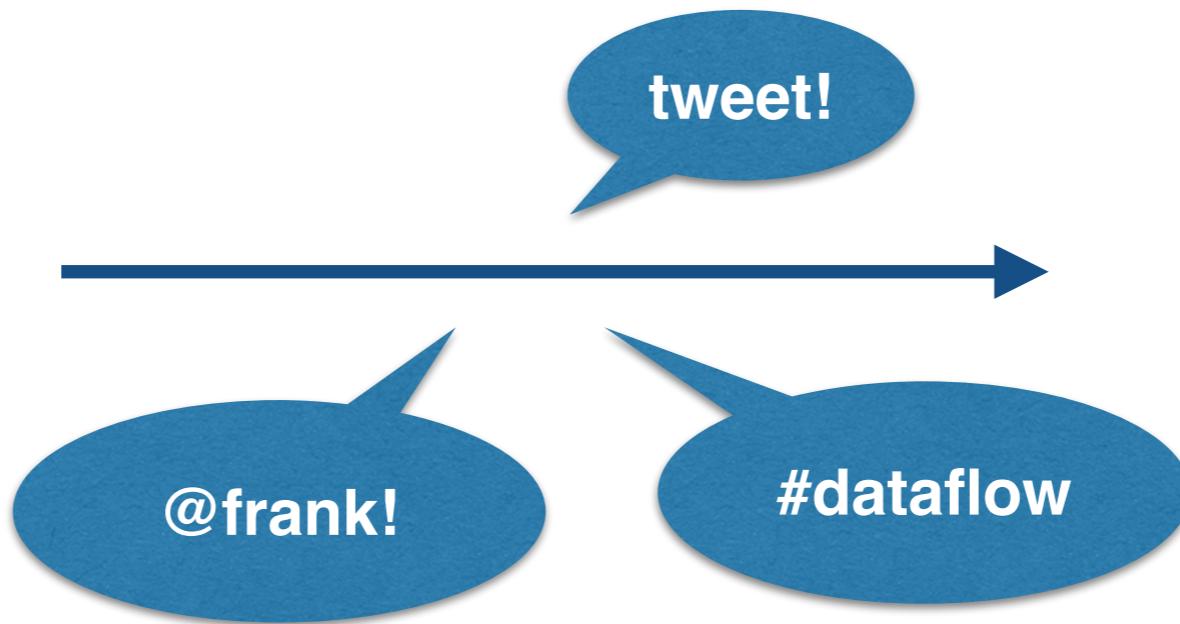




MATERIALIZE

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A motivating problem



What is the most popular #hashtag?
by component of @mention graph?
in real-time (millisecond latencies)?

Differential dataflow

An expressive programming framework that updates computation when inputs change.

<http://github.com/TimelyDataflow/differential-dataflow>

[based off of Differential Dataflow, CIDR 2013]

Goal: collection-oriented programming language,
but then allow the collections to change.

People are good at programming with collections.
(at least, better than with streams)

```
fn your_prog: [D] -> [R] = /* .. */;  
  
// Intended experience:  
for t in times {  
    let output[t] = your_prog(input[t]);  
}  
  
d_output: Stream<(Data, Time, Diff)>
```

input streams of changes

```
let nodes = /* pairs (node, bool) */;  
let edges = /* pairs (node, node) */;
```

“program” : dataflow assembly

```
nodes.join(&edges)    // one hop neighbors  
    .concat(&nodes) // plus original nodes  
    .distinct()    // extended neighborhood
```

dataflow execution

```
for t in times {  
    nodes.insert(..);  
    edges.insert(..);  
}
```

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for t in times {
    nodes.insert(..); nodes.remove(..);
    edges.insert(..); edges.remove(..);
}
```

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let nodes = /* pairs (node, bool) */;
let edges = /* pairs (node, node) */;

nodes.join(&edges)    // one hop neighbors
    .concat(&nodes) // plus original nodes
    .distinct()     // extended neighborhood

for t in times {
    nodes.insert(..); nodes.remove(..);
    edges.insert(..); edges.remove(..);
}
```

```
let nodes = /* pairs (node, bool) */;
let edges = /* pairs (node, node) */;

nodes.iterate(|reach| {
    nodes.join(&edges)    // one hop neighbors
        .concat(&nodes) // plus original nodes
        .distinct()     // extended neighborhood
}) ;

for t in times {
    nodes.insert(..); nodes.remove(..);
    edges.insert(..); edges.remove(..);
}
```

```
let nodes = /* pairs (node, bool) */;
let edges = /* pairs (node, node) */;

nodes.iterate(|reach| {
    reach.join(&edges) // one hop neighbors
        .concat(&nodes) // plus original nodes
        .distinct() // extended neighborhood
}) ;
```

Stream<((node, bool), (Time, u64), int)>

```
for t in t: {
    nodes. ...
    edges. ...
}
```

Secret sauce: Incremental computation done with respect to a partial order, rather than a total order.

collection(t) = sum_{s ≤ t} differences(s)

Reach	cores	livejournal	orkut	•	•
GraphX	128	36s	48s		
SociaLite	128	52s	67s	•	•
Myria	128	5s	6s		
BigDatalog	128	17s	20s	•	•

[BigDatalog, SIGMOD 2016]

Reach	cores	livejournal	orkut
GraphX	128	36s	48s
SociaLite	128	52s	67s
Myria	128	5s	6s
BigDatalog	128	17s	20s
Differential	1	7s	15s

