

Embedded Software: Better Models, Better Code

Tom Henzinger

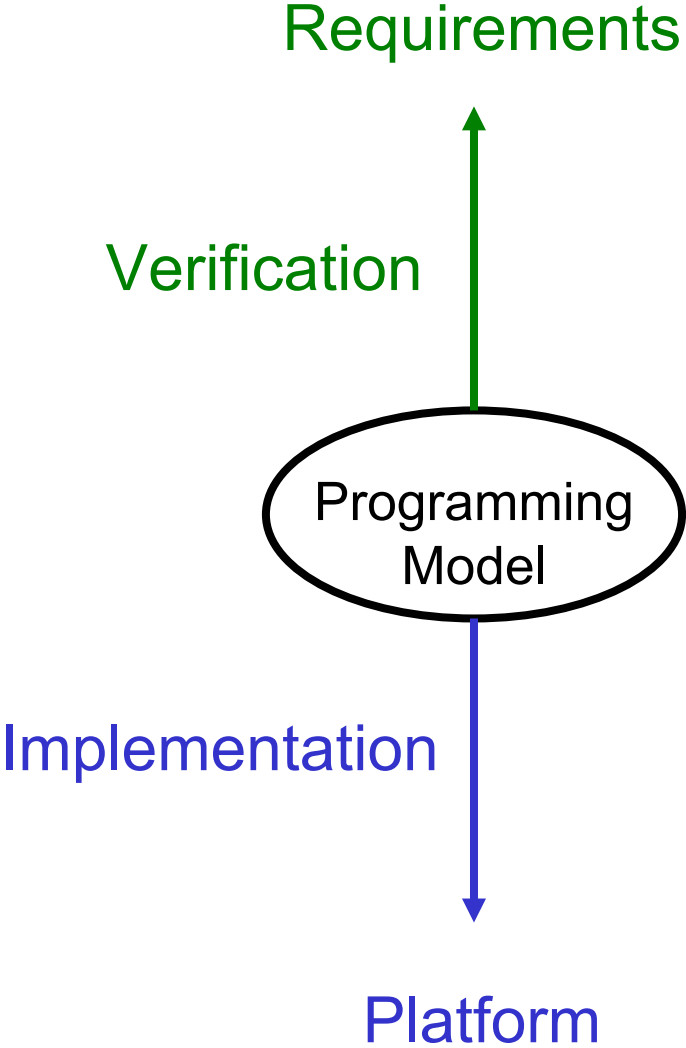
University of California, Berkeley

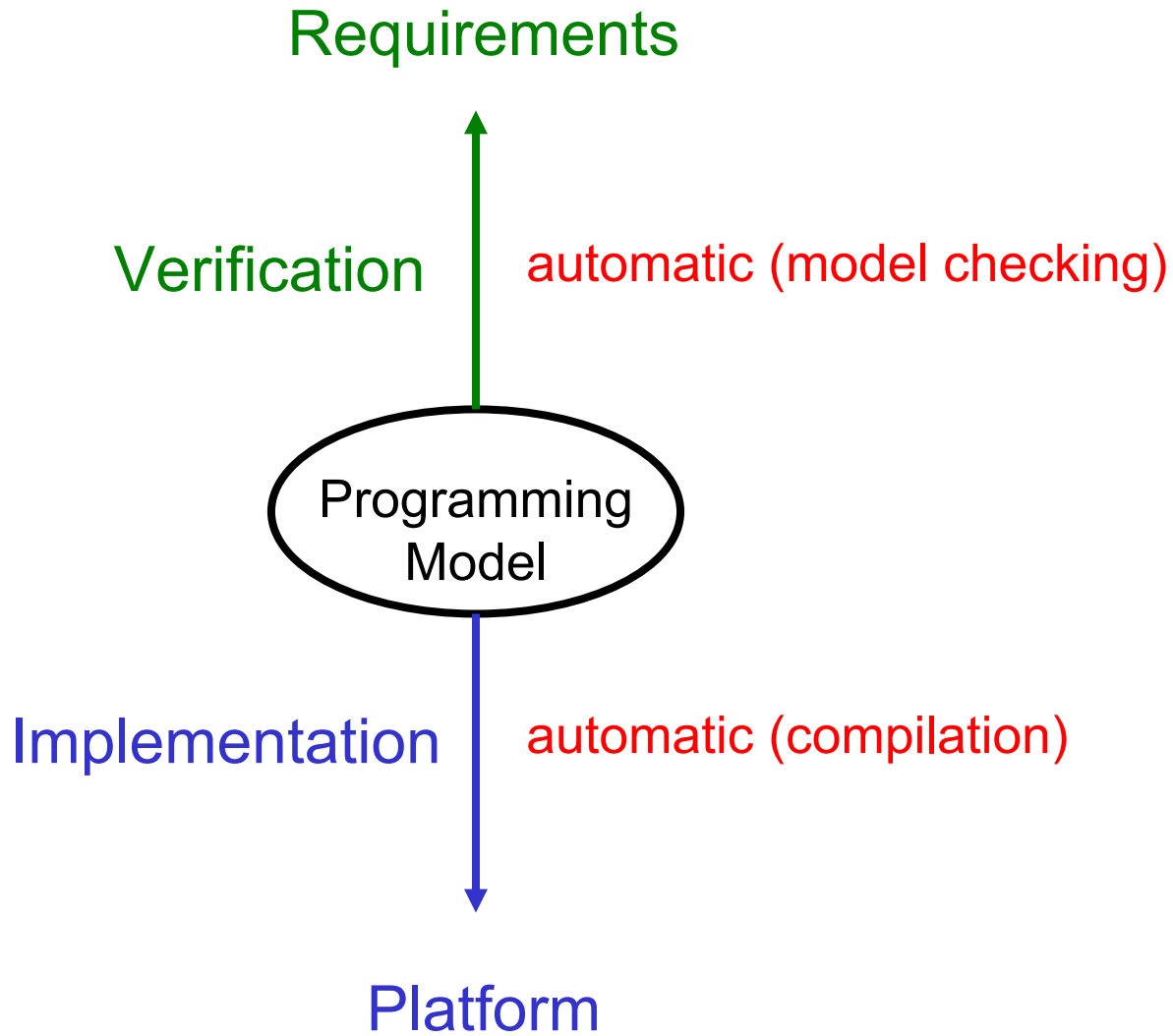
Our Research Explores Three Paradigms

In **modeling**, use discounted quantitative measures.

