

Time-Aware Stream Processing in Isabelle/HOL

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Introduction

Time-Aware Stream Processing

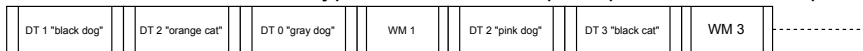
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 - Timestamp: A partially-ordered value associated with the data (e.g., unix-time, int, pairs of ints, etc...)
 - Watermark: A value of the same type of the timestamp. Represents data-completeness.



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- Asynchronous Dataflow Programming: Directed graph of interconnected operators that perform event-wise transformations
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- Why?
 - Highly Parallel
 - Low latency (output as soon as possible)
 - Incremental computing (re-uses previous computations)

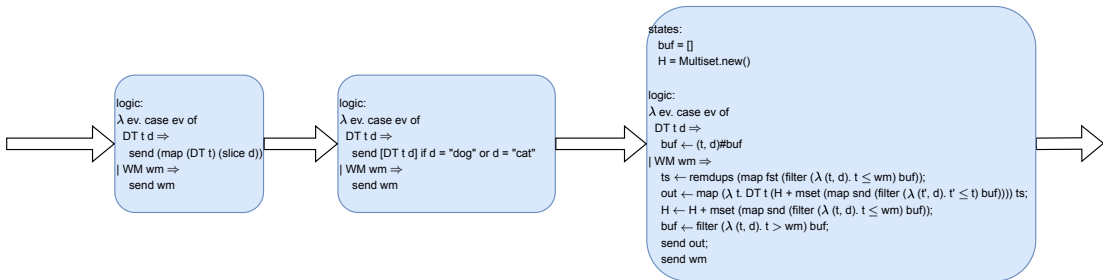
Time-Aware Stream Processing Example

Example:

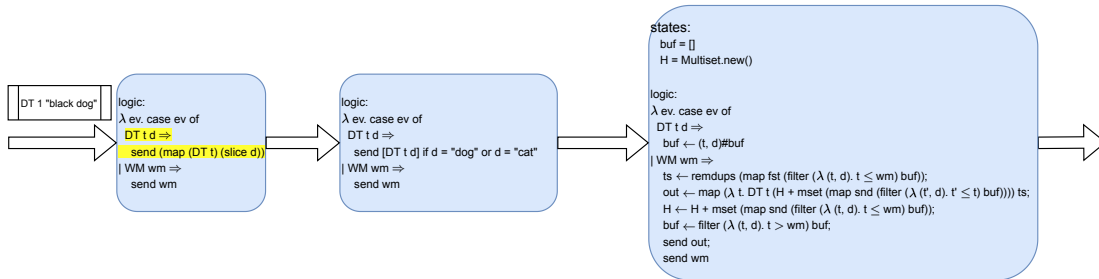
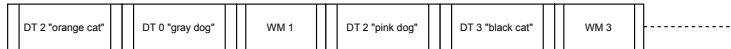
Incrementally count the occurrences of the words “dog” and “cat”