Reach

cores

livejournal

orkut

1000

800

600

400

200

0

update

10

100

1000

20,000

vars

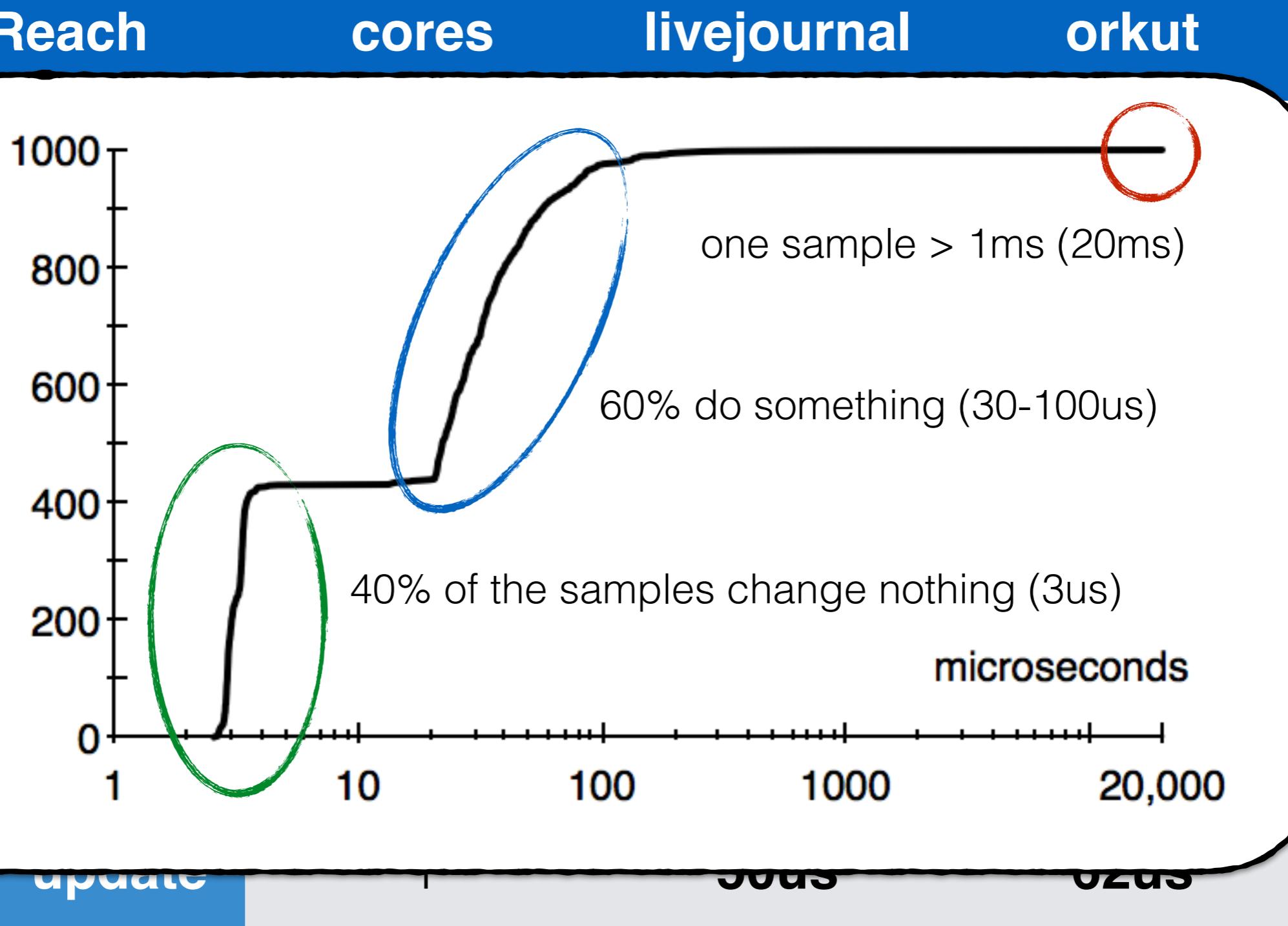
vars

microseconds

40% of the samples change nothing (3us)

60% do something (30-100us)

one sample > 1ms (20ms)



```
let nodes = /* pairs (node, bool) */;
let edges = /* pairs (node, node) */;

nodes.iterate(|reach| {
    reach.join(&edges) // one hop neighbors
        .concat(&nodes) // plus original nodes
        .distinct() // extended neighborhood
}) ;

for t in times {
    nodes.insert(..);
    edges.remove(..);
}
```

```
let nodes = /* pairs (node, node) */;
let edges = /* pairs (node, node) */;

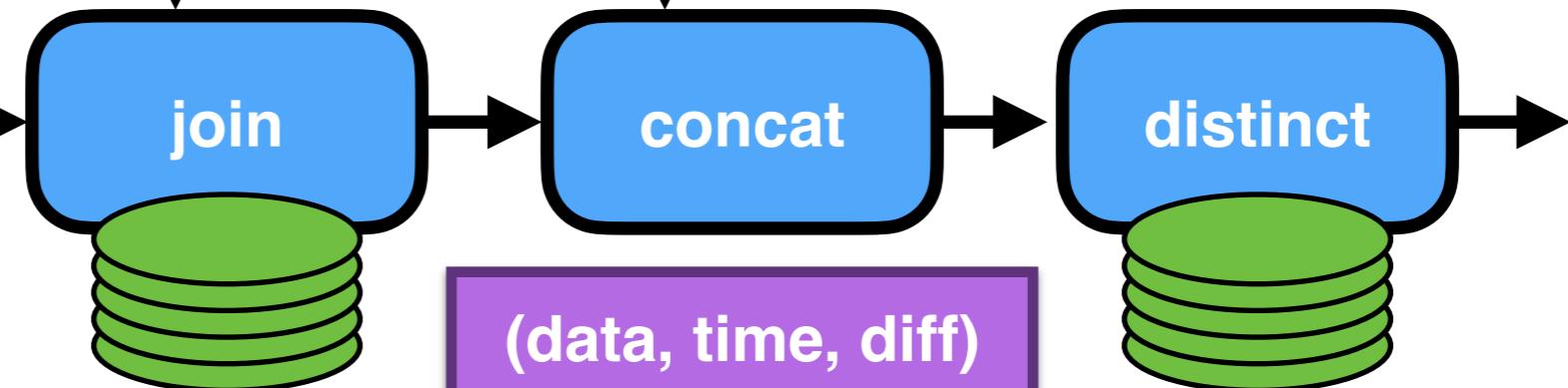
nodes.iterate(|reach| {
    reach.join(&edges) // one hop neighbors
        .concat(&nodes) // plus original nodes
        .min() // smallest labels
});

for t in times {
    nodes.insert(..);
    edges.remove(..);
}
```