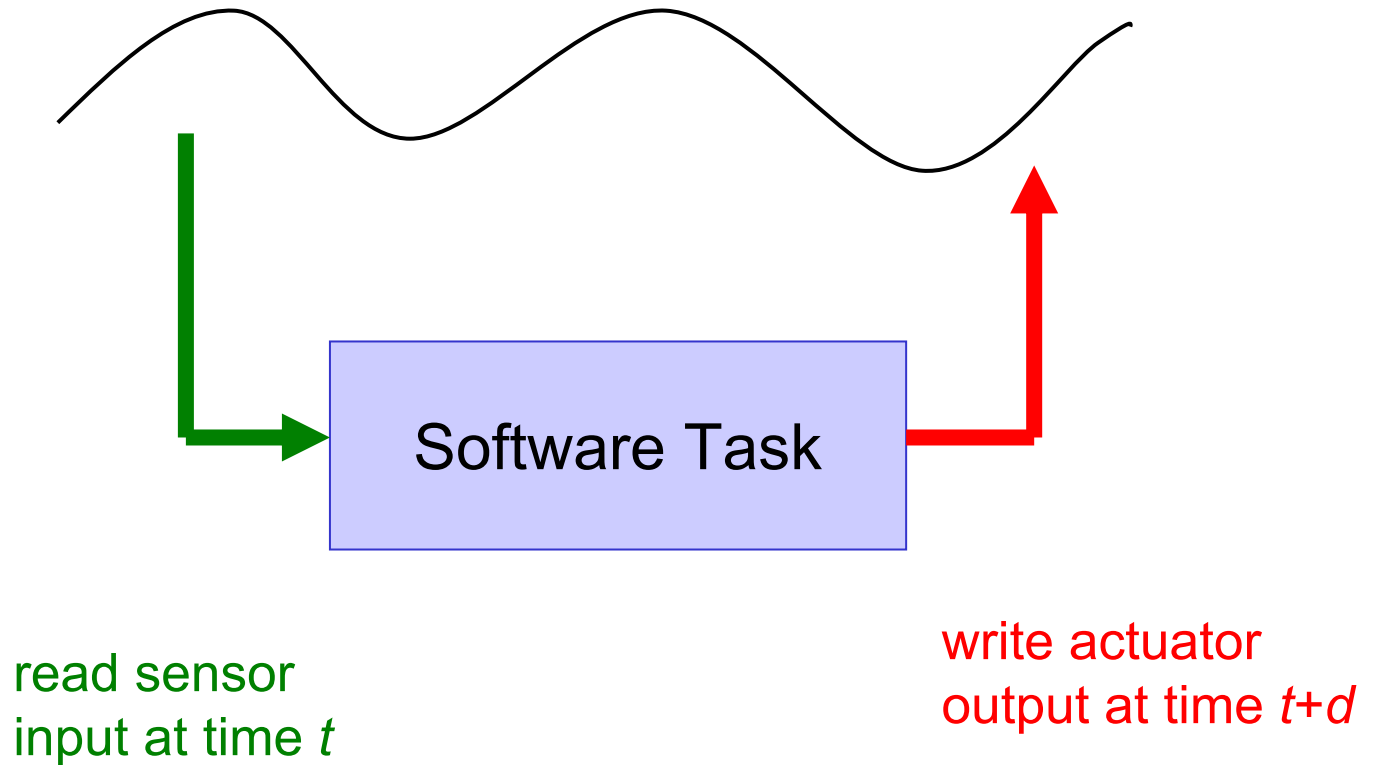
In **composition**, treat inputs and outputs contra-variantly.

In **implementation**, preserve logical execution times.

