

On Verifying Causal Consistency

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Geo-Replicated Data Structures

- Strong (sequential) consistency



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Geo-Replicated Data Structures

- Strong (sequential) consistency is **impossible** while being **available** and tolerating **network partitions**: the CAP theorem ¹



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- Tolerating **faults** while preserving **availability** leads to **anomalies** w.r.t. strong (sequential) consistency



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Updates are seen in different orders

Goal: Verifying Causal Consistency

The set of allowed anomalies are defined by **weak consistency** criteria, e.g., eventual consistency, causal consistency.

Algorithmic methods for checking **causal consistency**.

Single-Trace Verification: Check if **one trace** is causally consistent

- Application to testing, monitoring (by enumerating traces)

All-Traces Verification: Check if **all traces** are causally consistent

- Static verification

Comparison with other Consistency Criteria

Single-Trace Verification:

- **NP-complete** for most consistency criteria²

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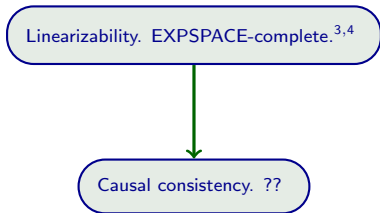
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Sequential consistency. Undecidable.¹



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(Input: **finite-state automaton** representing all traces)

Key Observation: Implementations Are Data Independent

Key-value store implementations are **data independent**

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⇒ **Writes can be assumed to be unique**

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X contains **4-6** bad patterns

Results: Complexity/Decidability and Reduction to Reachability

Bad patterns implications for **data-independent** implementations:

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*Single-Trace Verification of causal consistency is **polynomial** when **writes are unique**.*

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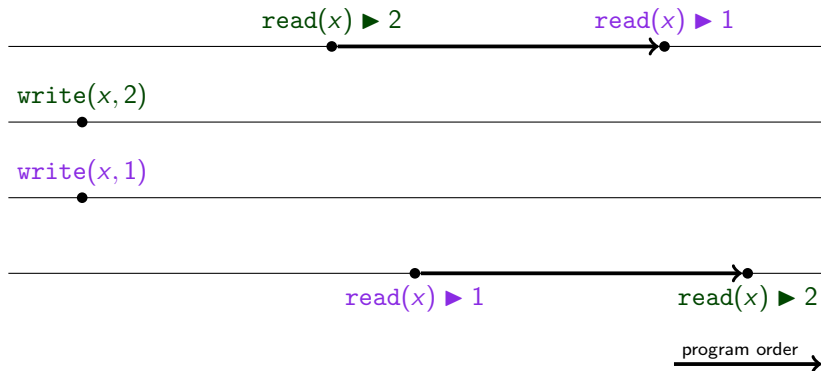
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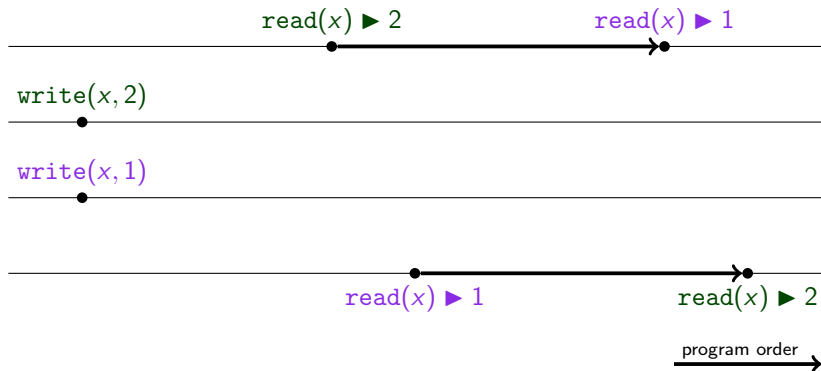
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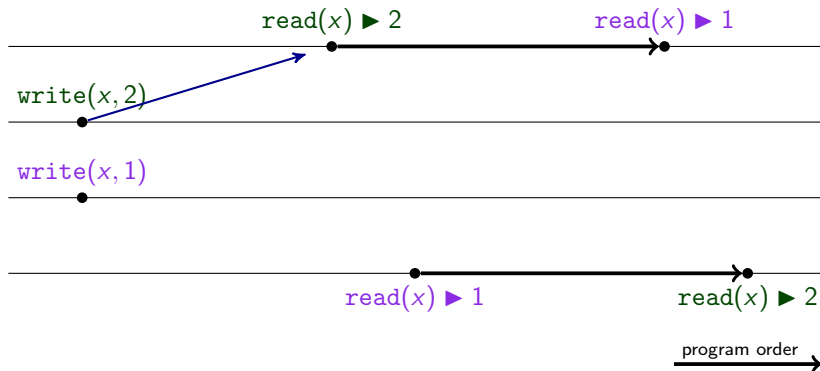