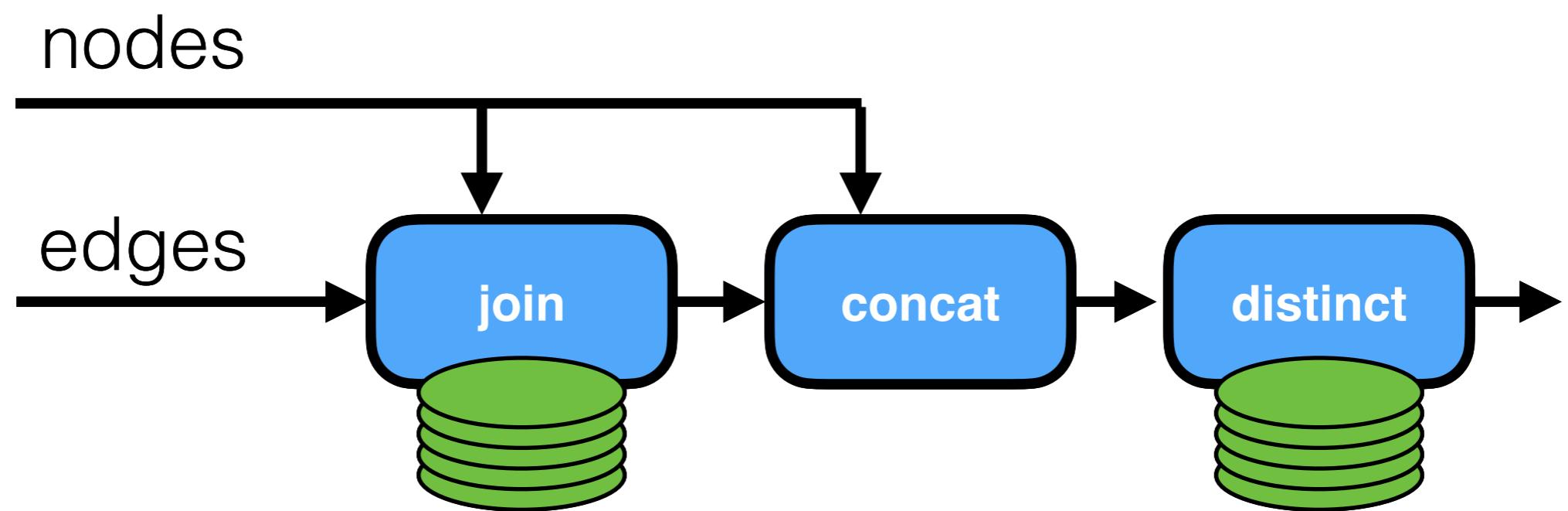
`Stream<(Data, Time, Diff)>`

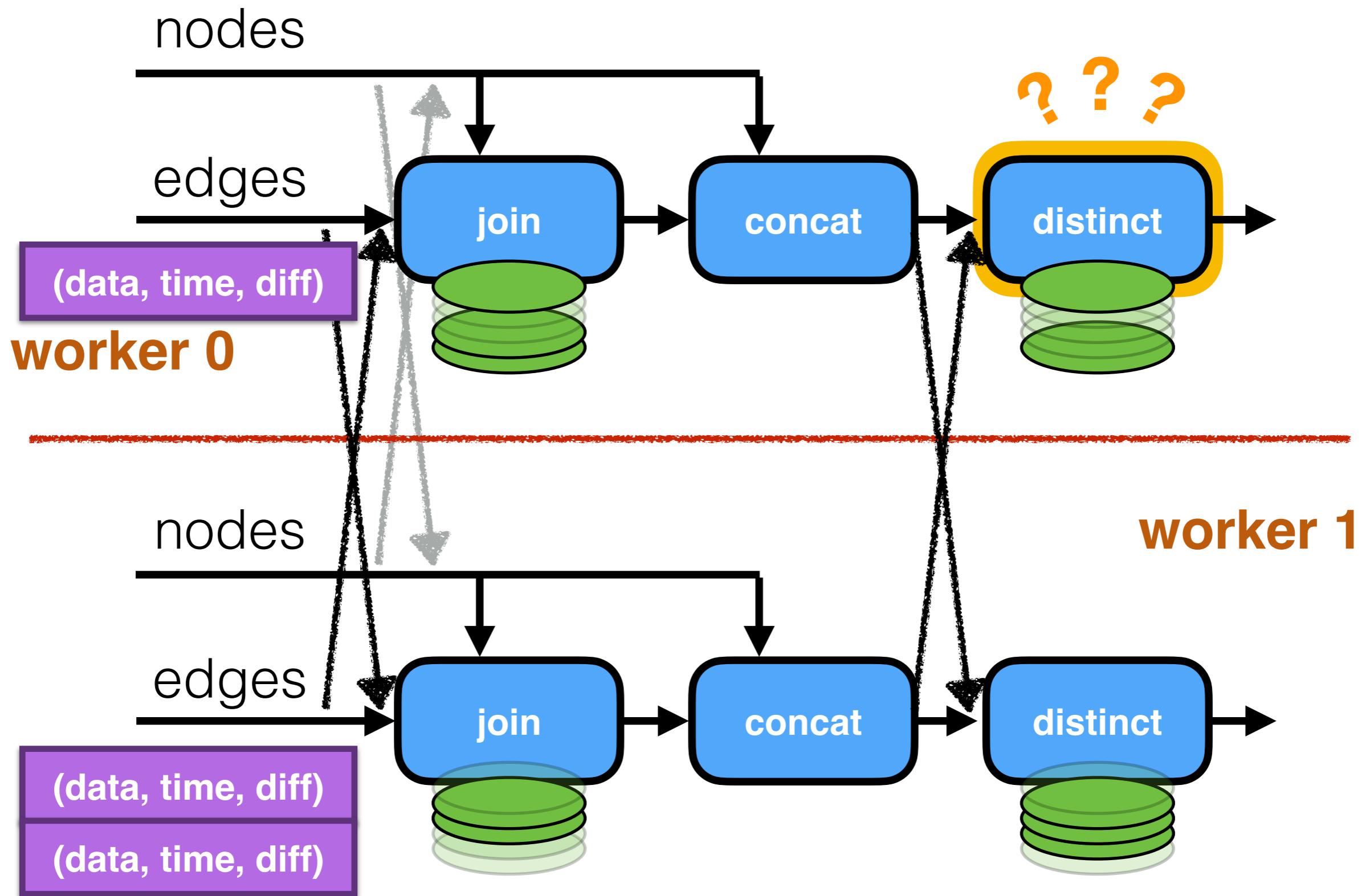
nodes

edges

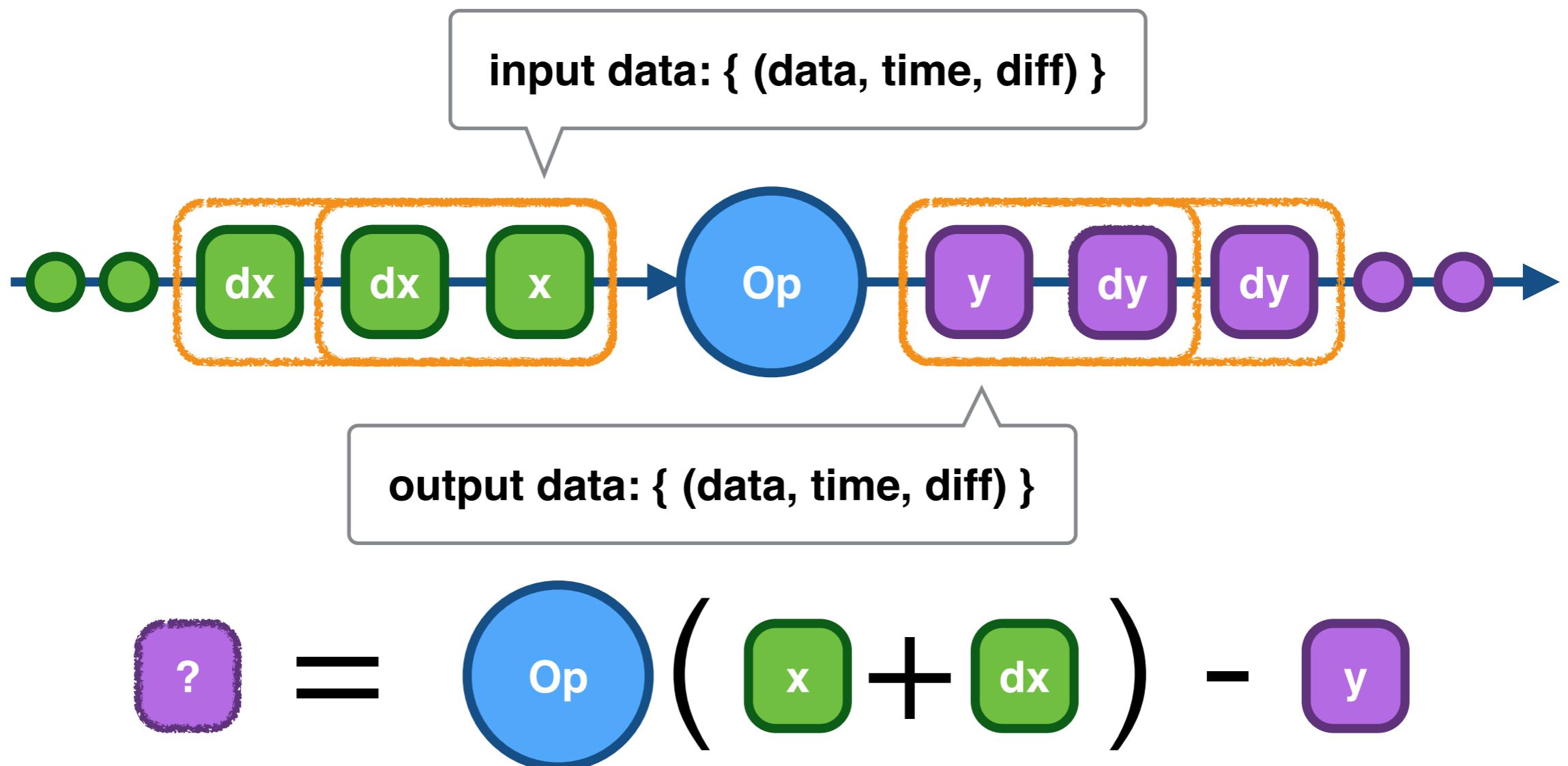


`(data, time, diff)`



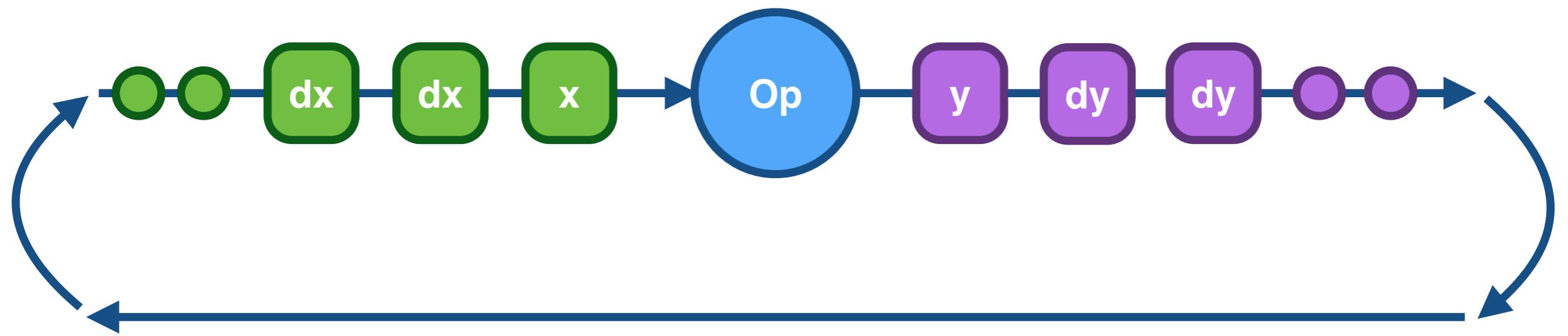


Incremental Dataflow

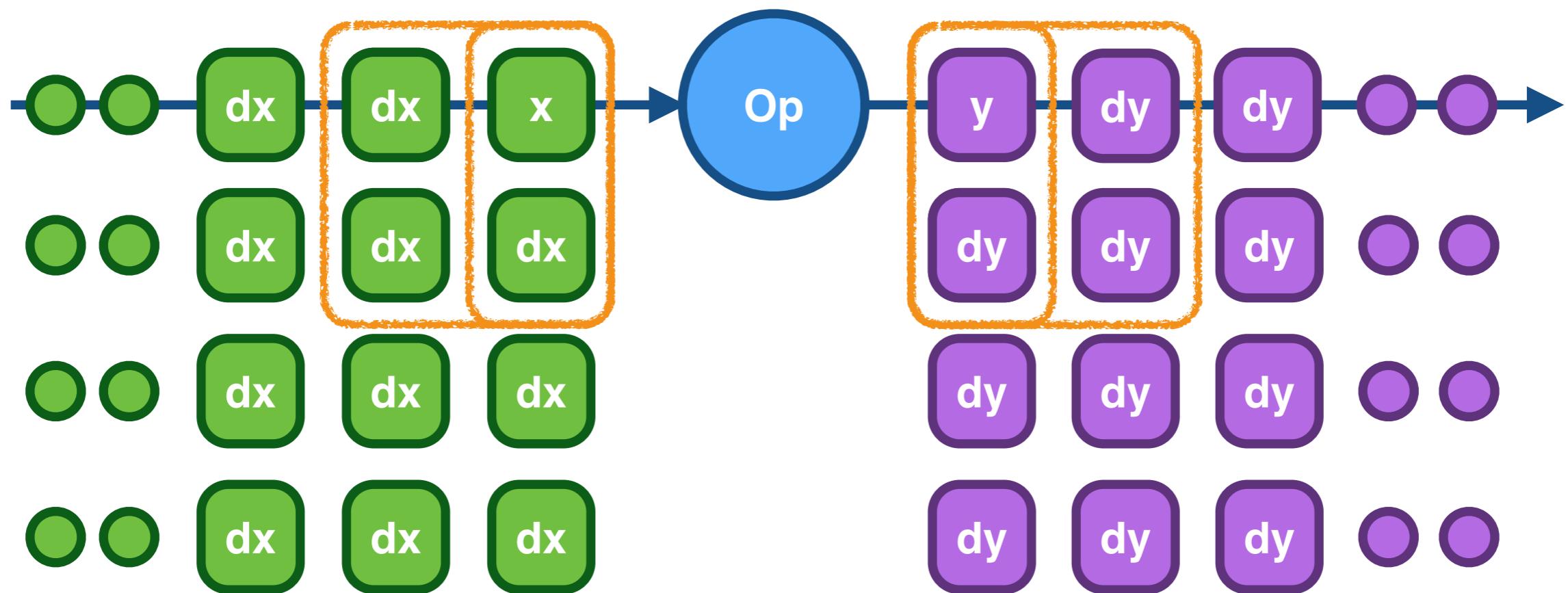


Iterative Dataflow