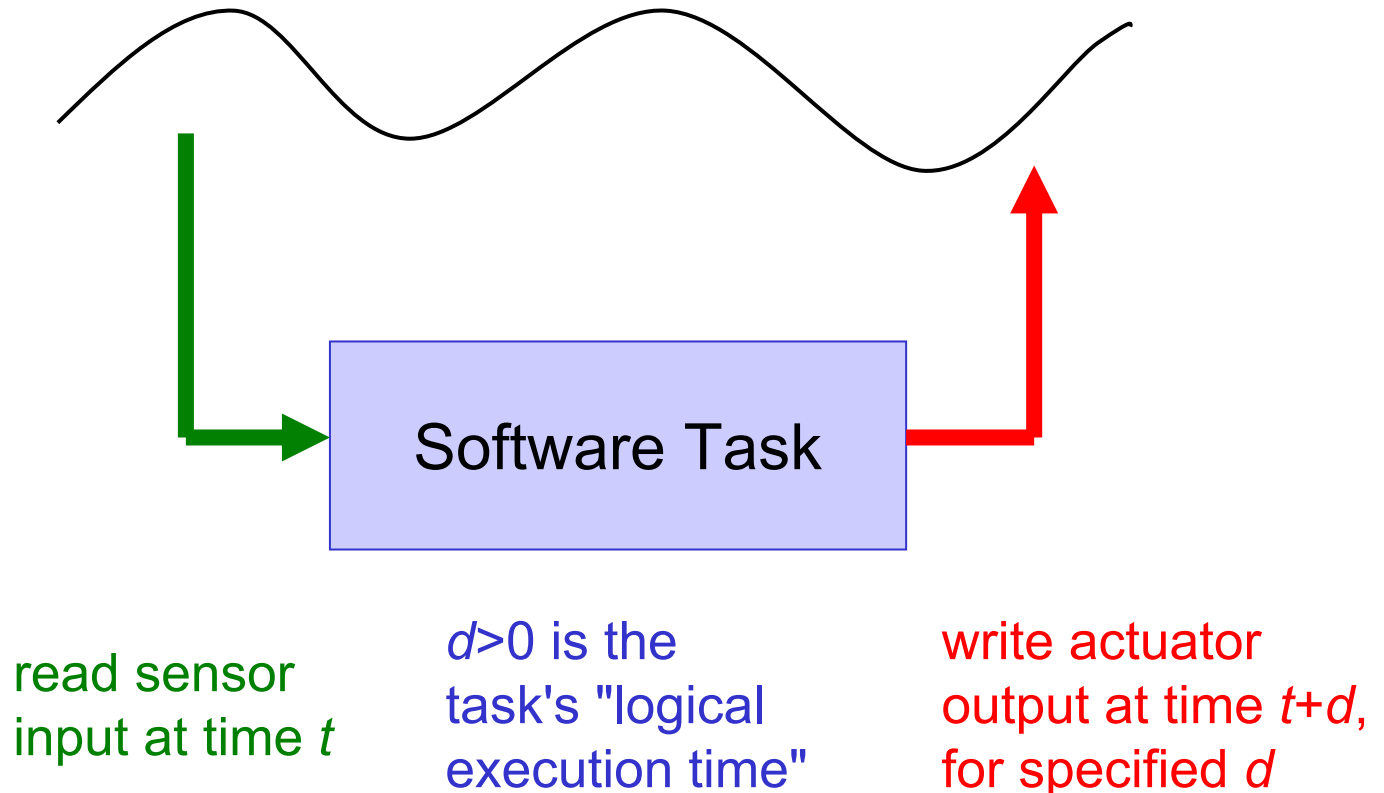




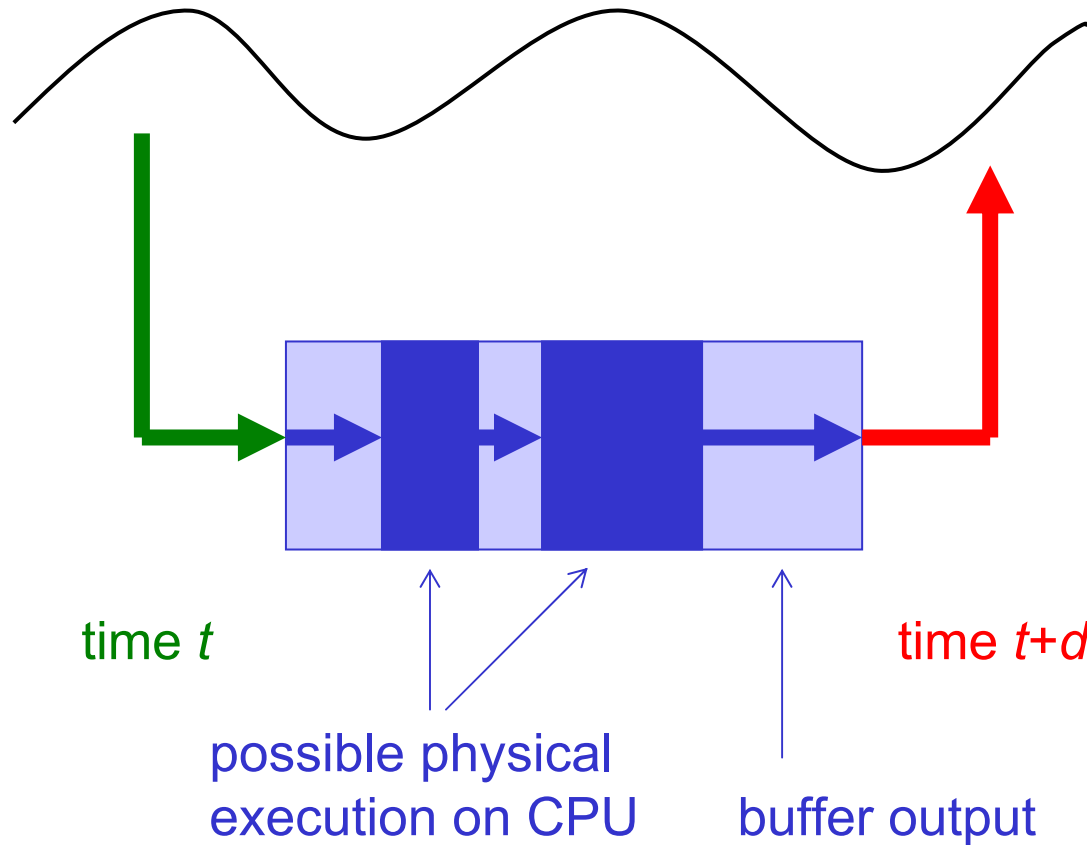
The LET (Logical Execution Time) Assumption



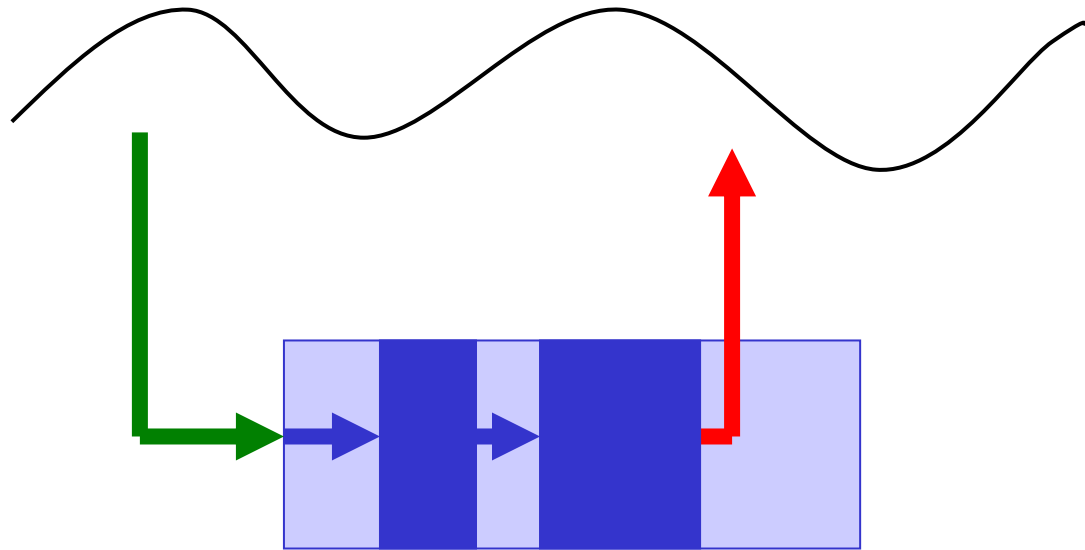
The LET (Logical Execution Time) Assumption



The LET (Logical Execution Time) Assumption



Contrast LET with the Standard Practice



output as soon
as ready

LET-based Real-Time Programming

The **programmer** specifies d to solve the control problem at hand.

The **compiler** ensures that d is met on a given platform; otherwise it rejects the program.

