some(sharding_factor));
```

Parallel processing possible without  
explicit parallel contracts



Throughput of clickstream analysis with  
increasing sharding factor

# Roadmap



# Outlook – Embedding SQL

```
SELECT udf(x,y)  
FROM tab  
WHERE udf2(r,q)
```

With common dataflow base we can also embed SQL in imperative program

```
let dat = run_query("SELECT...", x);  
for i in dat {  
    run_query("INSERT...", i);  
}  
run_query("DELETE...", x, y, z);
```

Initial goal: Embedding Imperative in SQL

SQL

Imperative

SQL compiles to dataflow

Dataflow

With Ohua, dataflow becomes common base

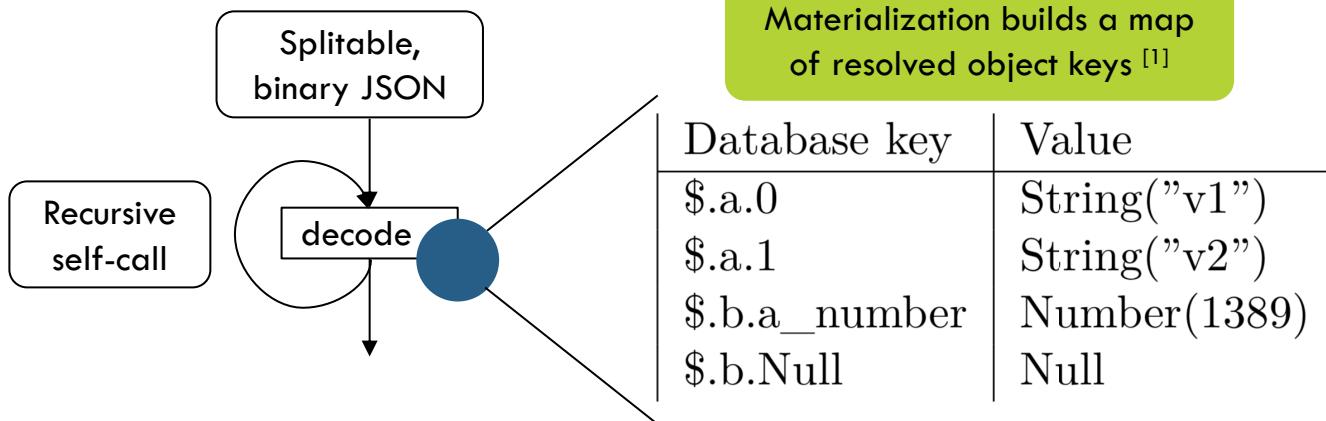
Created query dataflow representation for procedural programs

SQL involvement not necessary: Procedural-only query is possible

# Outlook - Recursion

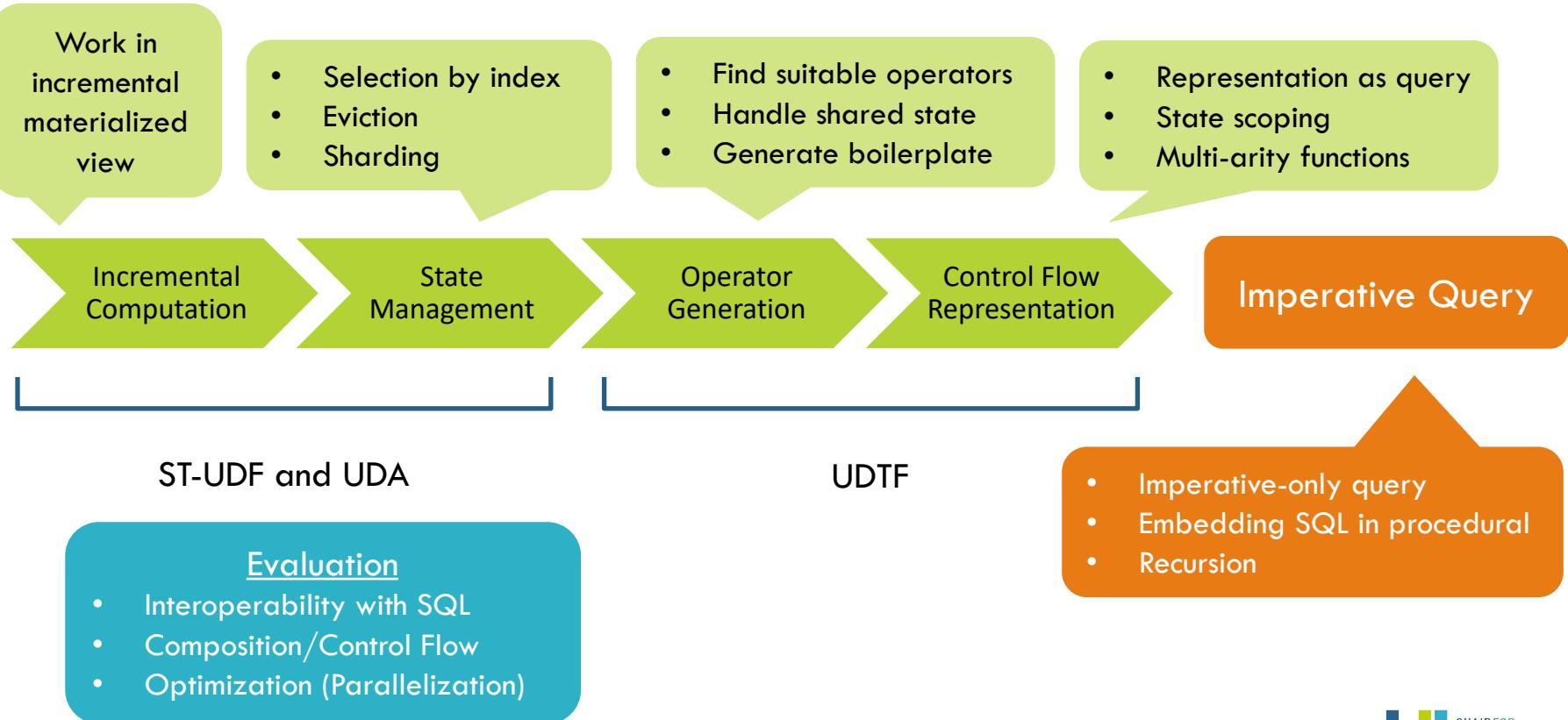
```
{  
    "a": ["v1", "v2"],  
    "b": {  
        "a_number": 1389,  
        "Null": null  
    }  
}
```

Arbitrary nesting needs  
recursive decoding of  
inner structure