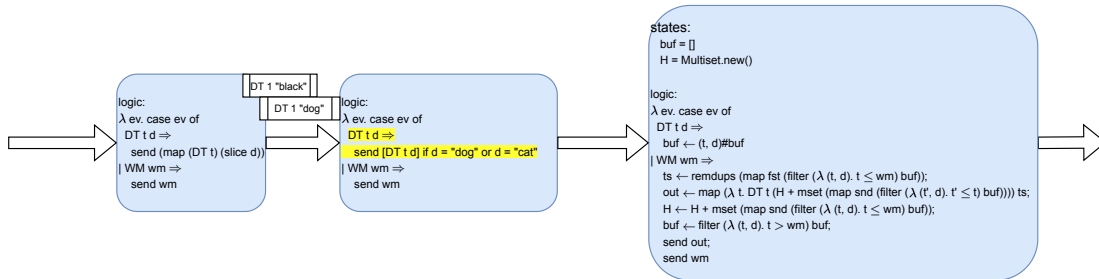
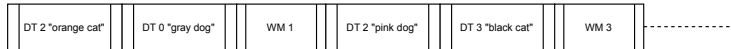
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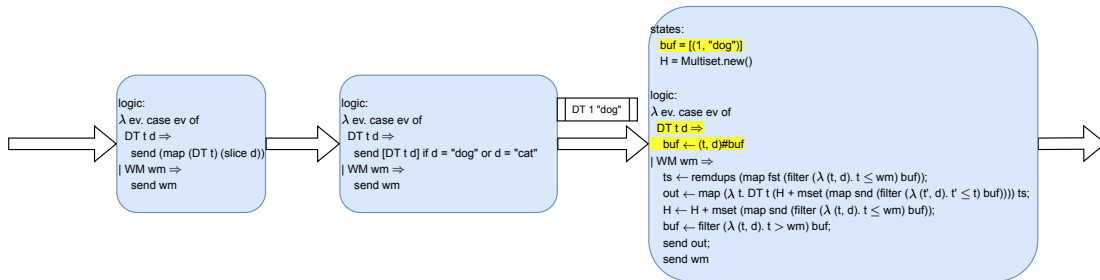
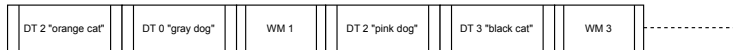
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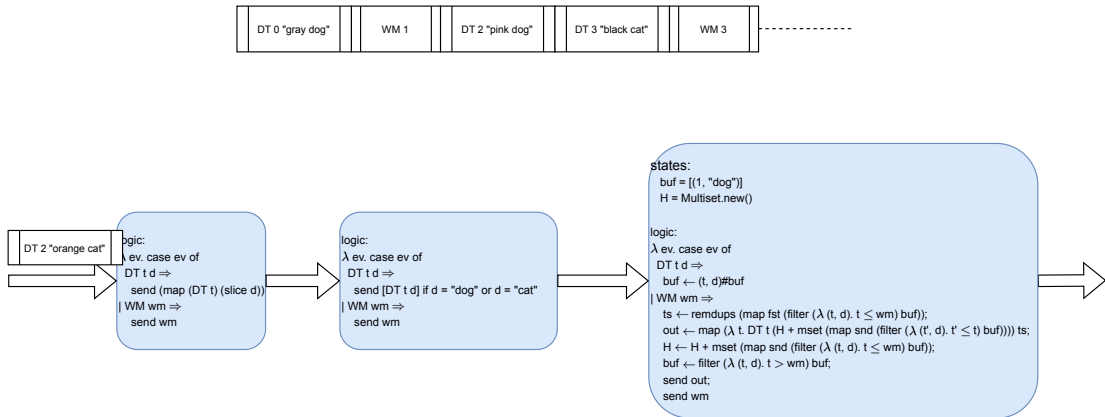
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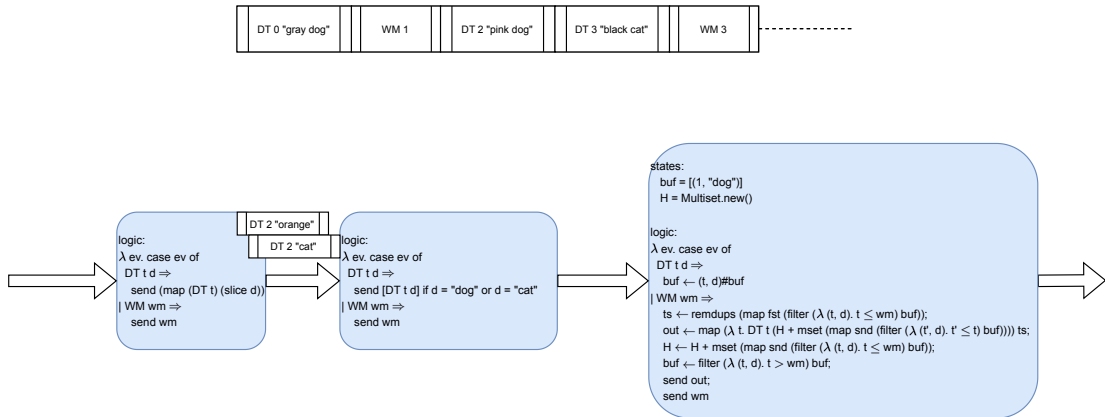