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the **causal past** of **every read** can explain its value

CO includes the program (site) order

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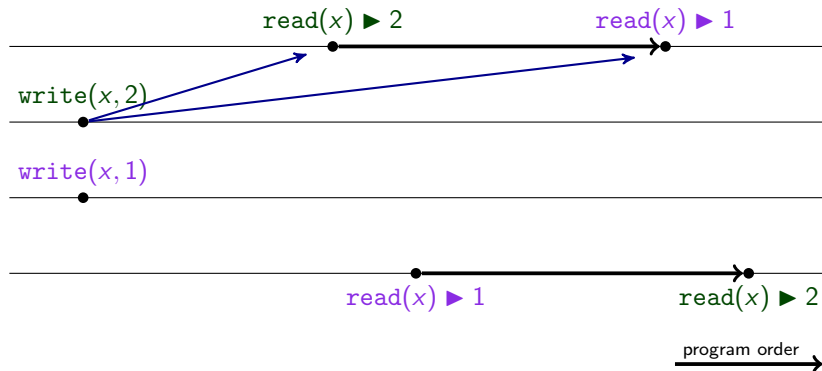


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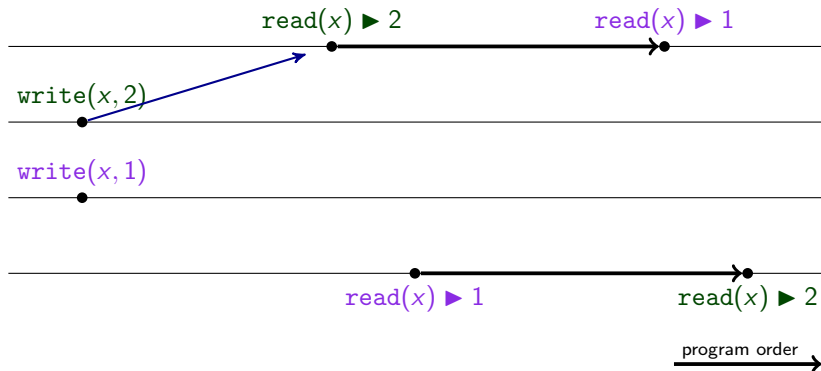


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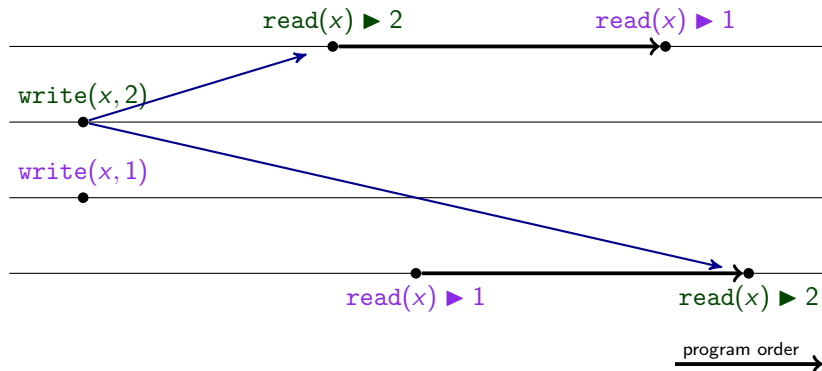