1. Zhen Hua Liu et al. Closing the Functional and Performance Gap Between SQL and NoSQL. 2016. SIGMOD '16. 227–238.

# Roadmap



# Simple Materialization

## Materialization

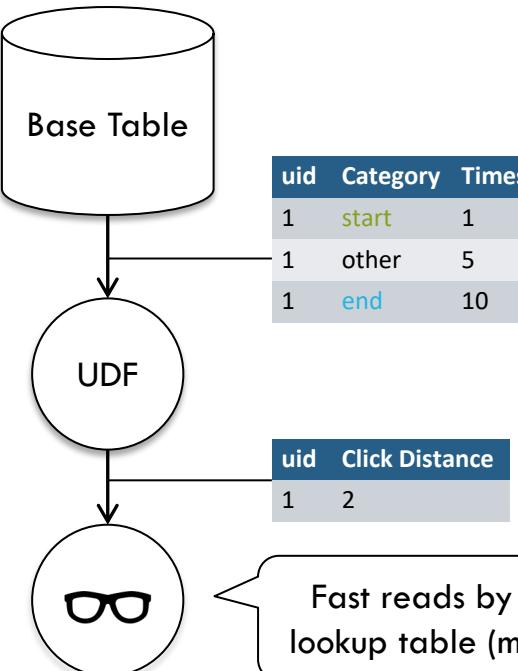
	uid	Category	Timestamp
inserted	1	start	1
	1	other	5
	1	end	10
deleted	1	other	11

Does no processing, hence same materialization as upstream

uid	Click Distance
1	2

## Data Transferred

Entire Table transferred and processed. Inefficient and with high latency



Fast reads by serving from lookup table (materialization)

# Incremental Materialization

## Materialization

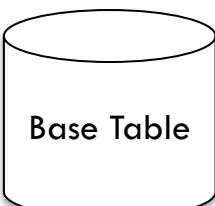
uid	Category	Timestamp
1	start	1
1	other	5
1	end	10
1	other	11

Private materialization  
as lookup table for  