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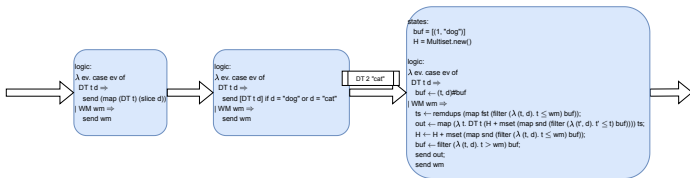
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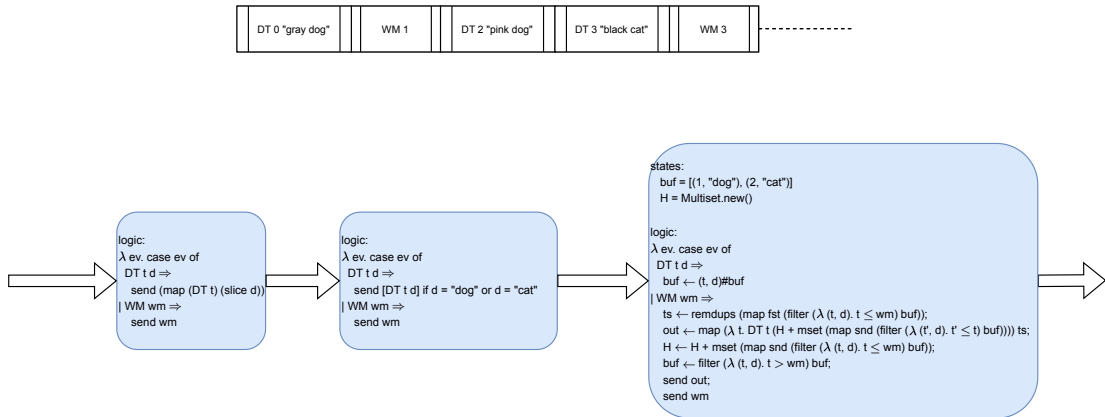
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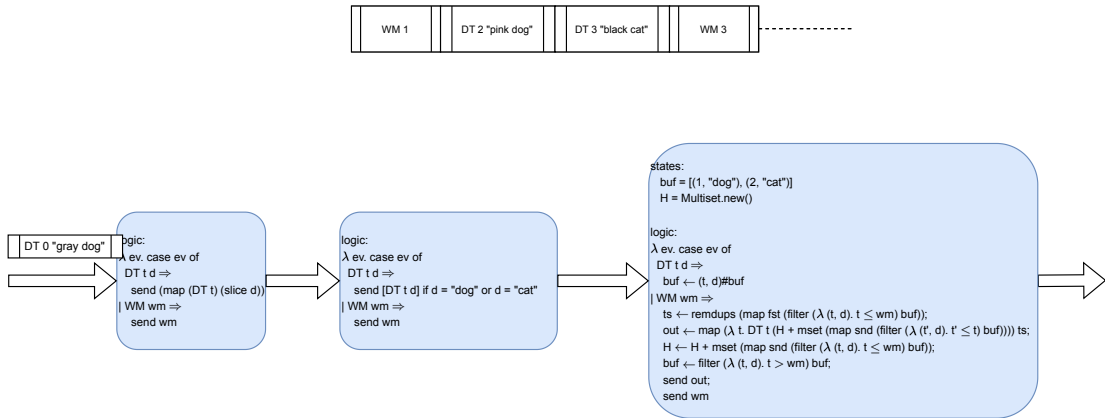
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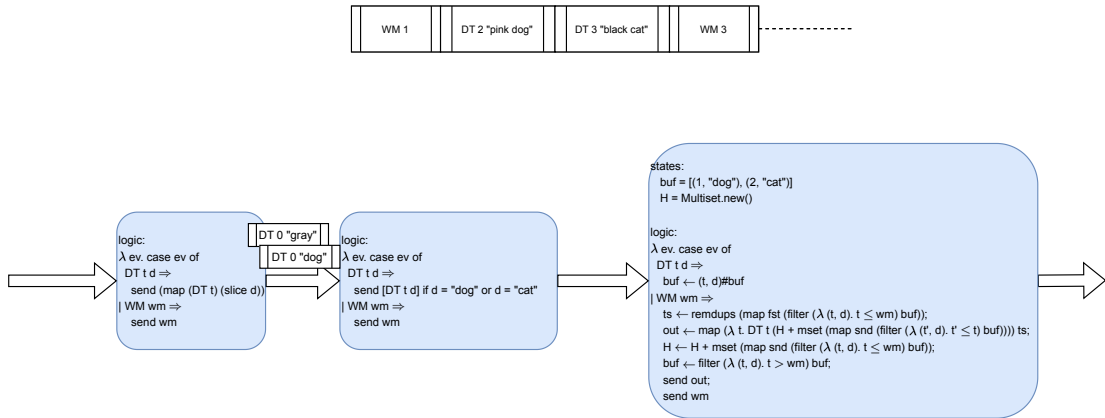