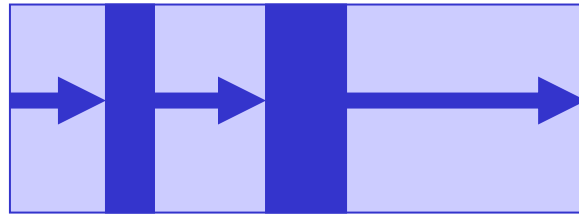
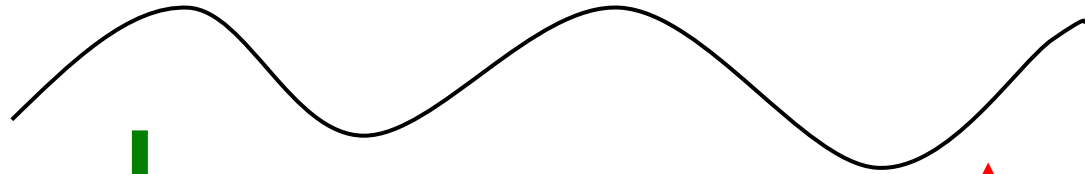
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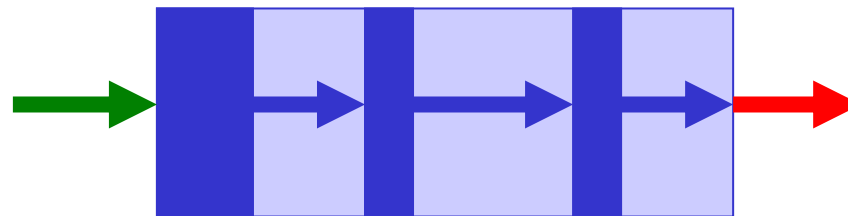
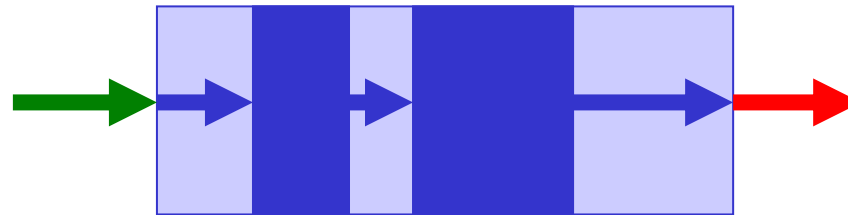
The proof that d is met (proof of “**time safety**”) produces a schedule → **Schedule Carrying Code** [H, Kirsch, Matic].

Portability

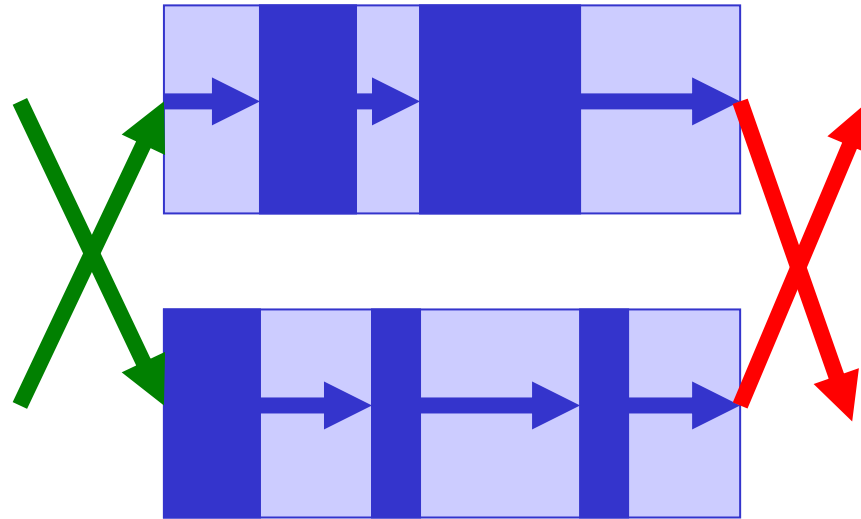


50% CPU speedup

Composability



Predictability / Verifiability

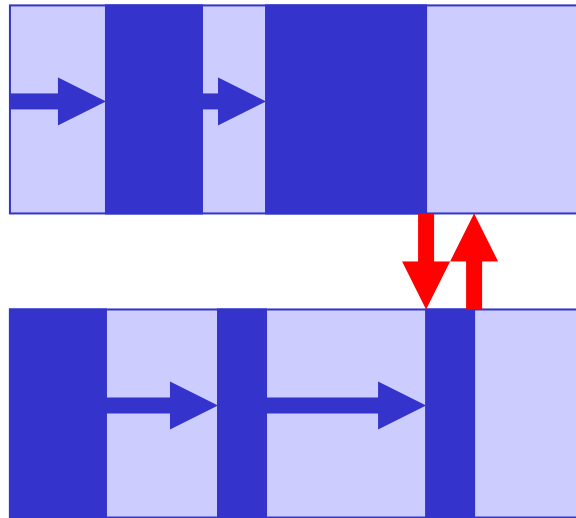


-timing predictability: minimal jitter

-value predictability: no race conditions

Environment determined behavior!