downstream operators

uid	Click Distance
1	2

uid	Click Distance
1	2

## Data Transferred

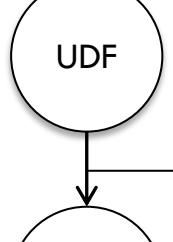


Base Table

Sign added  
to each row

Only deltas transferred  
and processed

sign	uid	Category	Timestamp
+	1	other	5
-	1	other	11



UDF



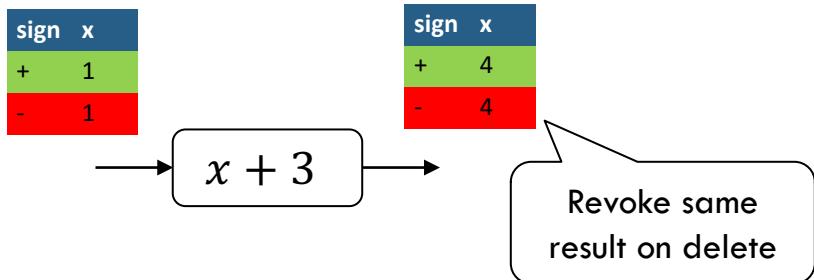
sign	uid	Click Distance
-	1	1
+	1	2

Operator must be  
able to adjust the  
result on delete

Output are deltas and  
delete outdated results

# Incremental ST-UDF and UDA

ST-UDF

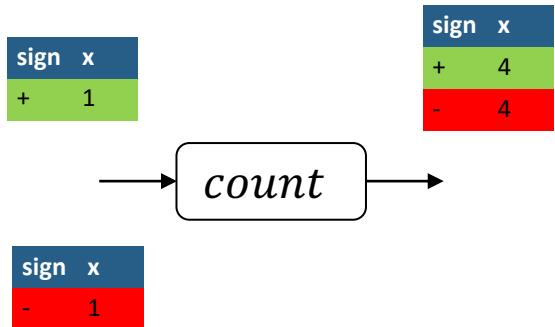


For a 1:1 function  $f(x)$  the incremental function  $f'$  is:

$$f'(+, x) = (+, f(x))$$

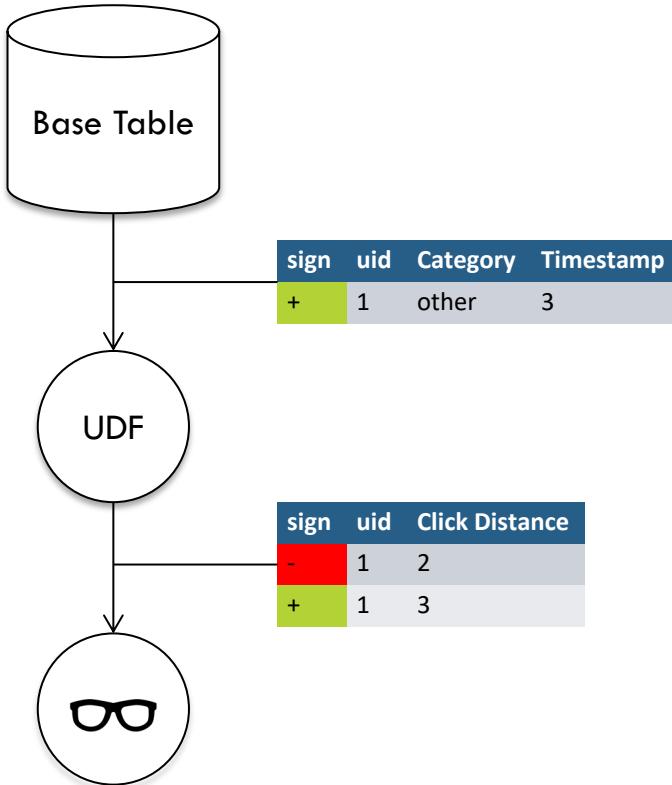
$$f'(-, x) = (-, f(x))$$

UDA



# Simple materialization

uid	Category	Timestamp