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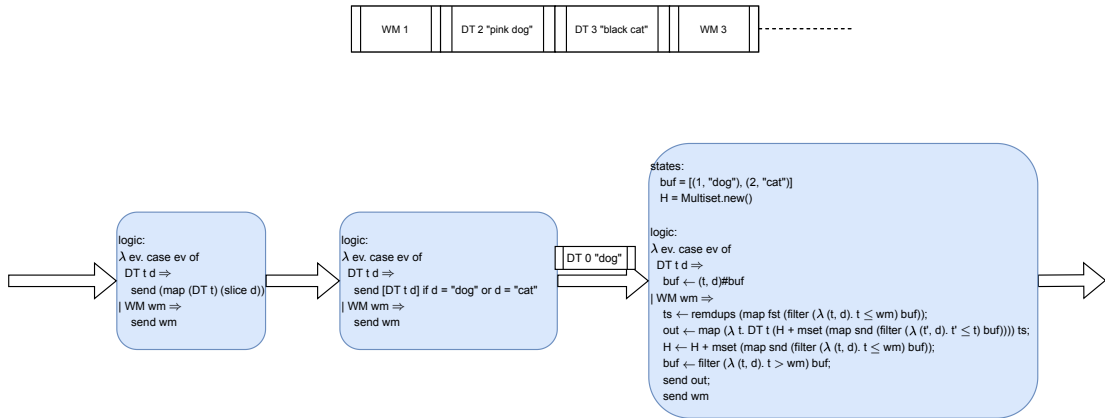
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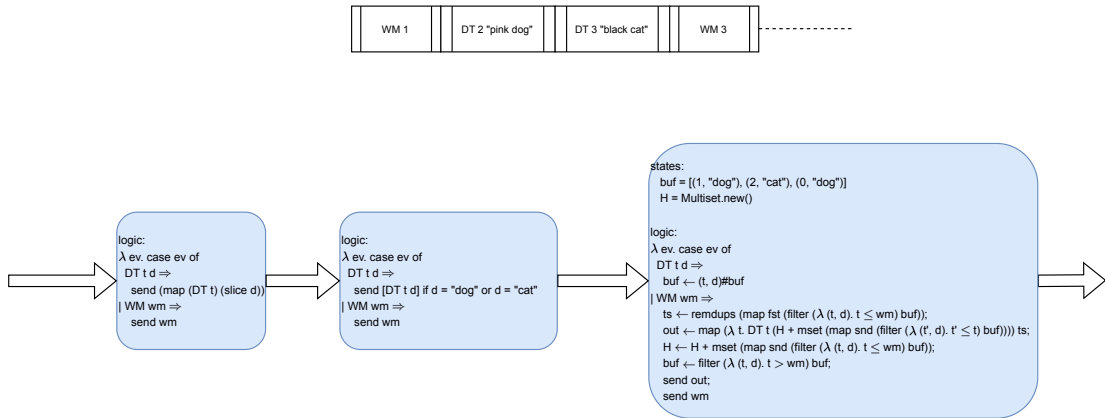
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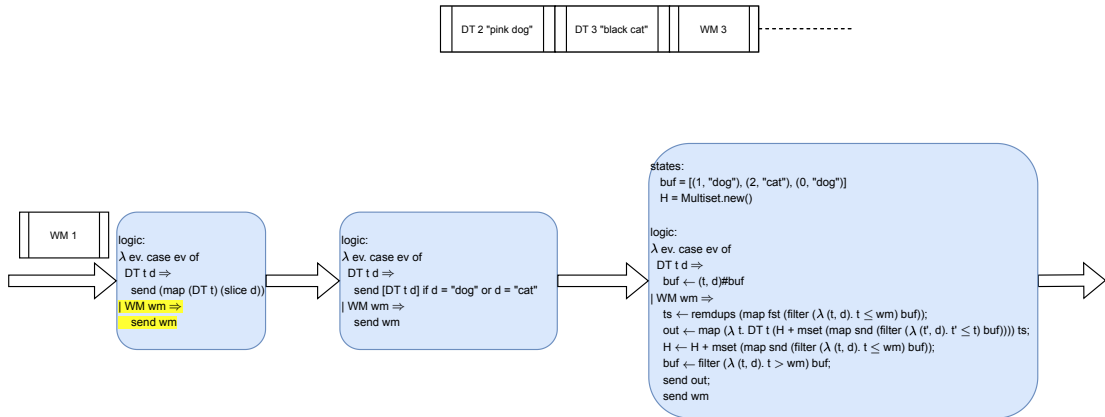
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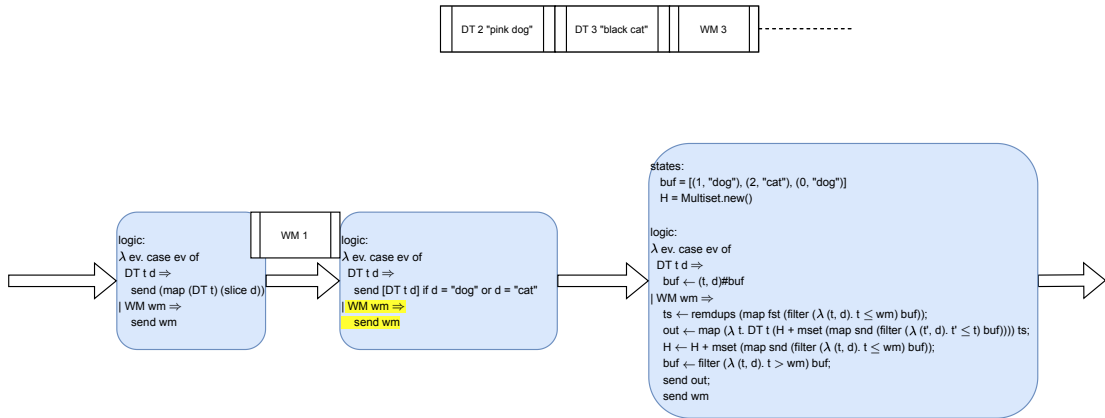