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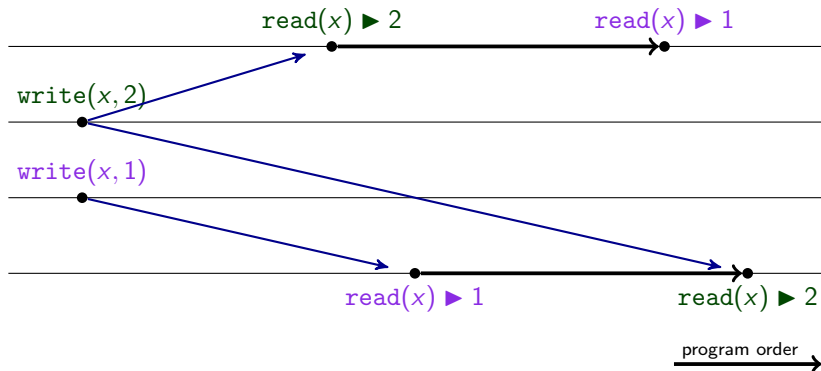


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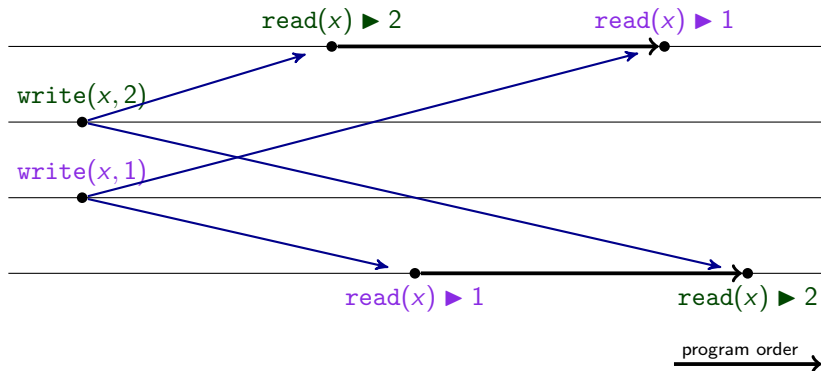


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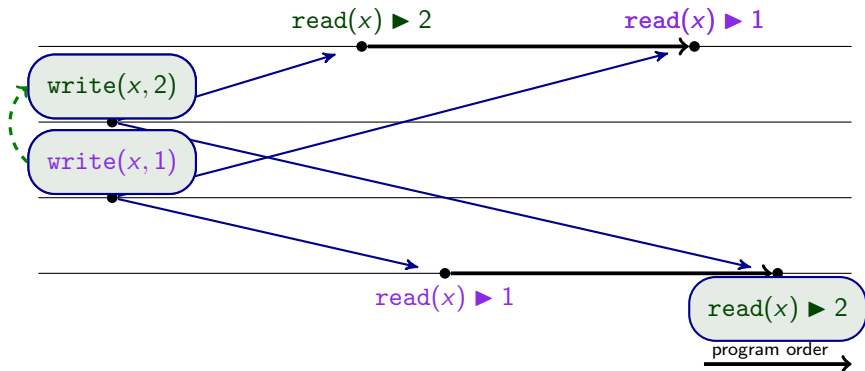