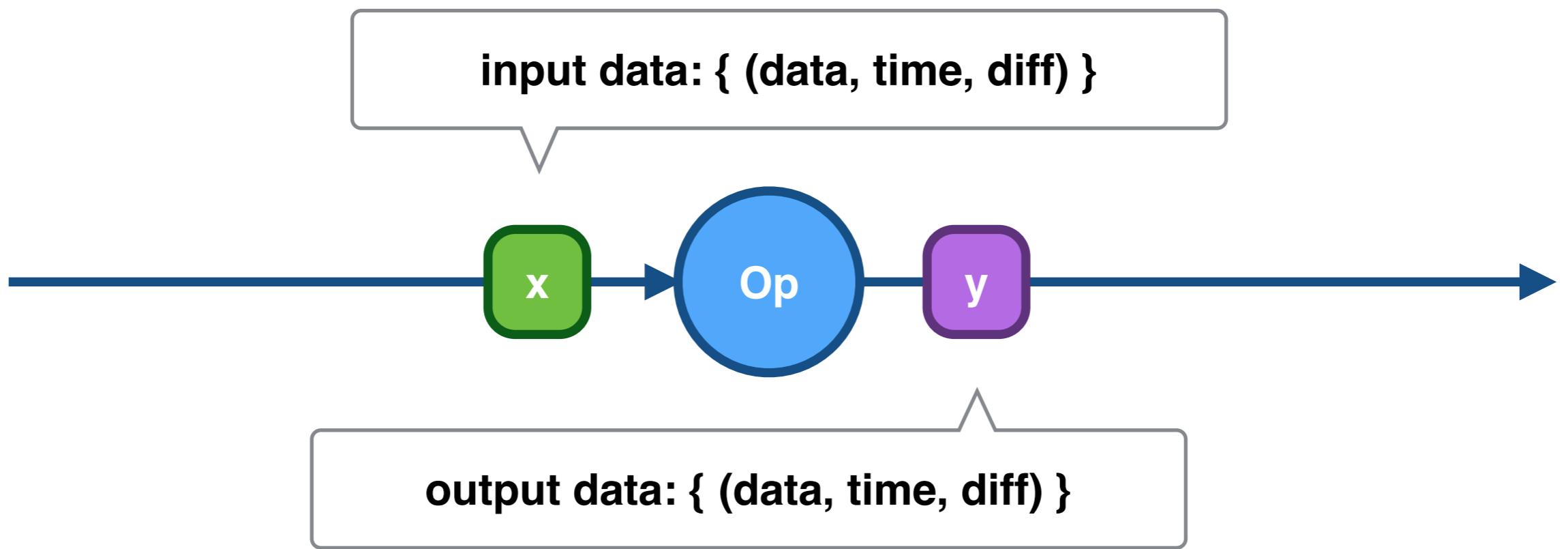
e.g. semi-naive bottom-up datalog



Differential Dataflow



Differential Dataflow



Important: times are only **partially** ordered

Differentiation on a discrete partial order

Differential Dataflow

with data-parallel operators



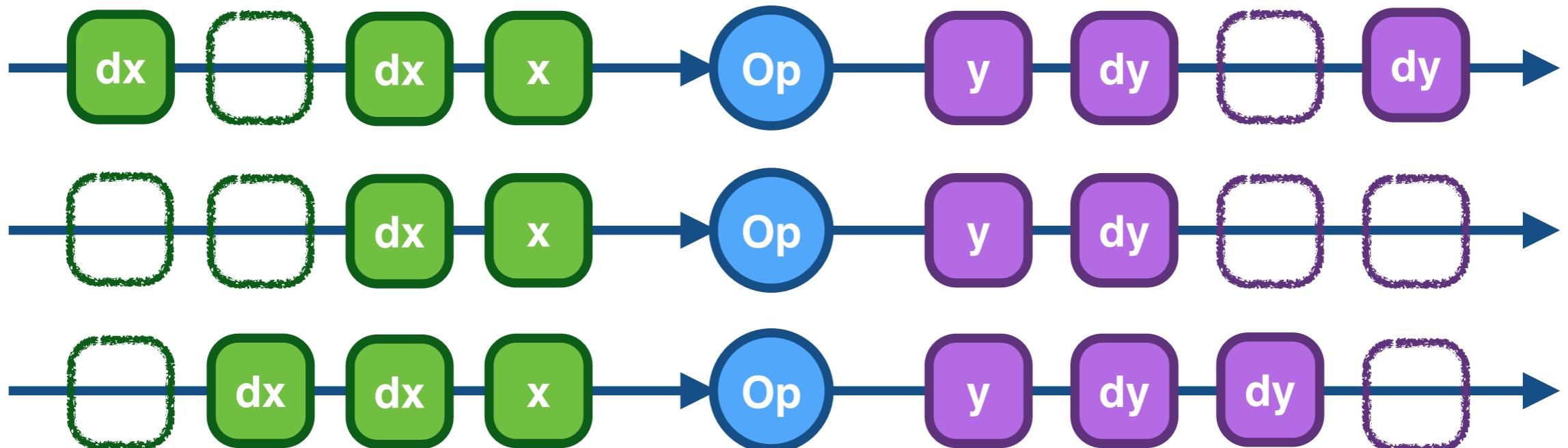