1	start	1
1	other	5
1	end	10
1	other	3

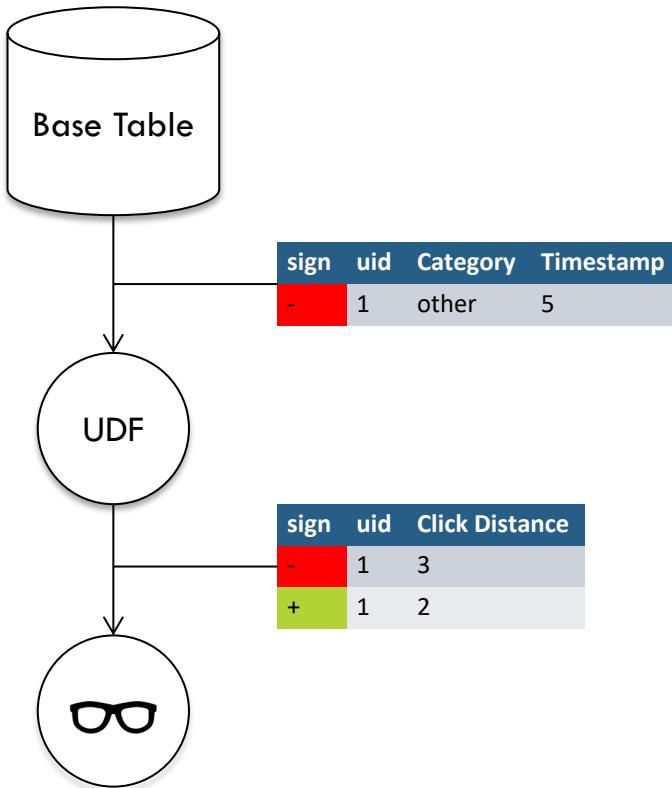


Update Path (Insert)

INSERT (1, other, 3)  
INTO 'Base Table';

# Noria Execution Model

uid	Category	Timestamp
1	start	1
1	other	5
1	end	10
1	other	3

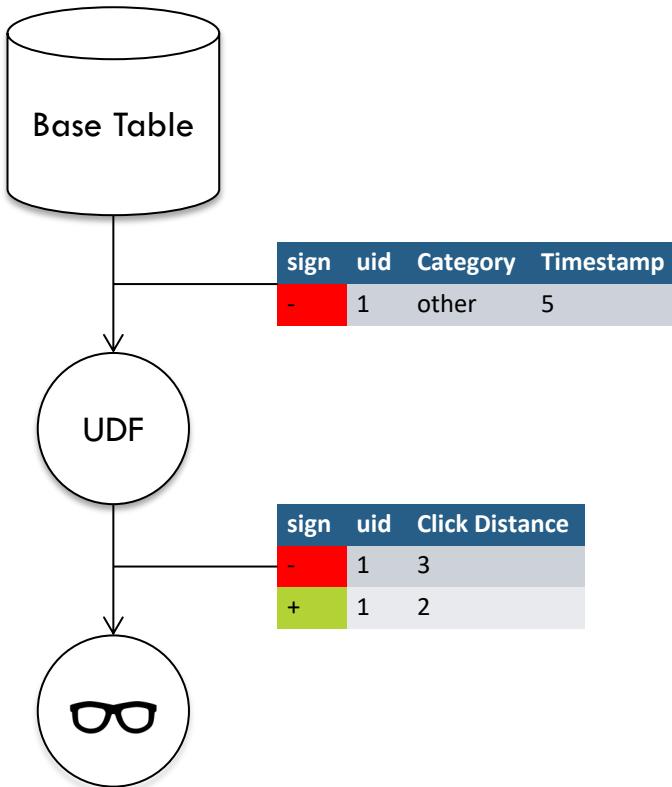


Update Path (Delete)

DELETE (1, other, 3)  
FROM ‘Base Table’;

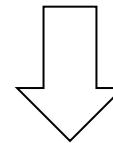
# Noria Execution Model

uid	Category	Timestamp
1	start	1
1	other	5
1	end	10
1	other	3



uid	Click Distance
1	2

- On-line inserts
- On-line deletes
- Order is random



- Commutative
- Incremental
- Reversible Operations

# UDF State Design

```
let state = iseq::Seq::new();  
  
for (_, cat, time) in stream {  
  
    if cat == start_cat {  