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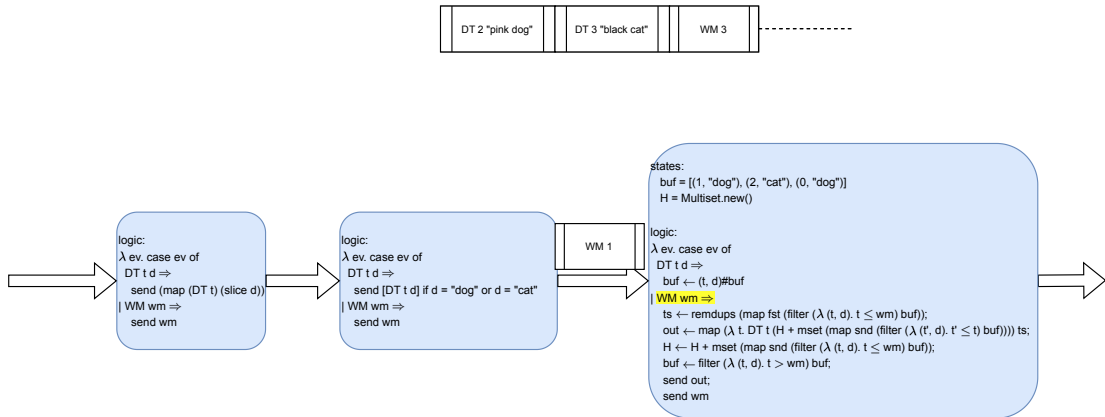
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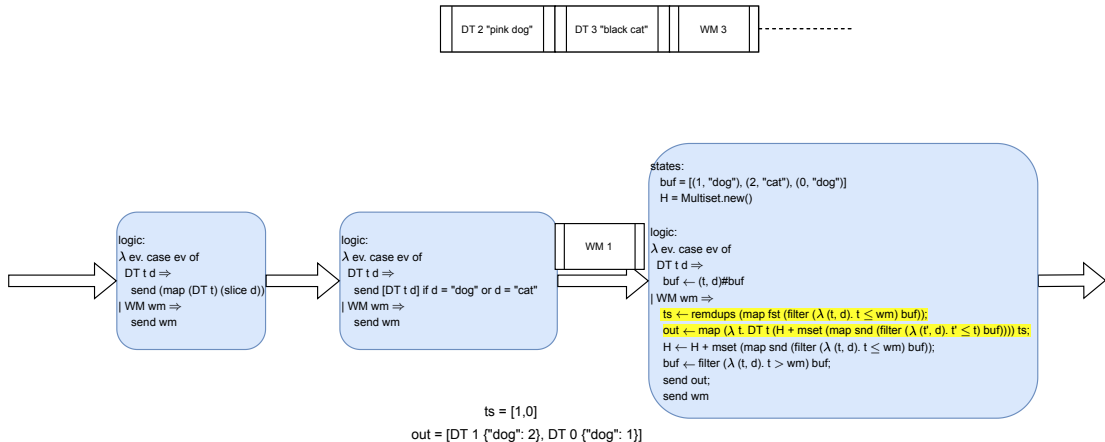
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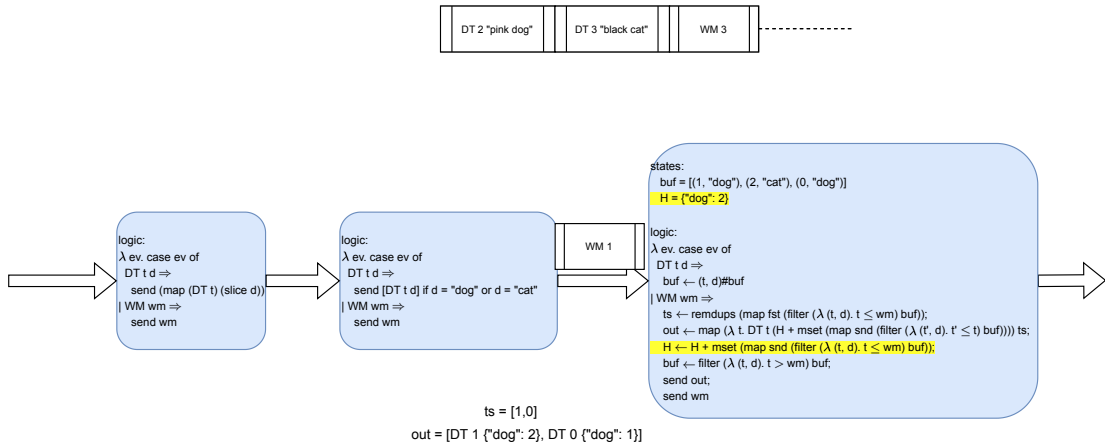
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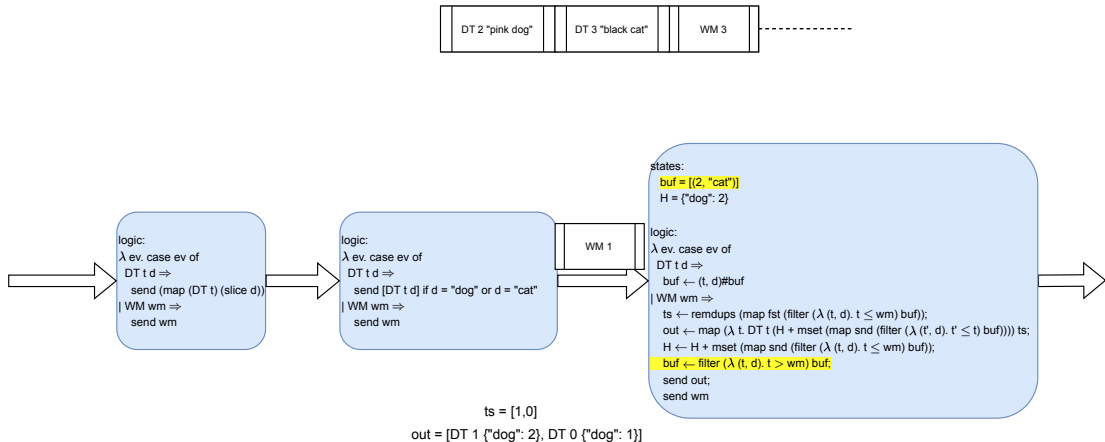