Differential Dataflow

with data-parallel operators



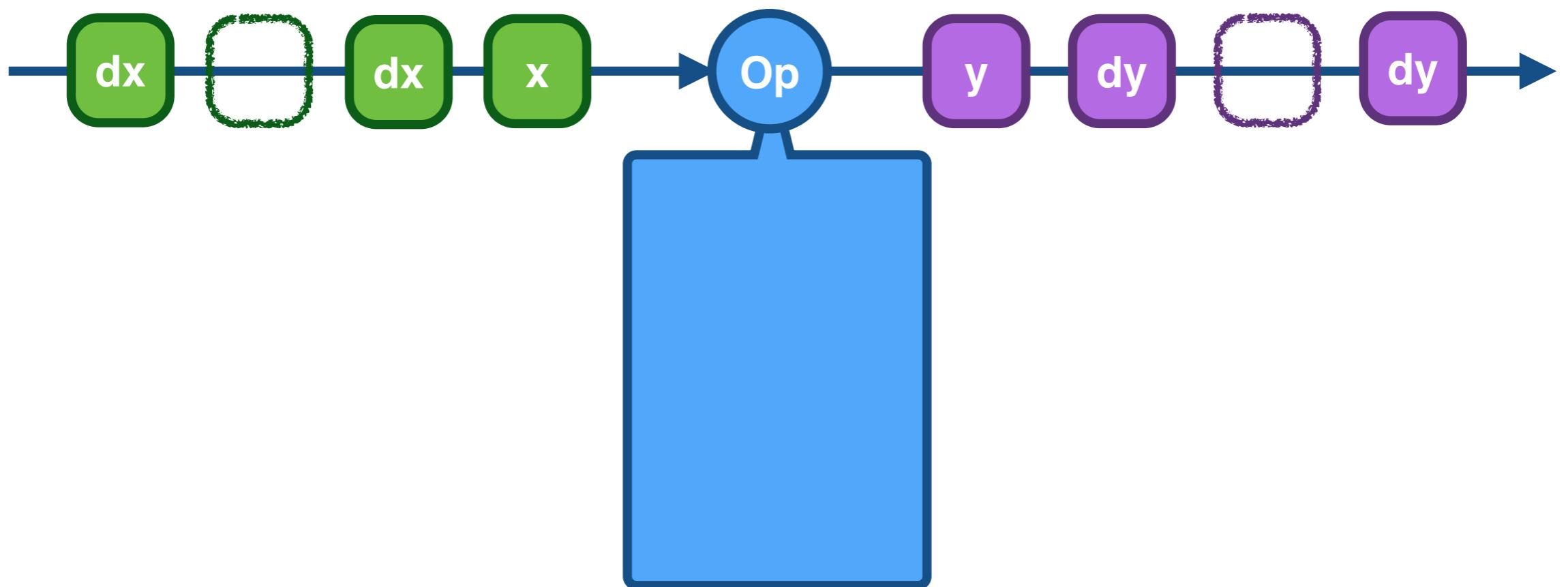
Differential Dataflow

with data-parallel operators



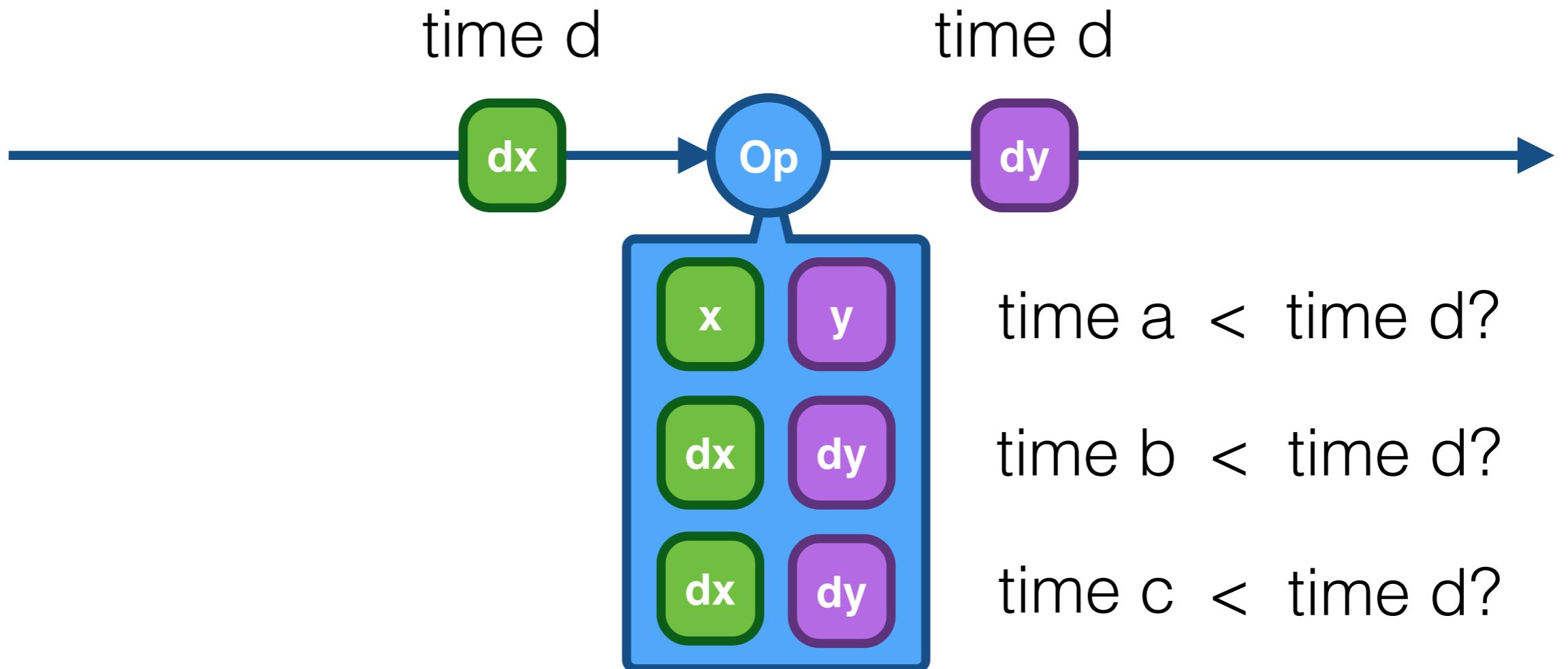
Differential Dataflow

with data-parallel operators



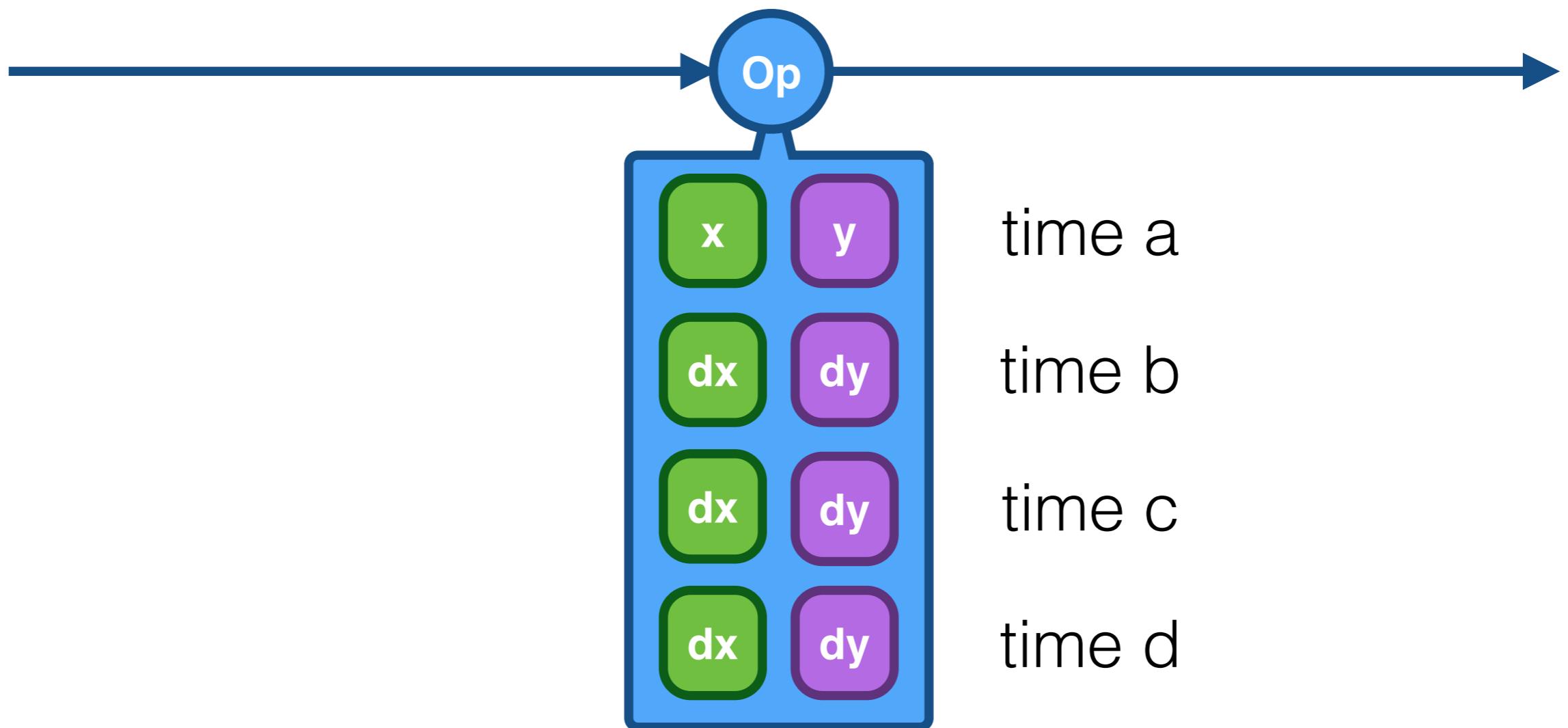