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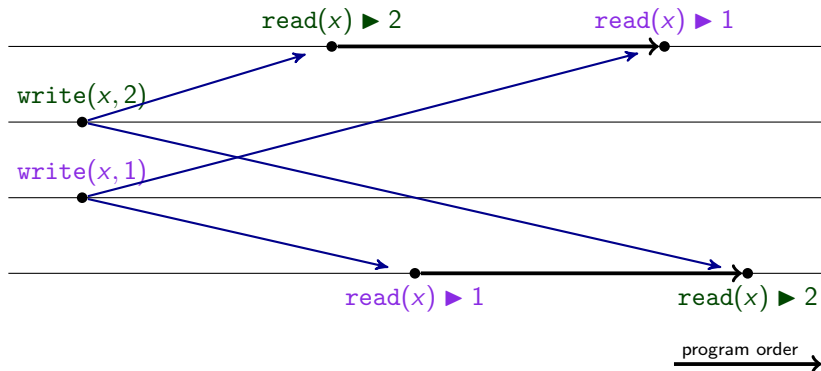


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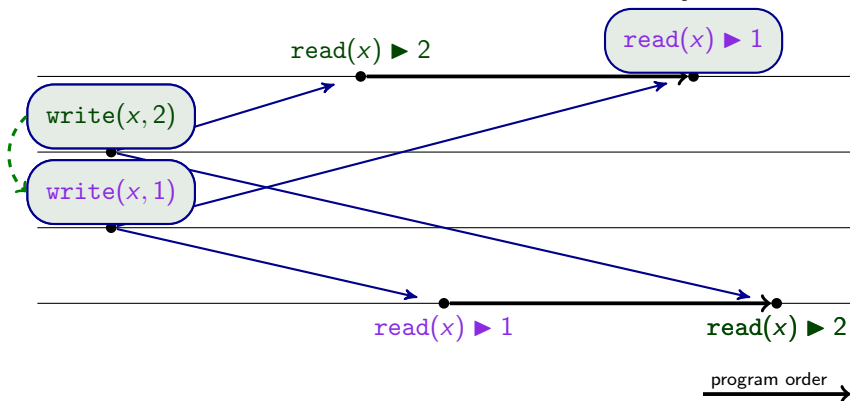


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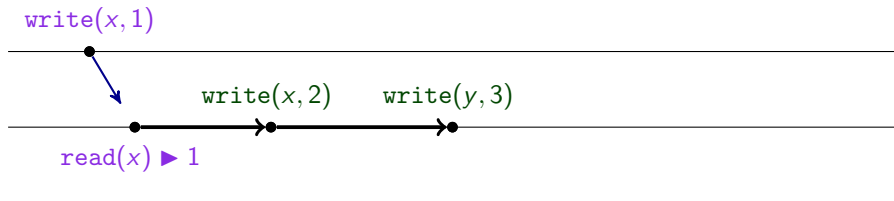
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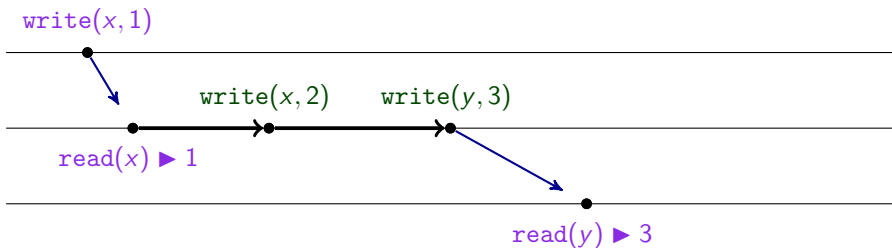
Causal Consistency Violations

Causally related writes must be seen by all sites in the same order.



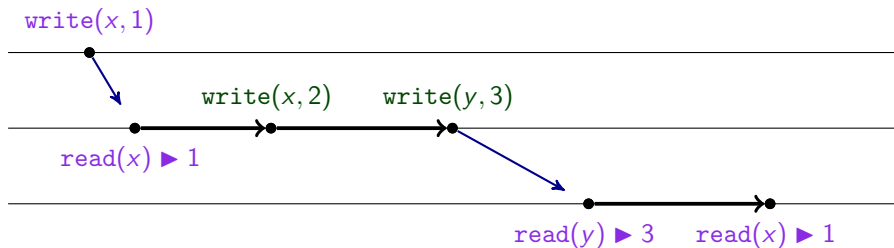
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Formalizing Causal Consistency