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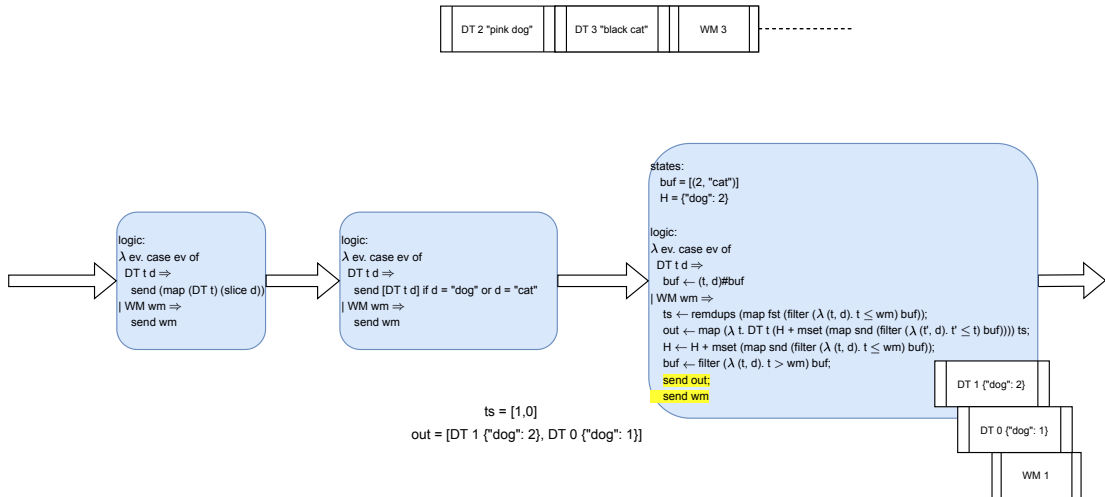
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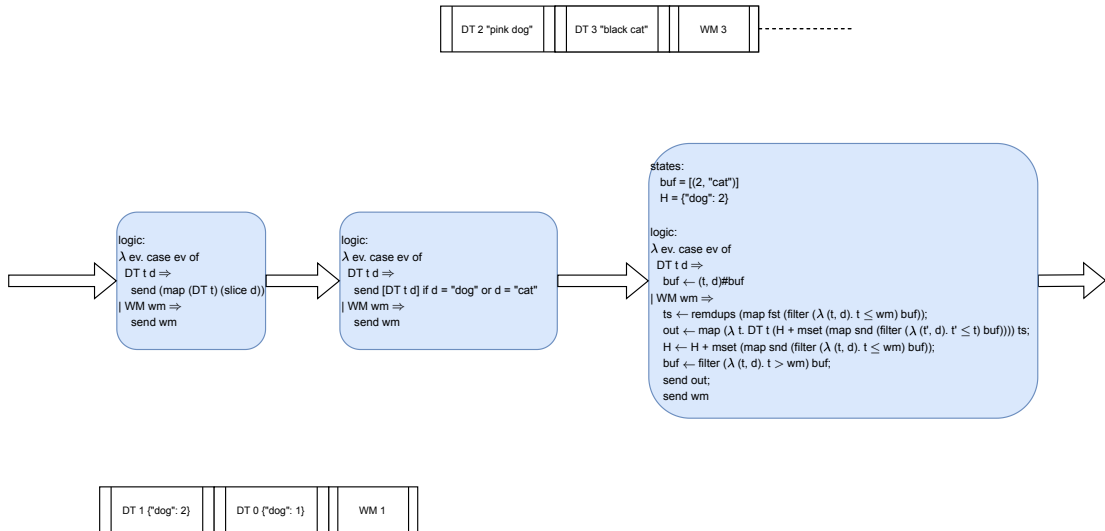
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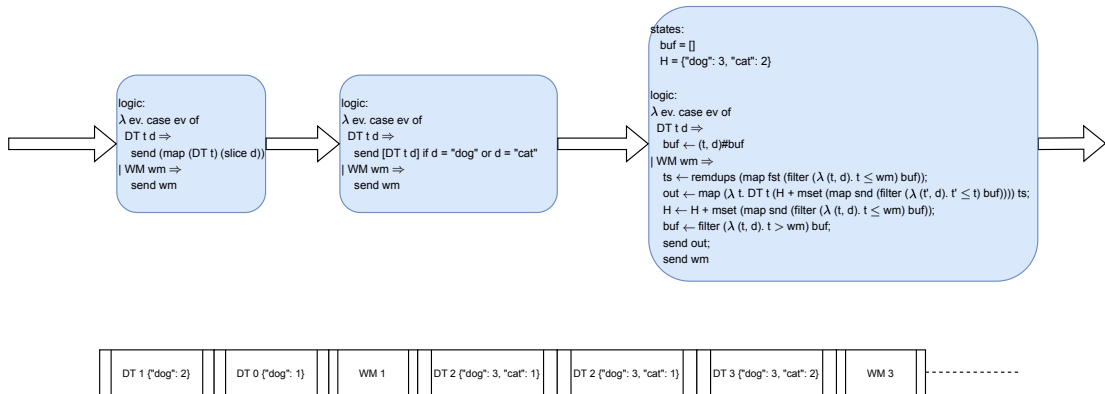
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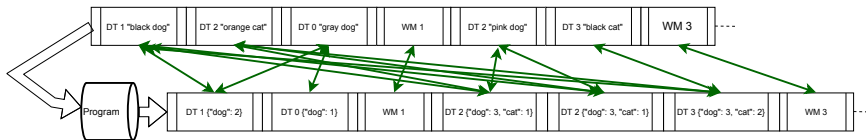