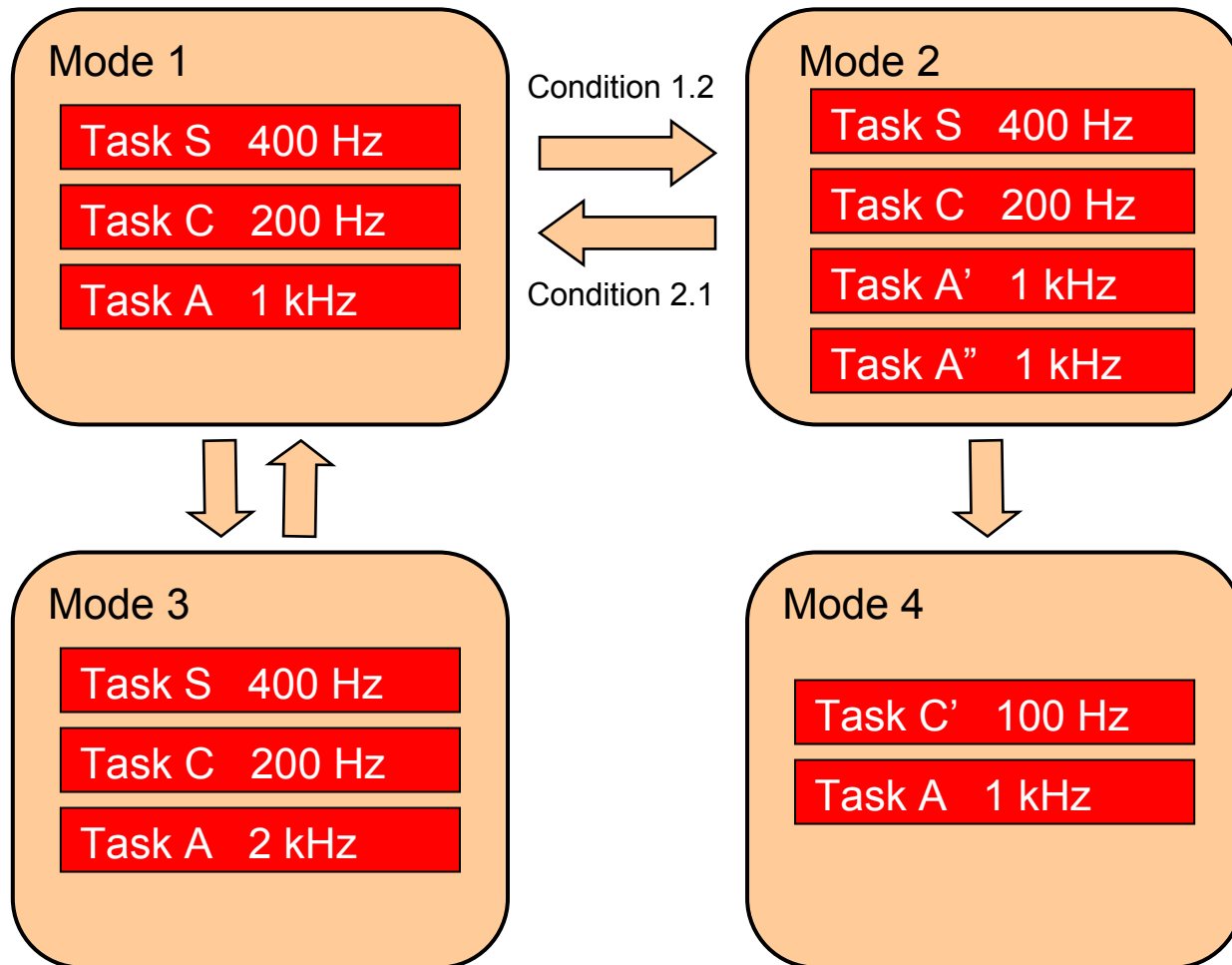
Contrast LET with the Standard Practice



Race

GIOTTO:

LET for periodic tasks with time-triggered mode switching



xGIOTTO:

Event-Triggered LET Programming [Ghosal, H, Kirsch, Sanvido]

If all events can happen at any time, then few programs would be time-safe.

However, nested reaction blocks can specify the selective listening to events (“**event scoping**”) → **Structured LET Programming**.

xGIOTTO:

Event-Triggered LET Programming [Ghosal, H, Kirsch, Sanvido]

1. Schedule Instruction:

```
schedule Task by Event ;
```

↑
logical deadline

2. Reaction Block:

```
react {  
  when Event do Block ;  
  whenever Event do Block ;  
  begin ... end ;  
} until Event ;
```

Our Research Explores Three Paradigms

In **modeling**, use discounted quantitative measures.

In **composition**, treat inputs and outputs contra-variantly.

In **implementation**, preserve logical execution times.

In composition, treat inputs and outputs contra-variantly.

This seems obvious:

The "type" of a component should be

$\text{Inputs} \rightarrow \text{Outputs}$

not

$\text{Inputs} \times \text{Outputs}$.

(These two are the same in set theory,
but not in type theory!)

In composition, treat inputs and outputs contra-variantly.

Surprisingly, this is rather non-standard:

If your notion of composition is **intersection** or **product**,

or your notion of refinement / abstraction is **simulation** or **language containment**,

then you treat inputs and outputs co-variantly
(and are in good company)!



Input constraint:
not $x=y=1$

Output constraint:
none

This is an assumption
about the environment.

This is an abstraction
of the component.



Input constraint:
not $x=y=1$

Possible behaviors:

x	y	z
0	0	0
0	0	1
0	1	0
0	1	1
1	0	0
1	0	1

Output constraint:
none

Compose with $y=z$, forgetting what is input, what output.



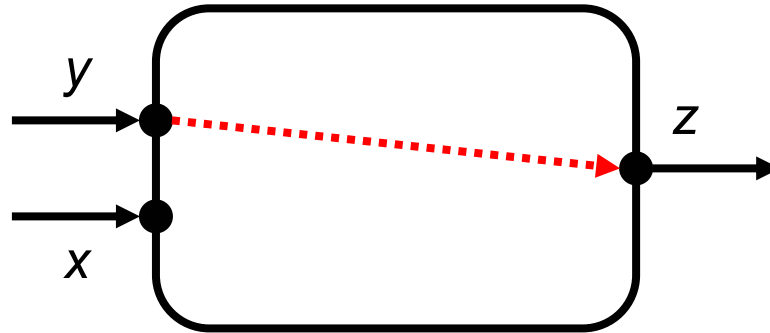
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Possible behaviors:

x	y	z
0	0	0
0	0	1
0	1	0
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Output constraint:
none

Compose with $y=z$, constraining only output (the component).



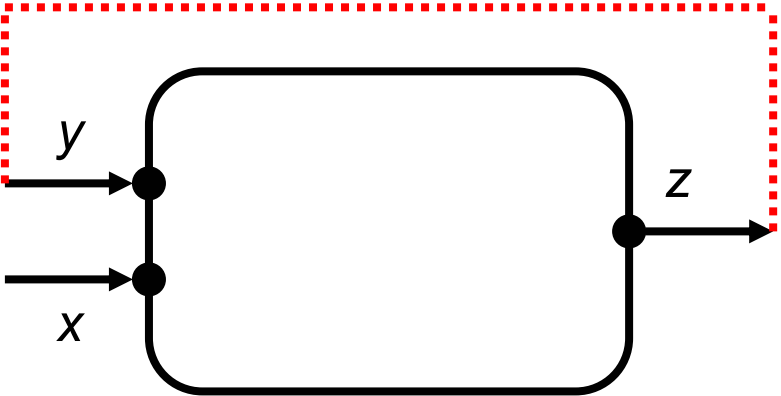
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Possible behaviors:

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1	0	0
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New output constraint:
 $z=y$

Compose with $y=z$, constraining only inputs (the environment).