        state.start(time)  
    } else if cat == end_cat {  
        state.end(time)  
    } else {  
        state.record(time)  
    }  
}
```

```
state  
.complete_intervals()  
.map(Interval::len)  
.average()
```

State  $S$

Defines actions  
 $A: \{\text{Start, End, Record}\}$

Projection  
 $f: input \rightarrow A$

Not affected by sign

Computation  
 $comp: S \rightarrow output$

Affected by sign

Successively apply all  
actions and sign to state  
 $app: \pm A \times S \rightarrow S$

$UDF: [\pm input] \rightarrow output$

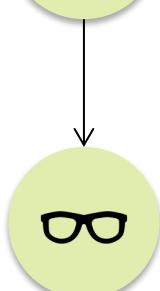
# Interval Sequence as State



uid	Category	Timestamp
1	start	1
1	other	5
1	end	10
1	other	3



`[[5,3,10].length()].average() == 3`



uid	Click Distance
1	3

$$s: [ [l_0, u_0), [l_1, u_1), [l_2, u_2)] ]$$

$t \in T$  such that

- $t \geq \begin{cases} l_1 & \text{if } l_1 \text{ exists} \\ u_0 & \text{otherwise} \end{cases}$
- $t < \begin{cases} u_1 & \text{if } u_1 \text{ exists} \\ l_2 & \text{otherwise} \end{cases}$

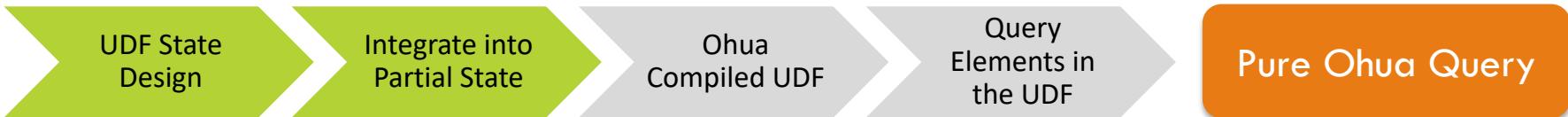
Invariants

- $l_1$  or  $u_0$  must exist
- $u_1$  or  $l_2$  must exist

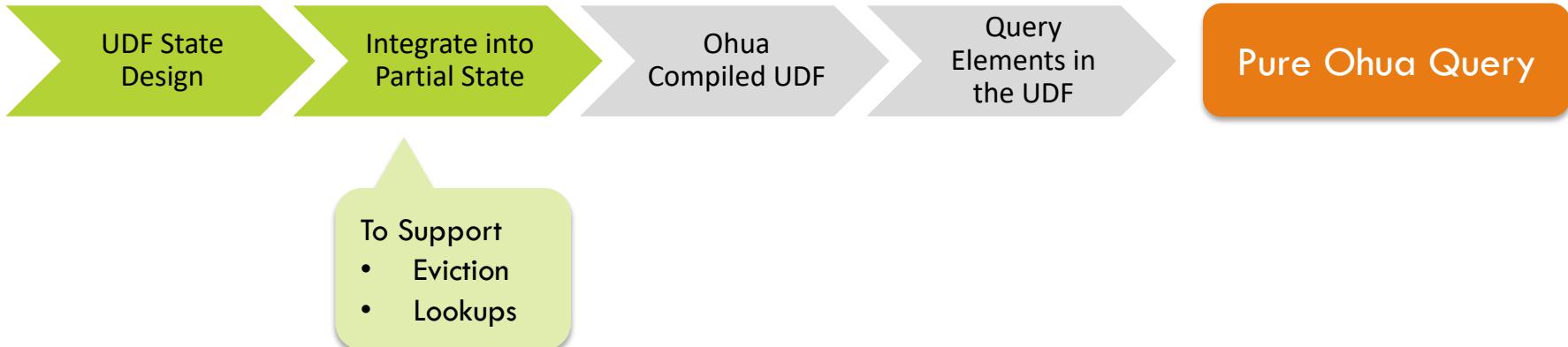
→ Merge intervals to maintain

# Conclusions

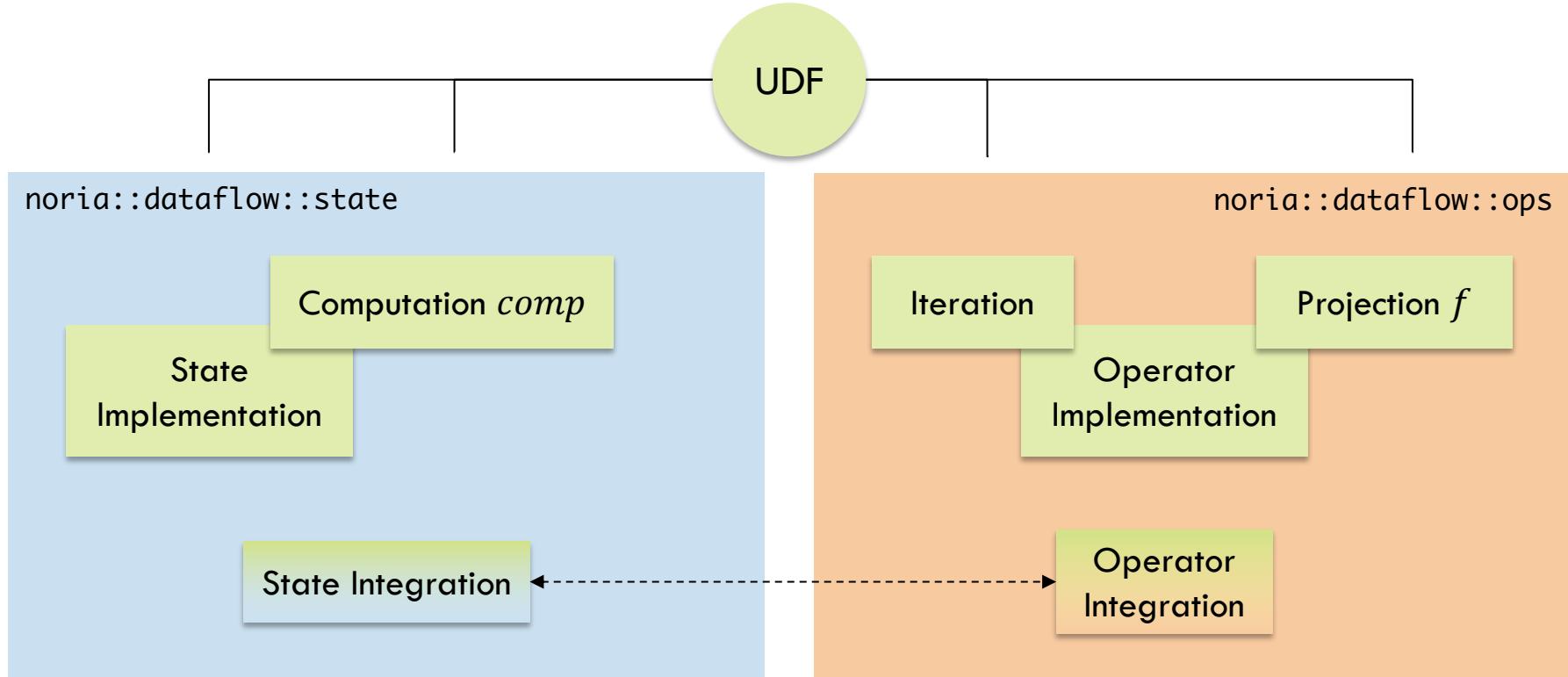
- Must be
- Reversible
  - Commutative



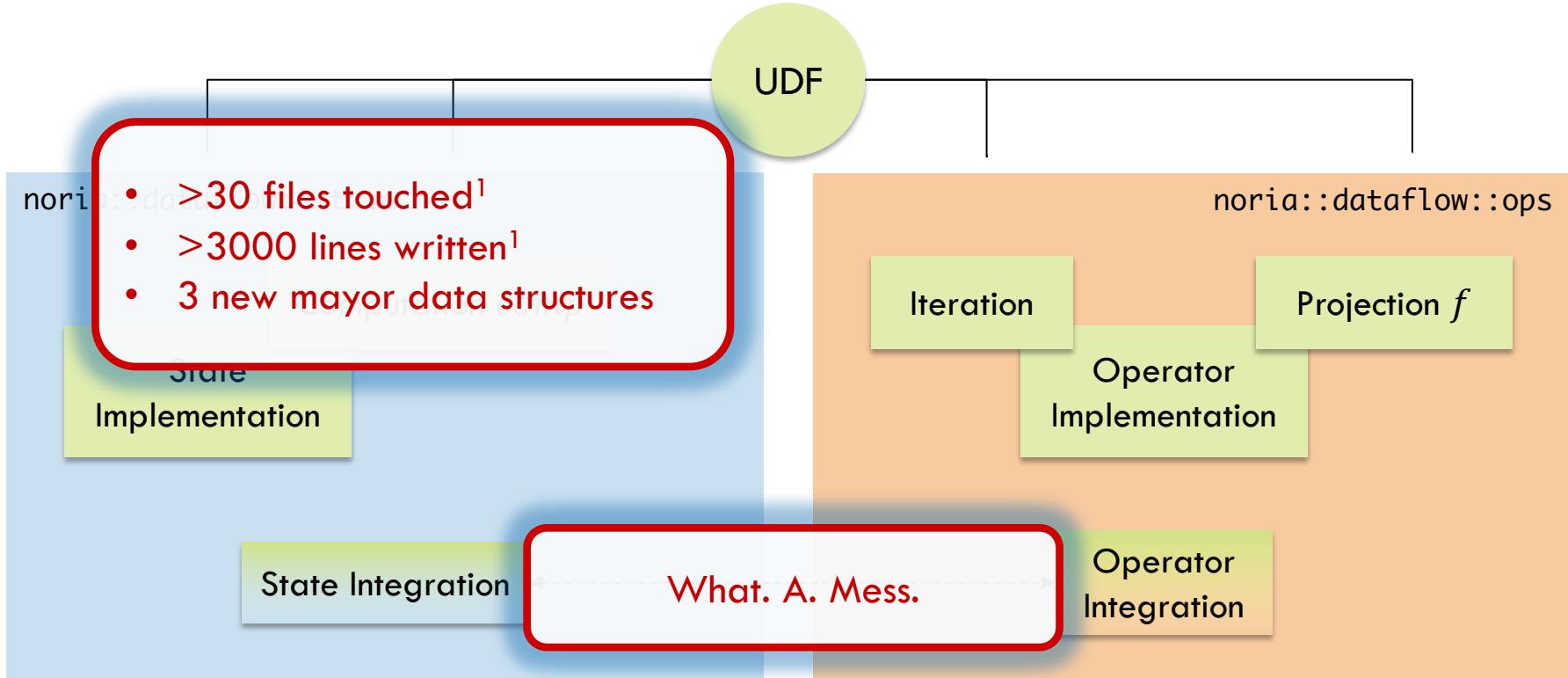
# Conclusions



# Manual Implementation

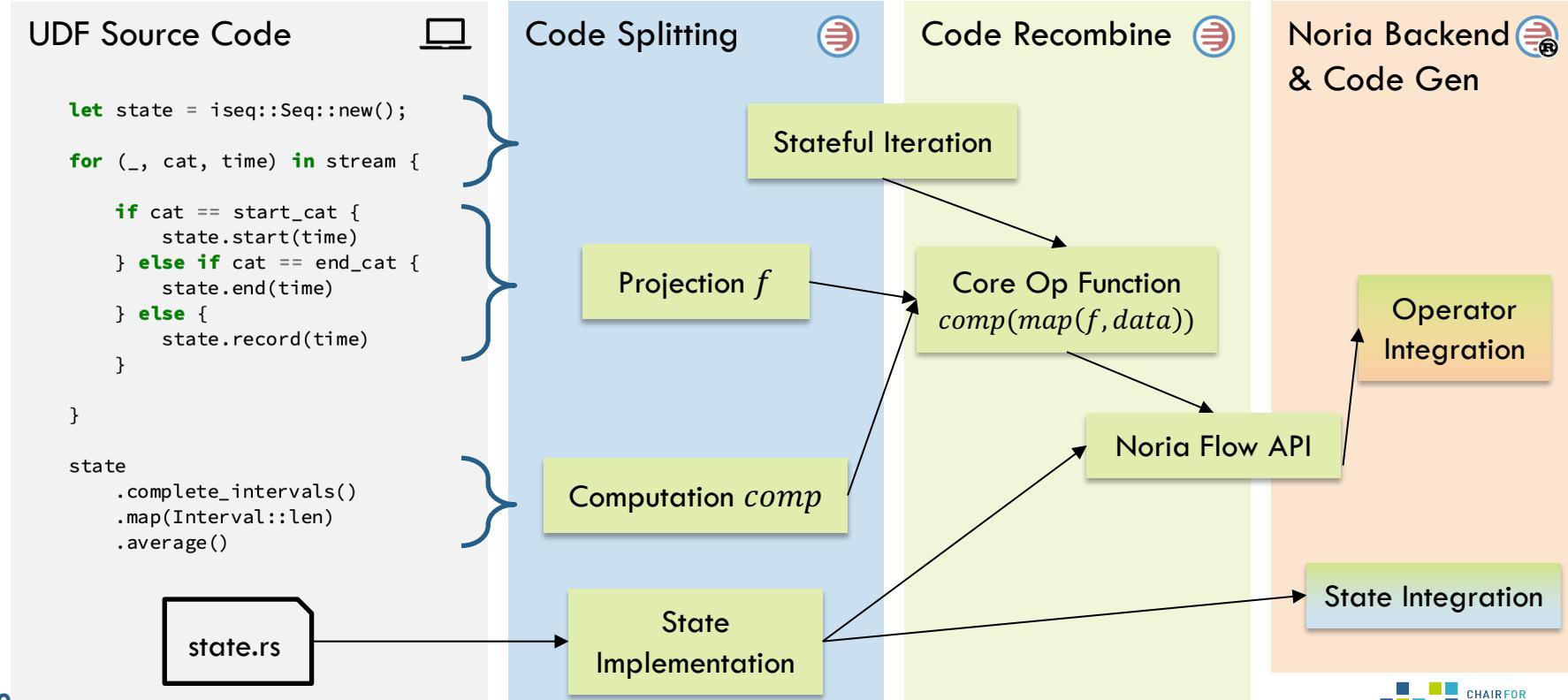


# Manual Implementation

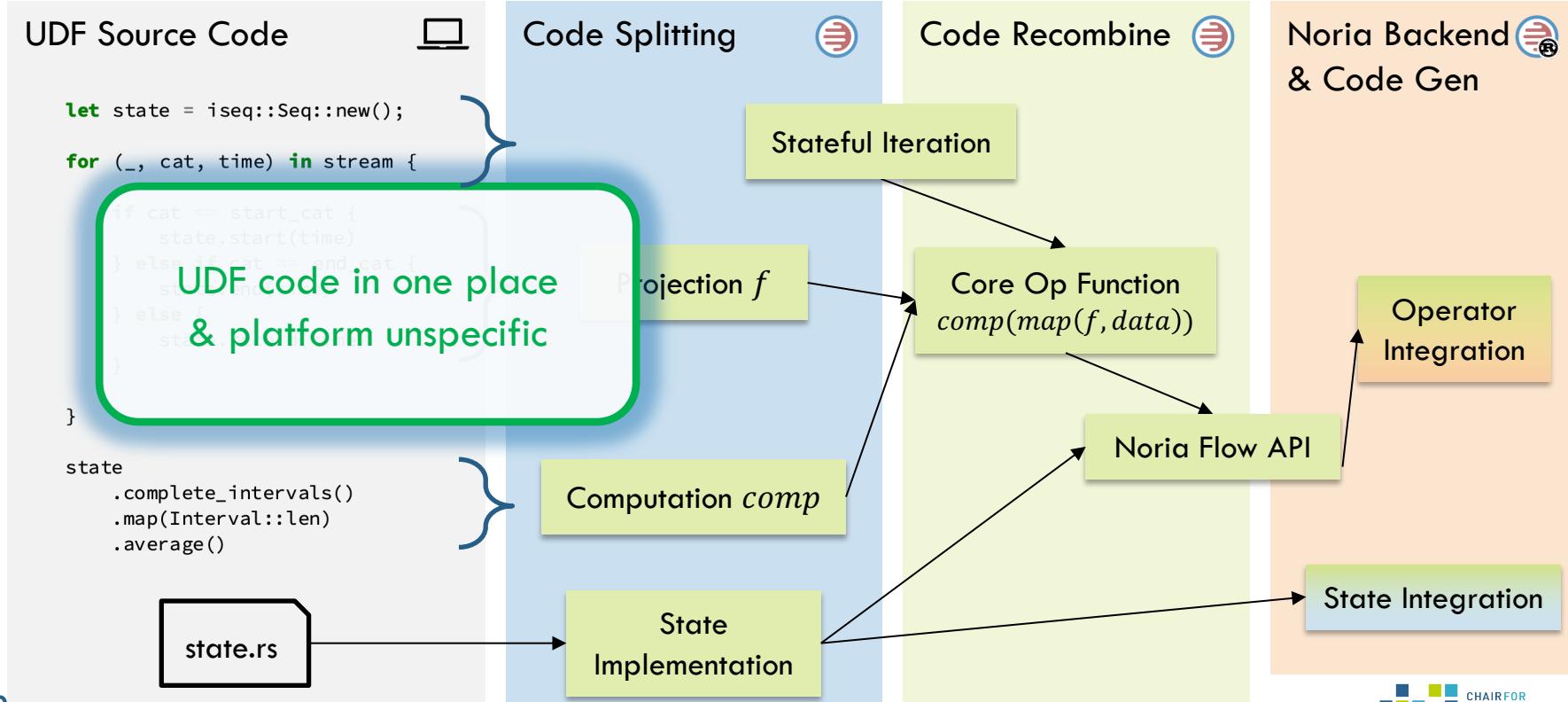