Differential Dataflow

with data-parallel operators

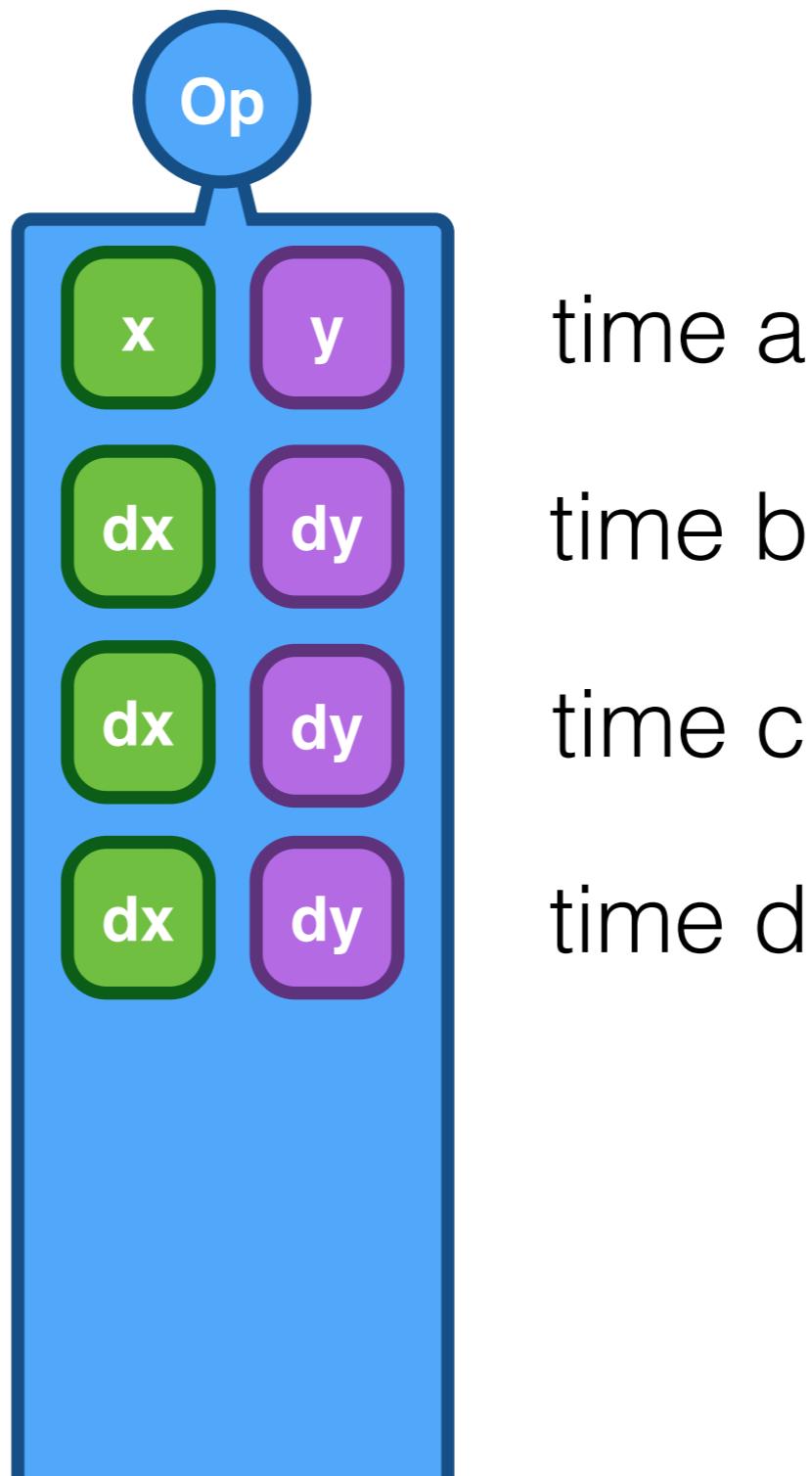


Differential Dataflow

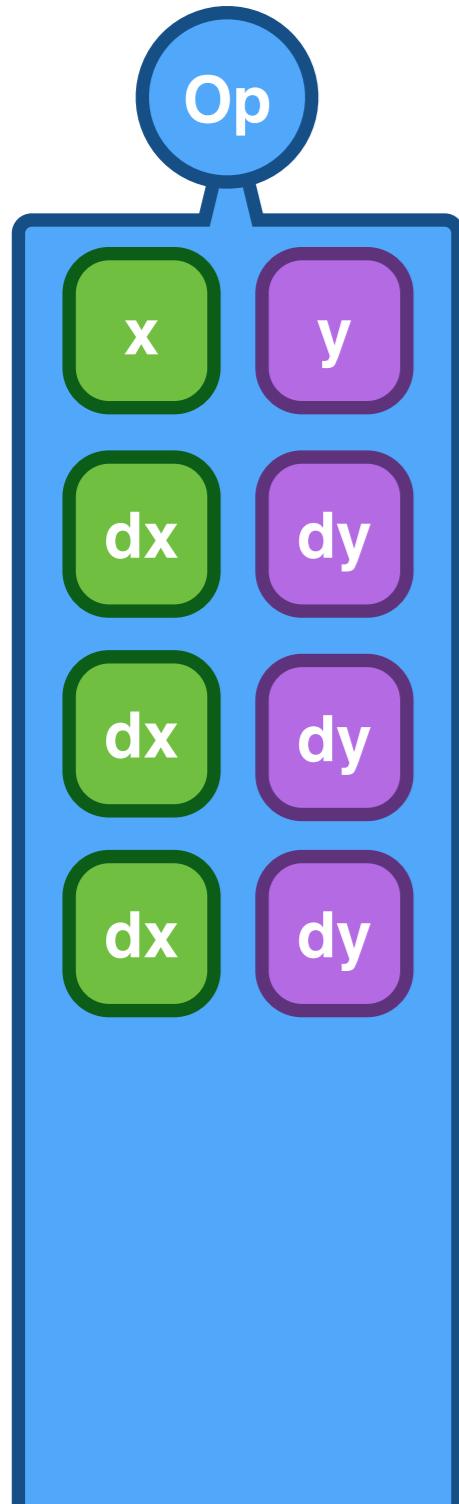
with data-parallel operators



Diversion: Compaction



Diversion: Compaction



Let F be a set of times lower-bounding those times we might see in the future.