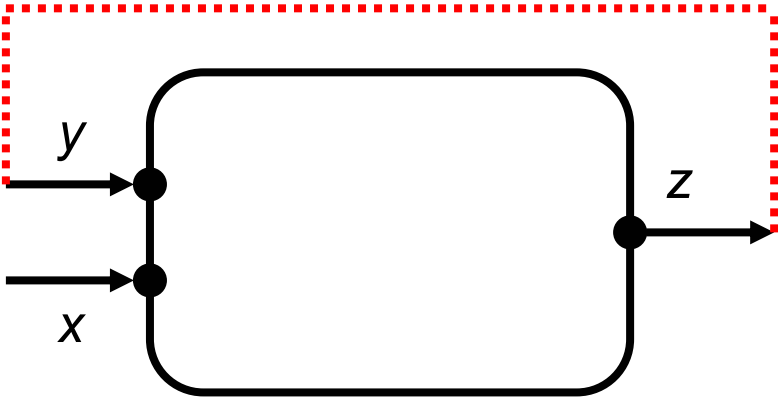
Input constraint:
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Possible behaviors:

Output constraint:
none

x	y	z
0	0	0
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Compose with $y=z$, constraining only inputs (the environment).



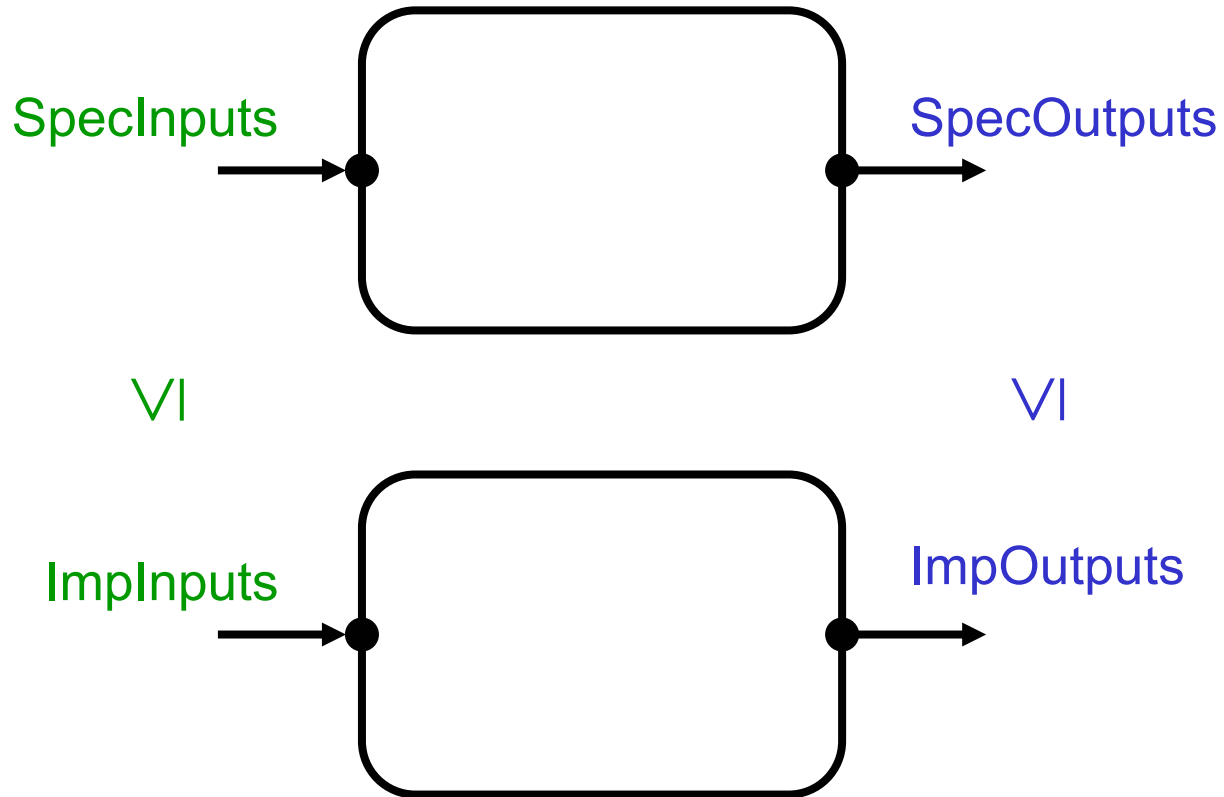
New input constraint:
 $x=0$

Possible behaviors:

Output constraint:
none

x	y	z
0	0	0
0	0	1
0	1	0
0	1	1
1	0	0
1	0	1

Traditional Behavioral Refinement: Simulation or Language Containmentment



Contra-variant Refinement: Implementations can be substituted for Specifications



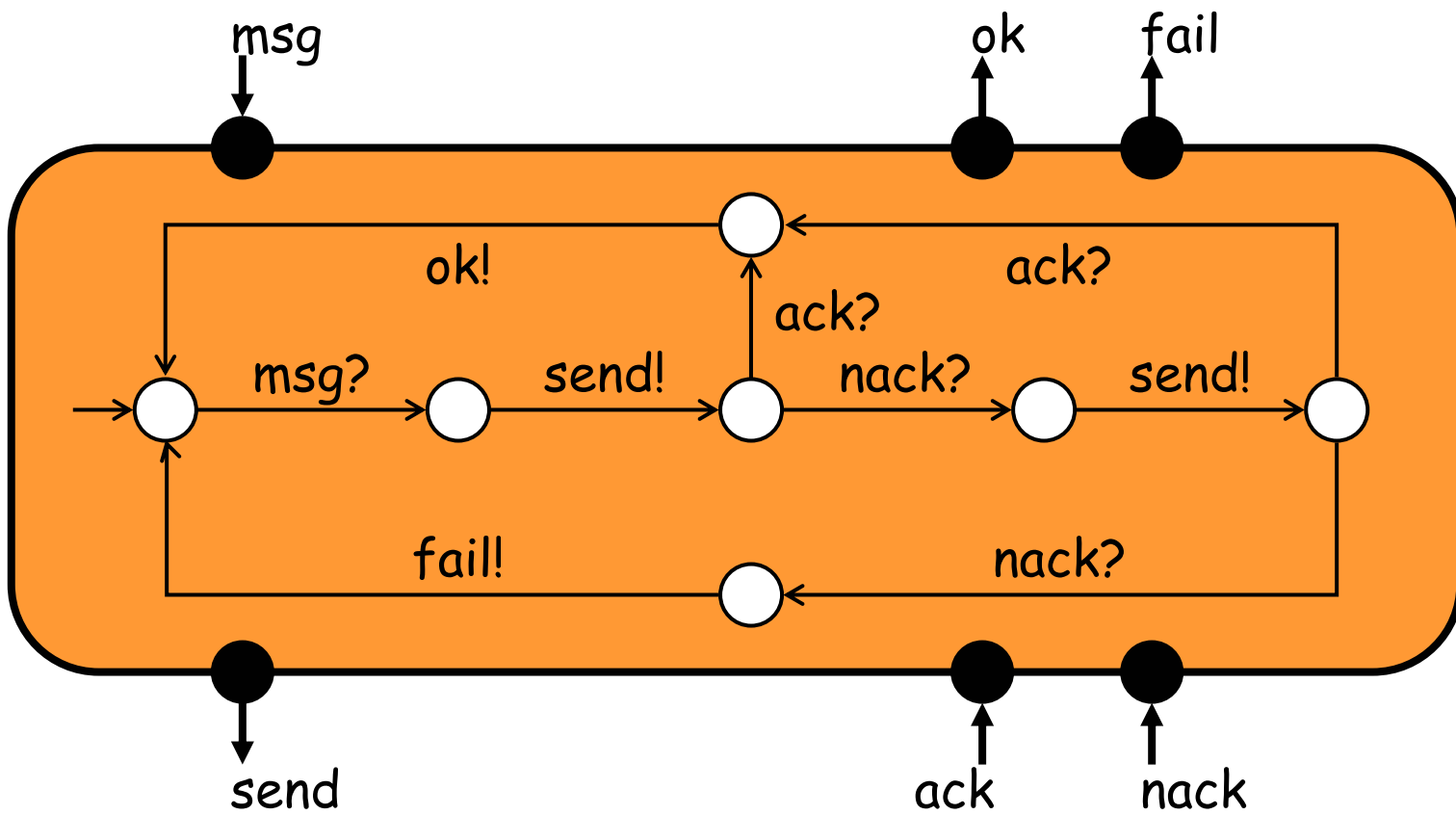
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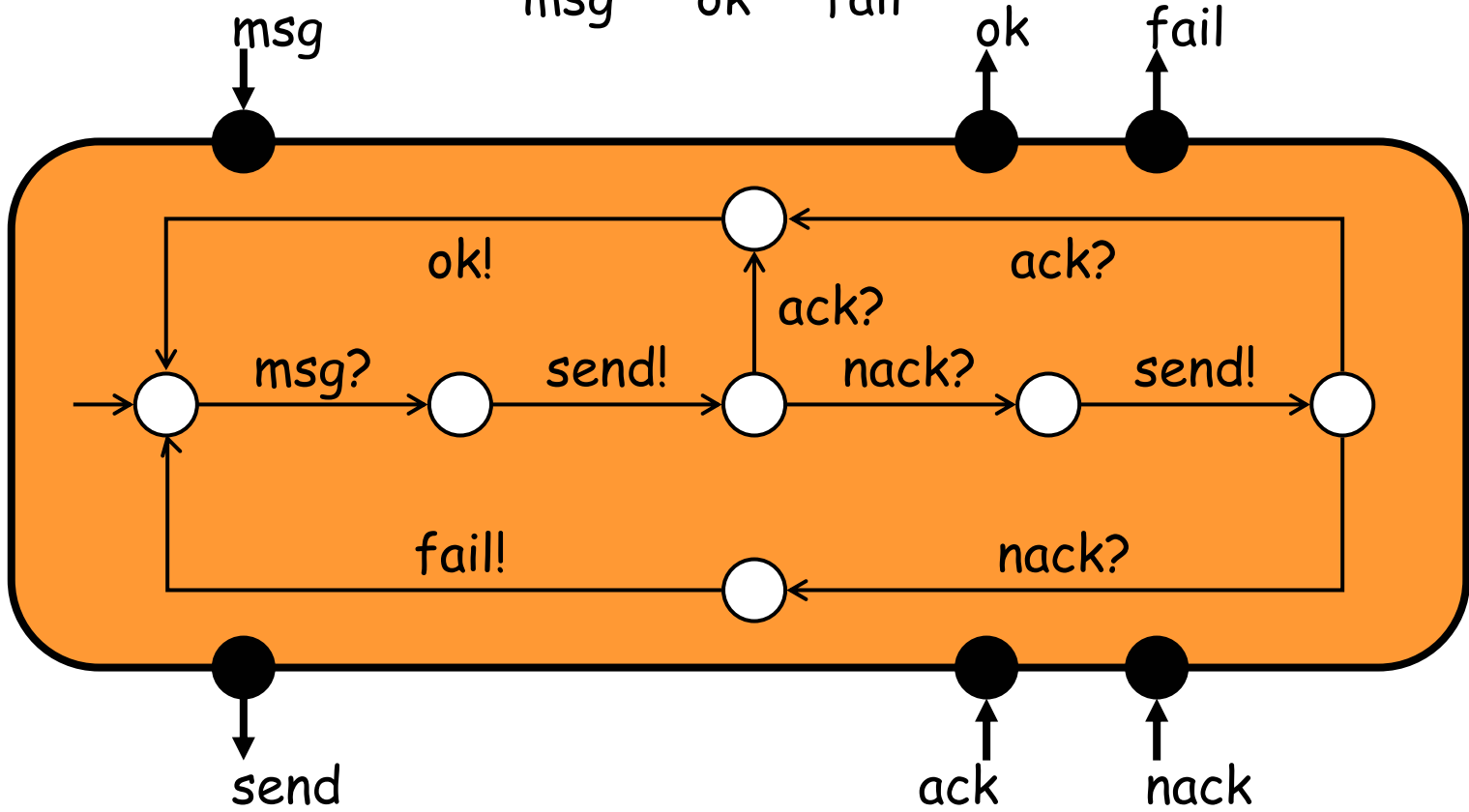
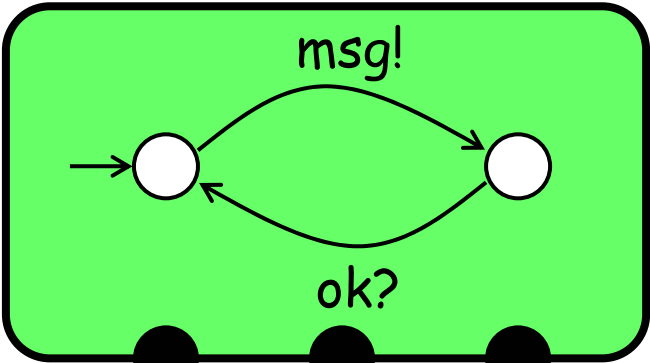
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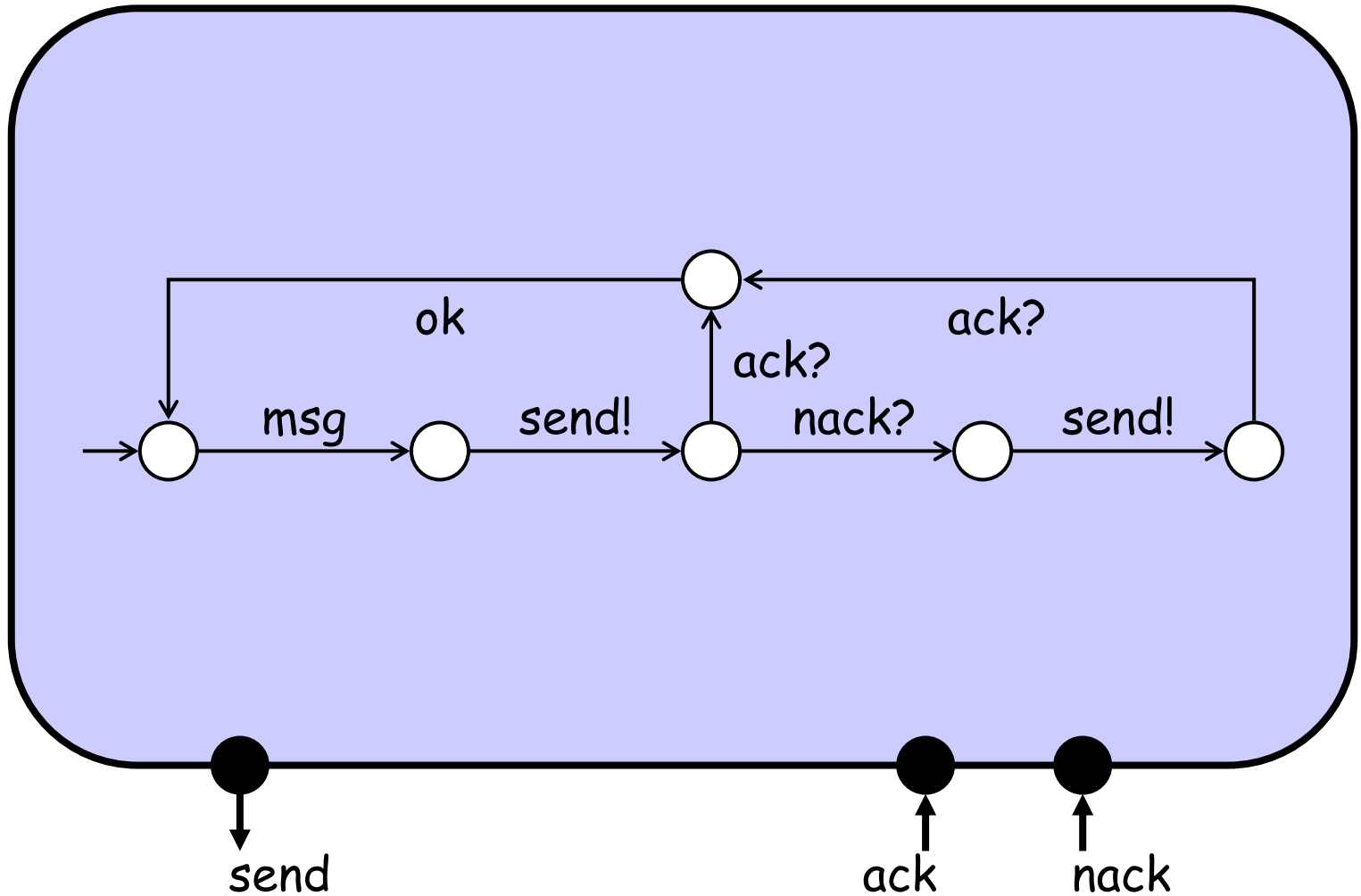
Implementation accepts all legal inputs.