Specification = a set of sequences of operations

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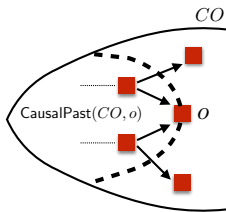
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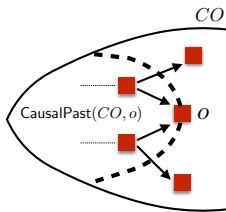
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\sqsubseteq means “can be linearized to”)

Causal Convergence⁸

- Conflicts are resolved using a global **arbitration order**
- **Strong eventual consistency:**
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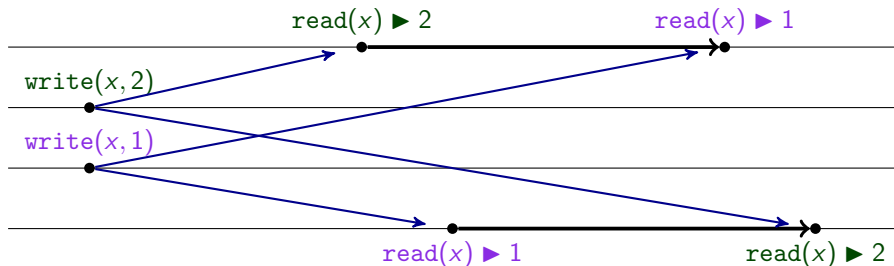
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Causal Convergence

A history $h = (O, PO)$ is **causally convergent** w.r.t. a **specification S** iff there exists a strict **partial order CO** and a strict **total order ARB (arbitration)** s.t.

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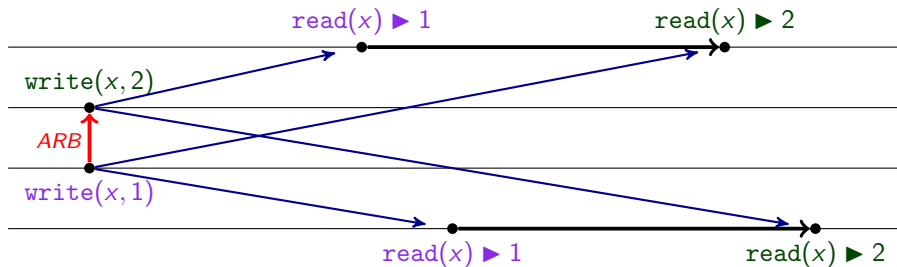
$$AX_{ARB} : CO \subseteq ARB$$

$$AX_{CAUSALARB} : \forall o \in O. \text{CausalPast}(CO, o) \oplus ARB \in S$$

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" $\oplus ARB$ " means adding the constraints in ARB)