$$t_1 \equiv_F t_2 \text{ when } \forall_{f \geq F} (t_1 \leq f \text{ iff } t_2 \leq f)$$

Each time t has a “representative” from this equivalence class, computed as

$$rep_F(t) := \vee_{f \in F} (t \wedge f)$$

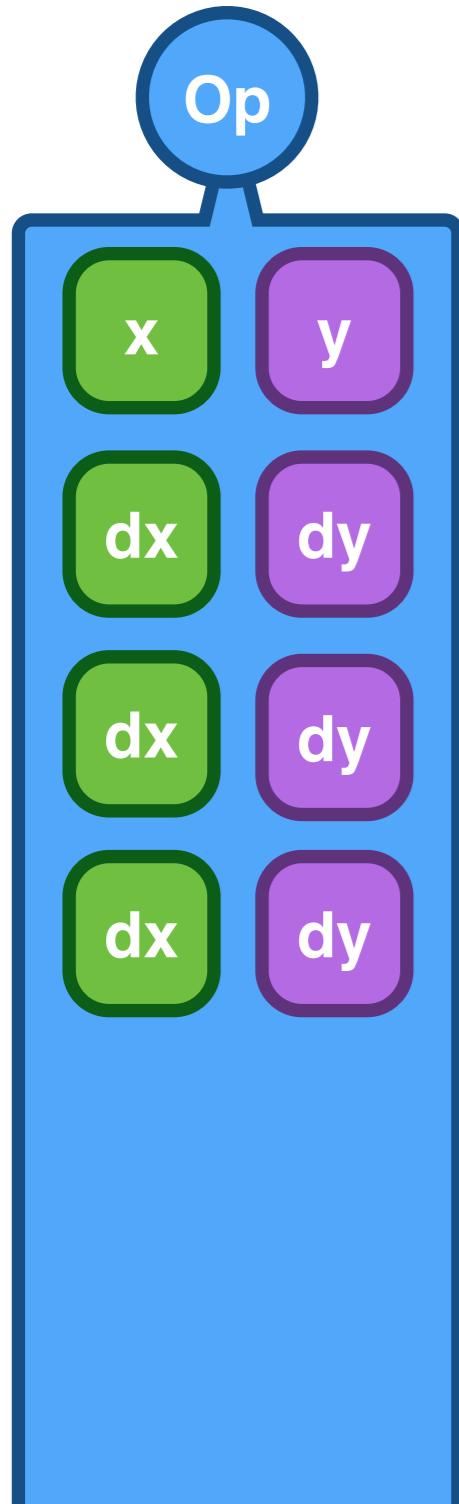
THEOREM A.1 (CORRECTNESS). *For any lattice element t and set F of lattice elements, $t \equiv_F \text{rep}_F(t)$.*

PROOF. We prove both directions of the implication in \equiv_F separately, for all $f \geq F$. First assume $t \leq f$. By assumption, f is greater than some element f' of F , and so $t \wedge f' \leq f$ by the (lub) property. As a lower bound, $\text{rep}_F(t) \leq t \wedge f'$ for each $f' \in F$, and by transitivity $\text{rep}_F(t) \leq f$. Second assume $\text{rep}_F(t) \leq f$. Because $t \leq (t \wedge f')$ for all $f' \in F$, then $t \leq \text{rep}_F(t)$ by the (glb) property and by transitivity $t \leq f$. \square

THEOREM A.2 (OPTIMALITY). *For any lattice elements t_1 and t_2 and set F of lattice elements, if $t_1 \equiv_F t_2$ then $rep_F(t_1) = rep_F(t_2)$.*

PROOF. For all $f \in F$ we have both that $t_1 \leq t_1 \wedge f$ and $f \leq t_1 \wedge f$, the latter implying that $t_1 \wedge f \geq F$. By our assumption, t_2 agrees with t_1 on times greater than F , making $t_2 \leq t_1 \wedge f$ for all $f \in F$. By correctness, $rep_F(t_2)$ agrees with t_2 on times greater than F , which includes $t_1 \wedge f$ for $f \in F$ and so $rep_F(t_2) \leq t_1 \wedge f$ for all $f \in F$. Because $rep_F(t_2)$ is less or equal to each term in the greatest lower bound definition of $rep_F(t_1)$, it is less or equal to $rep_F(t_1)$ itself. The symmetric argument proves that $rep_F(t_1) \leq rep_F(t_2)$, which implies that the two are equal (by antisymmetry). \square

Diversion: Compaction



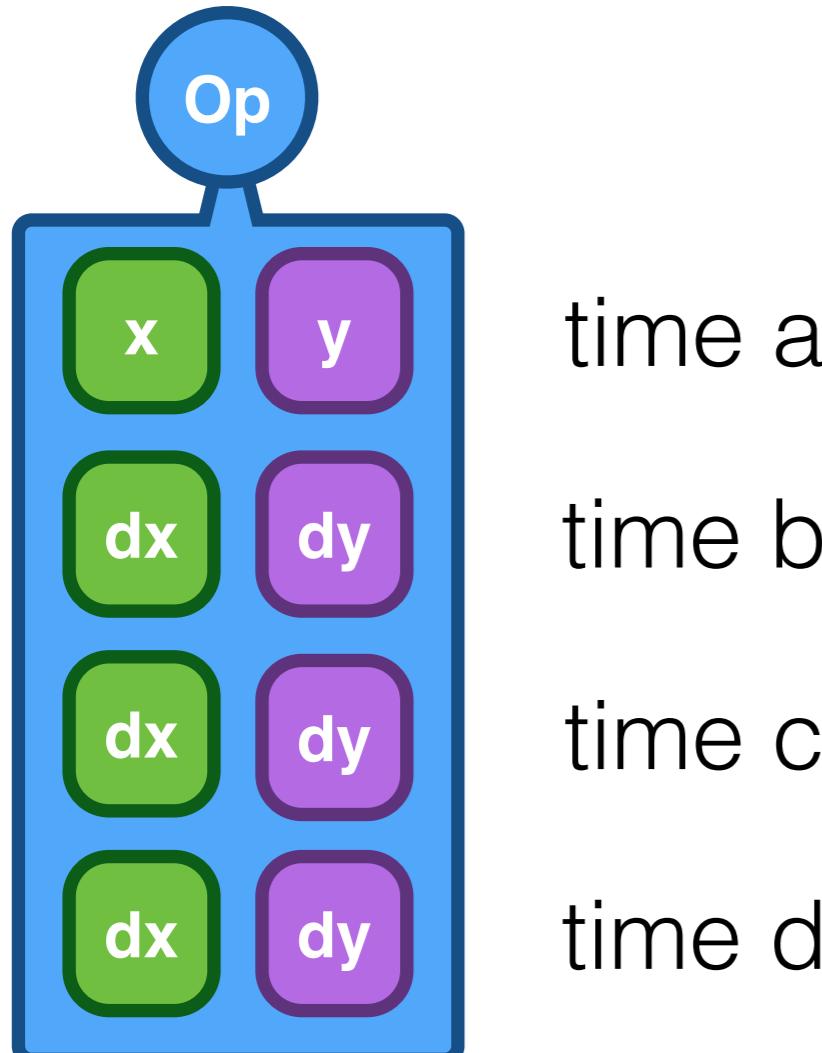
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Diversion: Compaction