1. For whole implementation including intermediate prototypes and test code. Approximately 50% used exclusively for UDF.

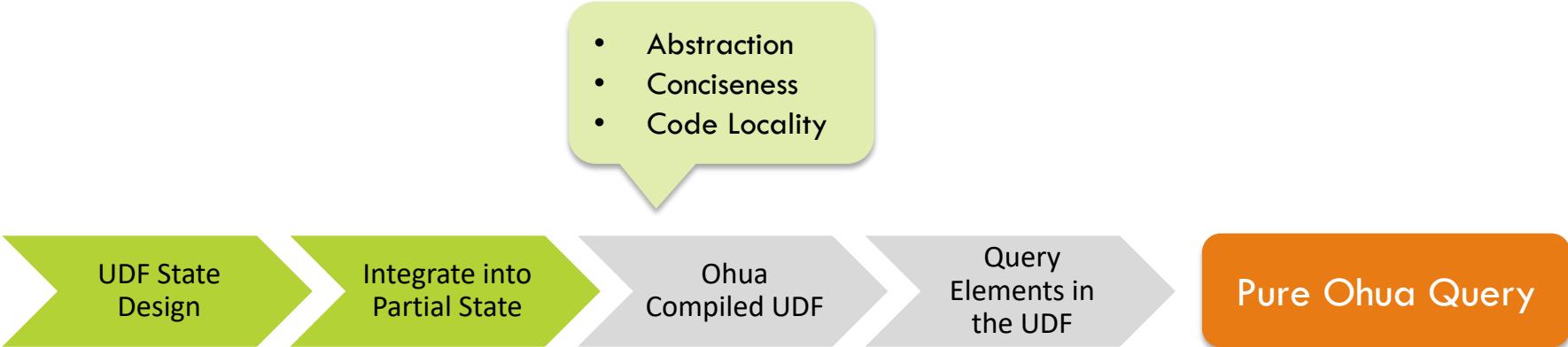
# Operator Compilation



# Operator Compilation



# Conclusions



# UDF Compilation

## UDF Source Code



```
fn click_ana(  
    start_cat: Category,  
    end_cat: Category,  
    clicks: Stream<(UID, Category, Time)>  
) -> f64 {  
  