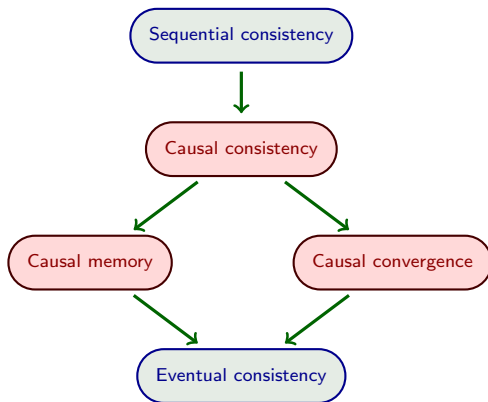
Satisfying Causal Convergence



Satisfying Causal Convergence but not Sequential Consistency



Different Notions of Causal Consistency



Causal memory = Causal consistency + local arbitration

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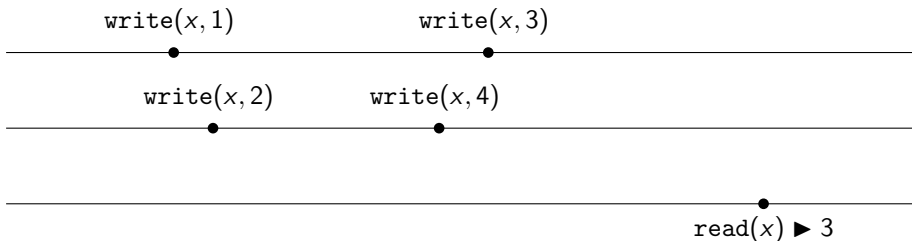
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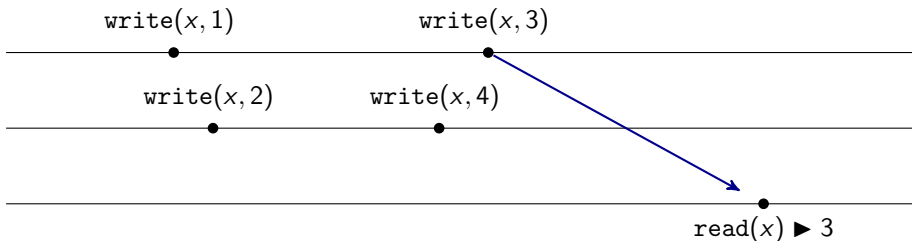
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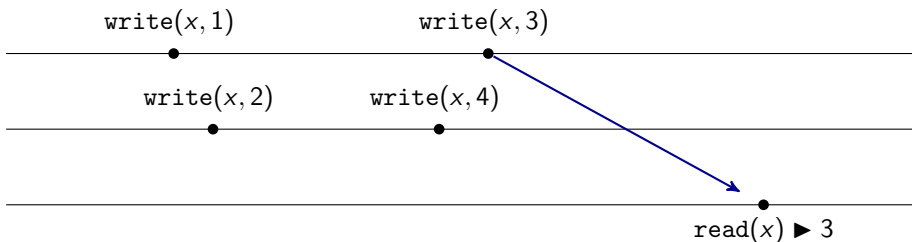
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Unicity of writes implies a **canonical causality** relation (included in every other causality relation).

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- *PO* (program order): connects operations from the same site
- *RF* (reads-from relation): connects write to read
- *CO* (causal order): defined as $(PO \cup RF)^+$

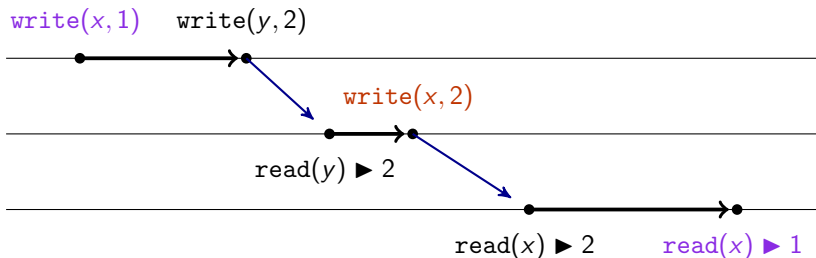
Bad Pattern for Causal Consistency: WriteCORead

- Two writes w_1 and w_2 , and one read r_1 on the same variable:
 - r_1 reads-from w_1
 - $w_1 <_{CO} w_2 <_{CO} r_1$

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Example:



WriteCOWRead: Litmus tests

$w_1 <_{PO} w_2 <_{PO} r_1$:

`write(x, 1)`

`write(x, 2)`

`read(x) ▶ 1`

WriteCORead: Litmus tests

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write(x, 1)

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write(y, 3)

||

read(y) ▶ 3

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$w_1 <_{PO} w_2 <_{CO} r_1$:

write(x, 1) read(y) ▶ 3
write(x, 2) || read(x) ▶ 1
write(y, 3)

$w_1 <_{CO} w_2 <_{PO} r_1$:

write(x, 1) read(y) ▶ 3
write(y, 3) || write(x, 2)
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$w_1 <_{CO} w_2 <_{CO} r_1$:

write(x, 1) read(y) ▶ 3 read(z) ▶ 4
write(y, 3) || write(x, 2) || read(x) ▶ 1
 write(z, 4)