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Properties

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- We need a correctness specification

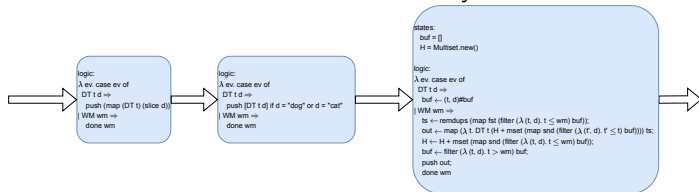
- How do we know if our Dataflow program is what we want?
- We need a correctness specification
- **Intuition of the specification:**
 - Soundness: for every output $DT\ t\ H$, the “dog” count in H is the count of events with timestamp $(\leq)t$ which contains the string “dog”; similarly for “cat”. The count for any other word is always 0.
 - Completeness: The other way around.



How to prove it

- lets break down the problem!:

- The correctness of the entire Dataflow emerges from the correctness of each part (operator)
 - Operator 1: Slicer
 - Operator 2: Filter
 - Operator 3: Incremental histogram
 - Assumptions about the incoming stream:
 - Monotone: after WM w_m no $DT\ t\ d$ such that $t \leq w_m$.
 - Productive: after $DT\ t\ d$ eventually WM w_m such that $t \leq w_m$



- The original incoming stream must respect monotonicity and productivity



- Each operator must preserve monotonicity and productivity!

Formalization in Isabelle/HOL

- Codatatypes

```
codatatype llist = (lnull: LNil) | LCons (lhd: 'a) (ltl: 'a llist)
```

- Selectors: *lhd*, *ltl*, Discriminator: *lnull*

- Examples:

- LNil
- LCons 1 (LCons 2 (LCons 3 LNil))
- LCons 0 (LCons 0 (LCons 0 (...)))

- Codatatypes

codatatype *llist* = (Inull: LNil) | LCons (lhd: 'a) (ltl: 'a *llist*)

- Selectors: *lhd*, *ltl*, Discriminator: *Inull*

- Examples:

- LNil
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- Coinduction principle for lazy list equality

$$\begin{aligned} R \text{ } lxs \text{ } lys &\implies \\ (\bigwedge lxs \text{ } lys. R \text{ } lxs \text{ } lys &\implies \\ \text{Inull } lxs = \text{Inull } lys \wedge \\ (\neg \text{Inull } lxs \longrightarrow \neg \text{Inull } lys \longrightarrow \text{lhd } lxs = \text{lhd } lys \wedge R \text{ } (ctl \text{ } lxs) \text{ } (ctl \text{ } lys))) &\implies \\ lxs = lys \end{aligned}$$

- More about *llist* at the **Coinductive** AFP entry!

- Recursion

```
fun lshift :: 'a list  $\Rightarrow$  'a llist  $\Rightarrow$  'a llist (infixr @@ 65) where  
  lshift [] lxs = lxs  
| lshift (x # xs) lxs = LCons x (lshift xs lxs)
```

- While Combinator

```
definition while_option :: ('a  $\Rightarrow$  bool)  $\Rightarrow$  ('a  $\Rightarrow$  'a)  $\Rightarrow$  'a  $\Rightarrow$  'a option where  
  while_option b c s = ...
```

- From HOL-Library
- *None* (Diverged), *Some* x (Finished with final state x)
- While rule for invariant reasoning (Hoare-style):
 - There is something that holds before a step; that thing still holds after the step

- Recursion: always eventually reduces an argument
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- Corec:

```
corec lapp :: 'a llist  $\Rightarrow$  'a llist  $\Rightarrow$  'a llist where
lapp lxs lys = case lxs of LNil  $\Rightarrow$  lys | LCons x lxs'  $\Rightarrow$  LCons x (lapp lxs' lys)

 $\neg$  lfinite lxs  $\implies$  lapp lxs lys = lxs
```

Isabelle/HOL: (Co)inductive Predicates

- Inductive predicate
 - Finite number of introduction rule applications

```
inductive in_llist :: 'a  $\Rightarrow$  'a llist  $\Rightarrow$  bool where  
  In_llist: in_llist x (LCons x xs)  
| Next_llist: in_llist x xs  $\Rightarrow$  in_llist x (LCons y xs)  
  
in_llist 2 (LCons 1 (LCons (2 (...))))
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- Coinductive predicate
 - Infinite number of introduction rule applications

```
coinductive lprefix :: 'a llist  $\Rightarrow$  'a llist  $\Rightarrow$  bool where  
  LNil_lprefix: lprefix LNil lxs  
| LCons_lprefix: lprefix lxs lxs  $\Rightarrow$  lprefix (LCons x lxs) (LCons x lxs)  
  
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- Coinduction principle

Lazy Lists Processors (a.k.a. Non-Time-Aware Stream Processing)

Operator formalization

- Operator as a codatatype¹

- Taking `'i` as the input type, and `'o` as the output type:

`codatatype ('o, 'i) op = Logic (apply: ('i \Rightarrow ('o, 'i) op \times 'o list))`

¹This is a simplification of the codatatype in the paper!