    let click_streams = group_by::<0>(clicks);  
  
    click_streams.map(|stream| {  
  
        ...  
  
        Operator Code  
    })  
}
```

## Code Splitting



Signature

Grouping

Operator

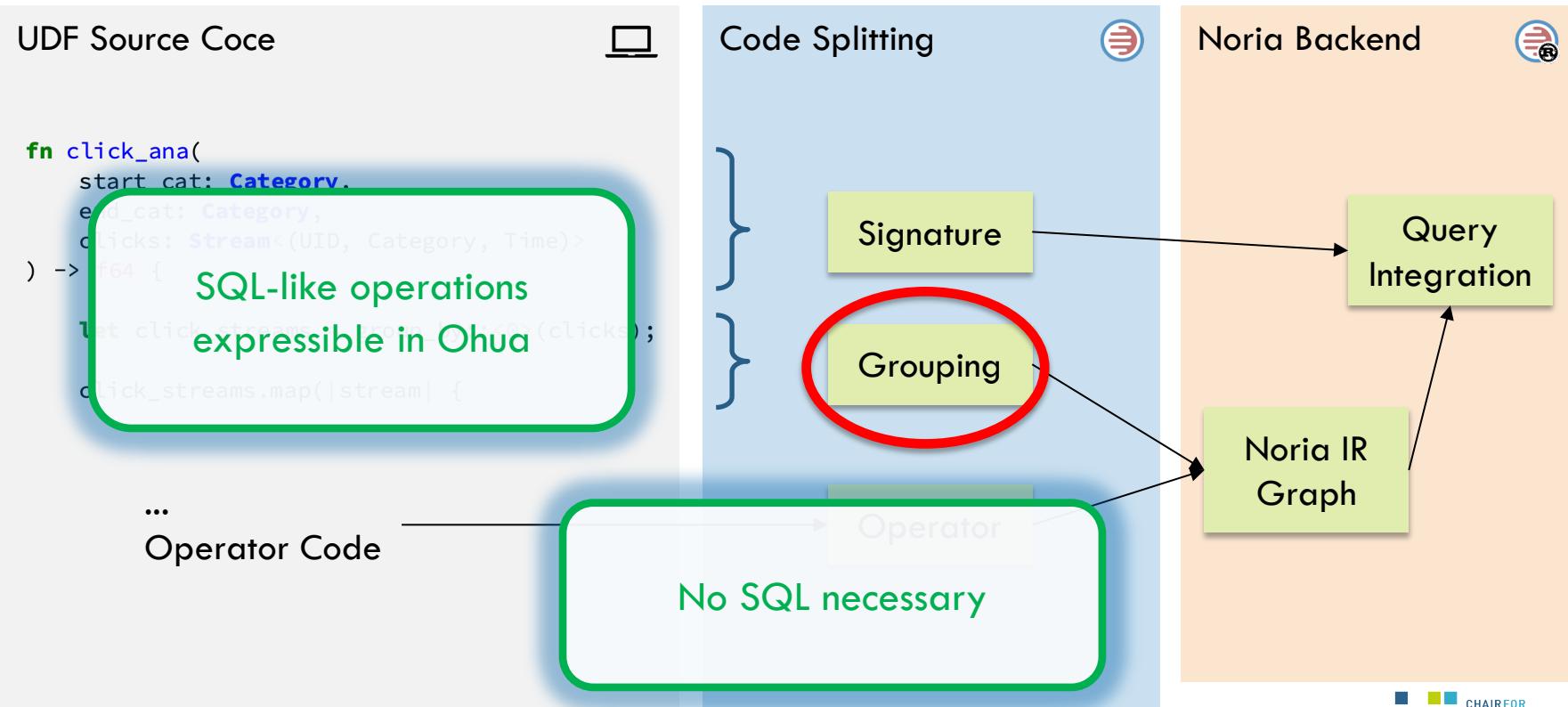
## Noria Backend



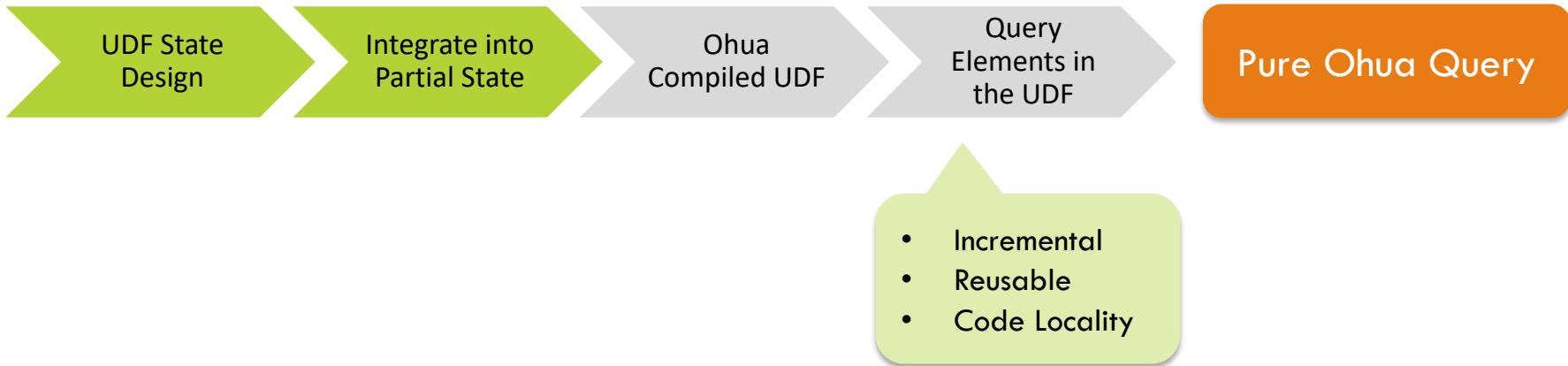
Query  
Integration

Noria IR  
Graph

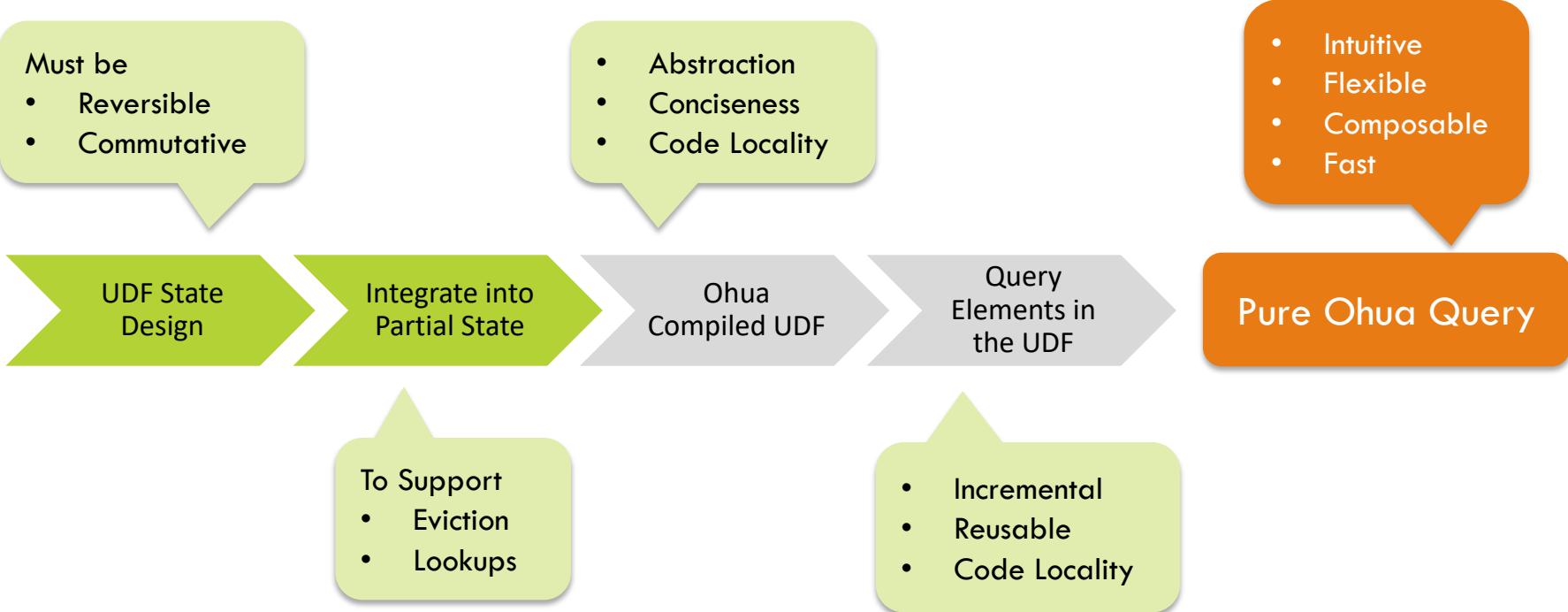
# UDF Compilation



# Conclusions



# Conclusions



# Conclusions