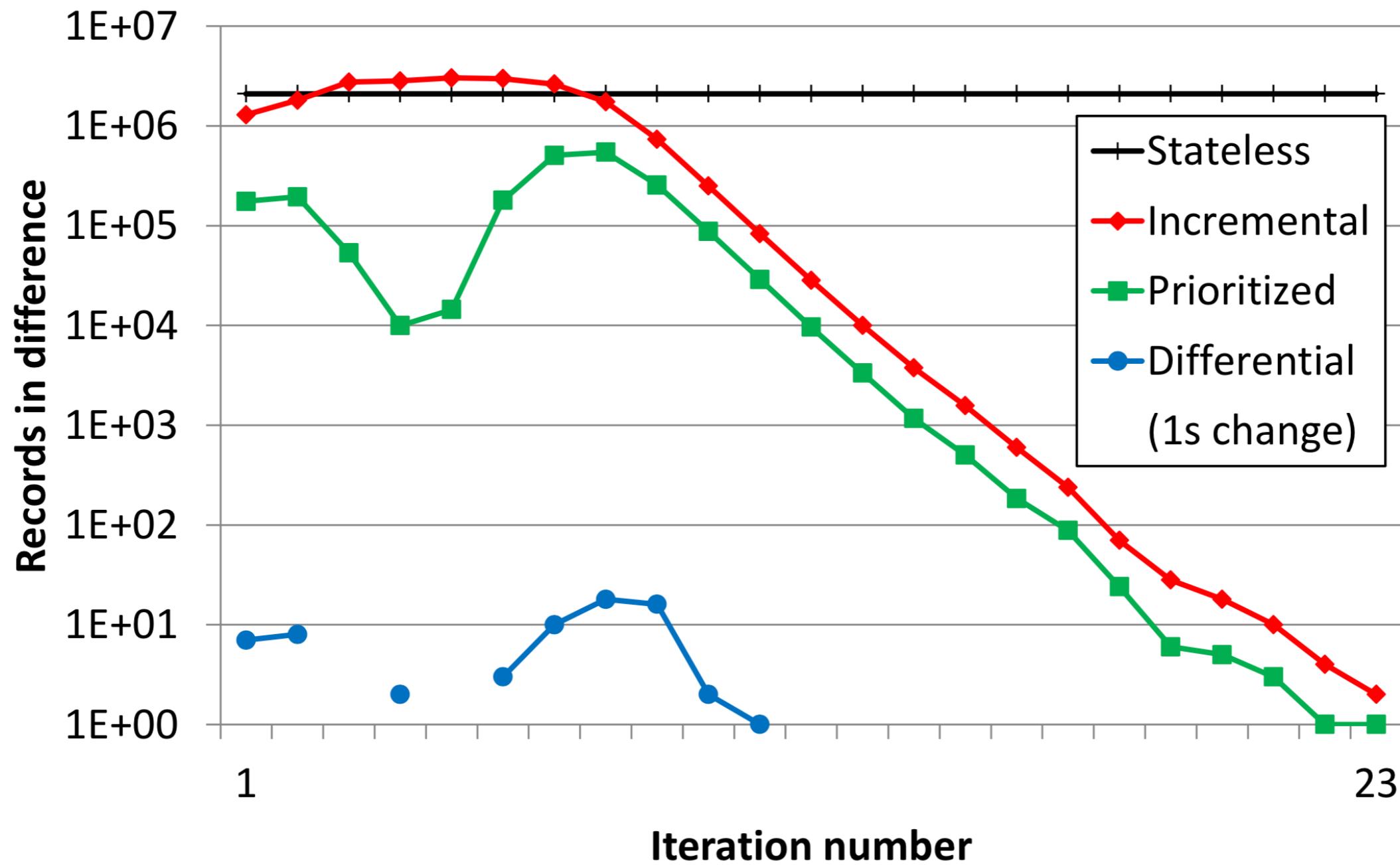
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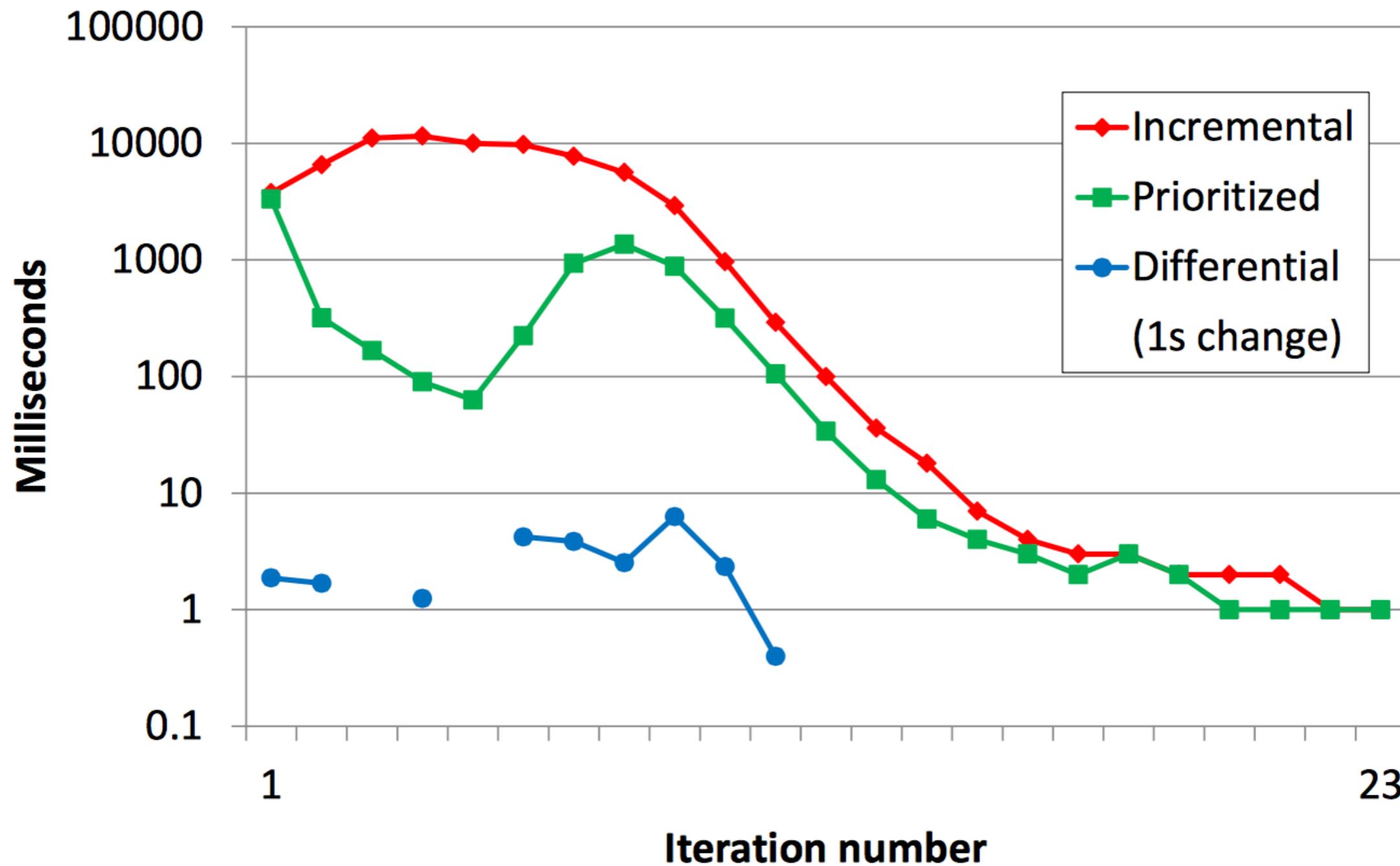
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Connected components on
the Twitter @mention graph





Pause

Things left unsaid

Arrangements

Operator state can be shared with others.
Credit due to: “declarative programming”.

Open- v. Closed-loop

The world changes even if we are not ready.
Systems should handle batches of changes.

Groups and Semi-groups

All “Diffs” can be arbitrary (semi-)groups.
Types encode the nature of changes.

<https://github.com/TimelyDataflow/differential-dataflow>
/timely-dataflow
/abomonation
/diagnostics



MATERIALIZE

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