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`codatatype ('o, 'i) op = Logic (apply: ('i \Rightarrow ('o, 'i) op \times 'o list))`

- Infinite trees: applying the selector `apply` “walks” a branch of the tree

```
corec count_op where count_op P n =  
  Logic ( $\lambda e$ . if P e then (count_op P (n + 1), [n+1]) else (count_op P n, []))
```

```
apply (count_op is_even 0) 0 = (count_op is_even 1, [1])
```

¹This is a simplification of the codatatype in the paper!

- Produce function: applies the logic (co)recursively throughout a lazy list

definition produce₁ op lxs = while_option ...

corec produce **where**

produce op lxs = (case produce₁ op lxs of

None \Rightarrow LNil

| Some (op', x, xs, lxs') \Rightarrow LCons x (xs @@ produce op' lxs'))

- lshift (@@) is a friend of corec function!²

²More about it at the “Friends with benefits” paper by Jasmin Christian Blanchette, Aymeric Bouzy, Andreas Lochbihler, Andrei Popescu, and Dmitriy Traytel

- Produce function: applies the logic (co)recursively throughout a lazy list

definition $\text{produce}_1 \text{ op } lxs = \text{while_option } \dots$

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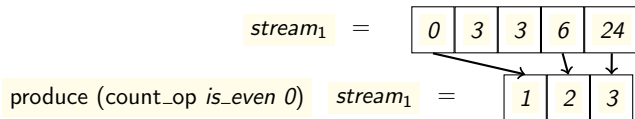
- lshift (@@) is a friend of corec function!²
- produce₁ has an induction principle based on the while invariant rule

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Operators: Count

- Example:

```
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```



Sequential Composition operator

- Sequential composition: take the output of the first operator and give it as input to the second operator.

- Correctness³:

$$\text{produce (comp_op } op_1 \text{ } op_2) \text{ } lxs = \text{produce } op_2 \text{ (produce } op_1 \text{ } lxs)$$

- Proof: coinduction principle for lazy list equality and produce_1 induction principle

³This is a simplification of the lemma in the paper!

Time-Aware Operators

- Time-Aware lazy lists

```
datatype ('t::order, 'd) event = DT (tmp: 't) (data: 'd) | WM (wmk: 't)
```

- Time-Aware lazy lists

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datatype ('t::order, 'd) event = DT (tmp: 't) (data: 'd) | WM (wmk: 't)
```

- Generalization to partial orders
 - Cycles
 - Timely Dataflow: uses pairs to count data cycles (e.g. *(data round n, at cycle i)*)
 - Operators with multiple inputs
 - sum type: represents the multiple input/output ports
- Productive and monotone streams: Coinductive predicates over lazy lists of events.

Proving histogram correct: building Blocks

- Histogram operator: batching and incremental computing
- Building Blocks: reusable operators
 - Batching: `batch_op`
 - Incremental computing: `incr_op`
 - Soundness, completeness, preservation of monotonicity and productivity
- Define histogram using the building blocks
- Compositional Reasoning: correctness follows from the correctness of the building blocks
 - Building blocks: more than 9000 loc
 - Incremental Histogram: 300 loc

Batching Operator

The *batch_op* operator⁴:

```
fun batches where batches [] tds = ([], tds)
| batches (wm # wms) tds = let (bs, tds') = batches wms [(t, d) ← tds. ¬ t ≤ wm]
  in (DT wm [(t, d) ← tds. t ≤ wm] # bs, tds')
```

corec batch_op **where**

```
batch_op tds = Logic (λev. case ev of DT t d ⇒ (batch_op (tds @ [(t, d)]), [])
| WM wm ⇒ let (out, tds') = batches [wm] tds
  in (batch_op tds', [x ← out. data x ≠ []] @ [WM wm]))
```

```
mono lxs W → DT wm b ∈ lset (produce (batch_op tds) lxs) →
(∀t ∈ set_t b. coll b t = lcoll lxs t + coll tds t) ∧
set_t b = batch_ts ((map (λ (t,d). DT t d) tds) @@ lxs) wm
```

```
mono lxs W → (∀t ∈ set_t tds. ∀wm ∈ W. ¬ t ≤ wm) → mono (produce (batch_op tds) lxs) W
```

⁴Another simplification from the paper

Conclusion

- Isabelle/HOL has a good tool set to formalize and reason about stream processing:
 - Codatatypes, coinductive predicates, corecursion with friends, reasoning up to friends (congruence),
 - Coinduction up to congruence principle is automatically derived for codatatypes (but not for coinductive principles)
- Next step: Feedback loop

Questions, comments and suggestions