Bad Patterns for Causal Consistency

- **WriteCORead**: two writes w_1 and w_2 , and one read r_1 on some x s.t.
 - r_1 reads-from w_1
 - $w_1 <_{CO} w_2 <_{CO} r_1$
- **CyclicCO**: $CO = (PO \cup RF)^+$ is cyclic
- **ThinAir**: a read operation $r = \text{read}(x) \blacktriangleright v$ with $v \neq 0$ s.t.
 - $w \not\prec_{RF} r$ for every write w
- **WriteCOInit**: a read operation $r = \text{read}(x) \blacktriangleright 0$ s.t.
 - $w <_{CO} r$ for some write w on x

Bad Patterns for Causal Consistency Variants

Causal Consistency	Causal Memory	Causal Convergence
CyclicCO	CyclicCO	CyclicCO
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Theorem (Bad Patterns)

A trace doesn't satisfy the criterion **X** iff it contains a **bad pattern** for **X**.

Outline

- Definition(s) of **causal consistency**
- Characterize **all causal consistency violations** using bad patterns
- Using bad patterns for verifying data-independent implementations
 - Single-Trace Verification: **polynomial time**
 - **Bad patterns** can be recognized with **state machines**
 - **Generic reduction** from causal consistency to **reachability**
 - All-Traces Verification: **decidable**

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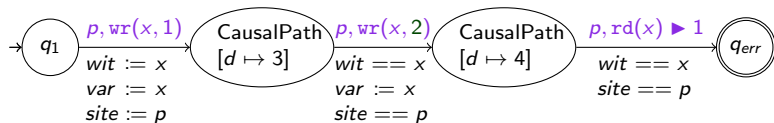
Theorem (Single-Trace Verification)

*Single-Trace Verification of causal consistency is **polynomial** when **writes are unique**.*

(By checking the absence of bad patterns.)

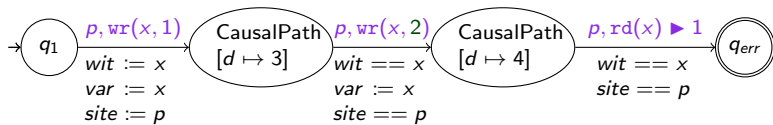
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- Registers are needed to store variable names while tracking causality paths
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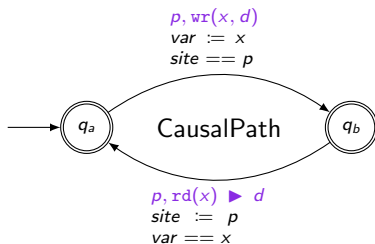


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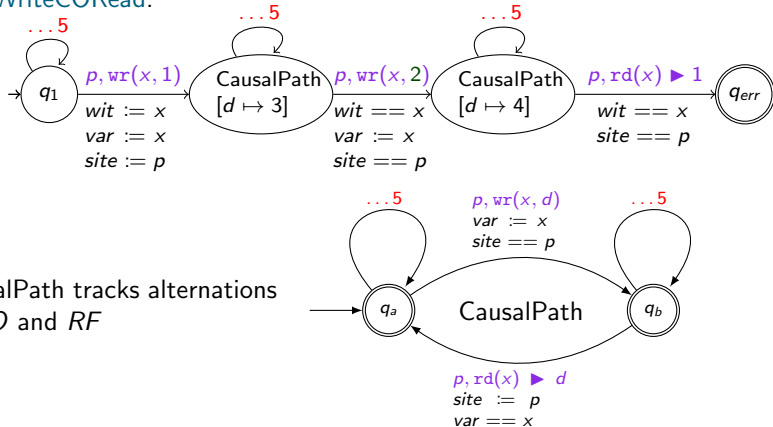


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Machine M tracking all bad patterns.

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Thank you