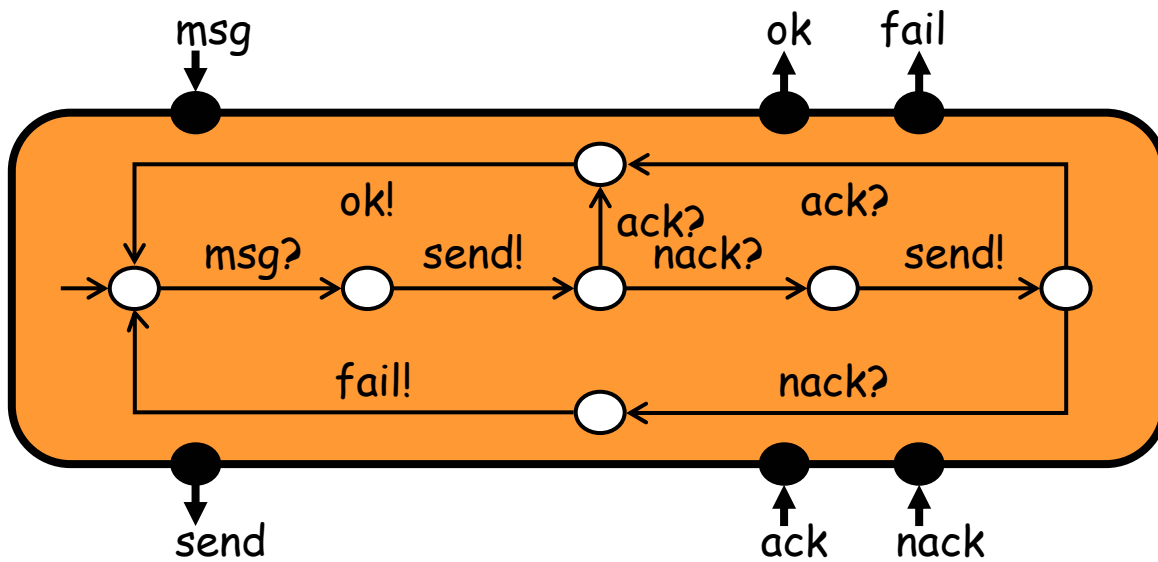
Implementation produces only legal outputs.



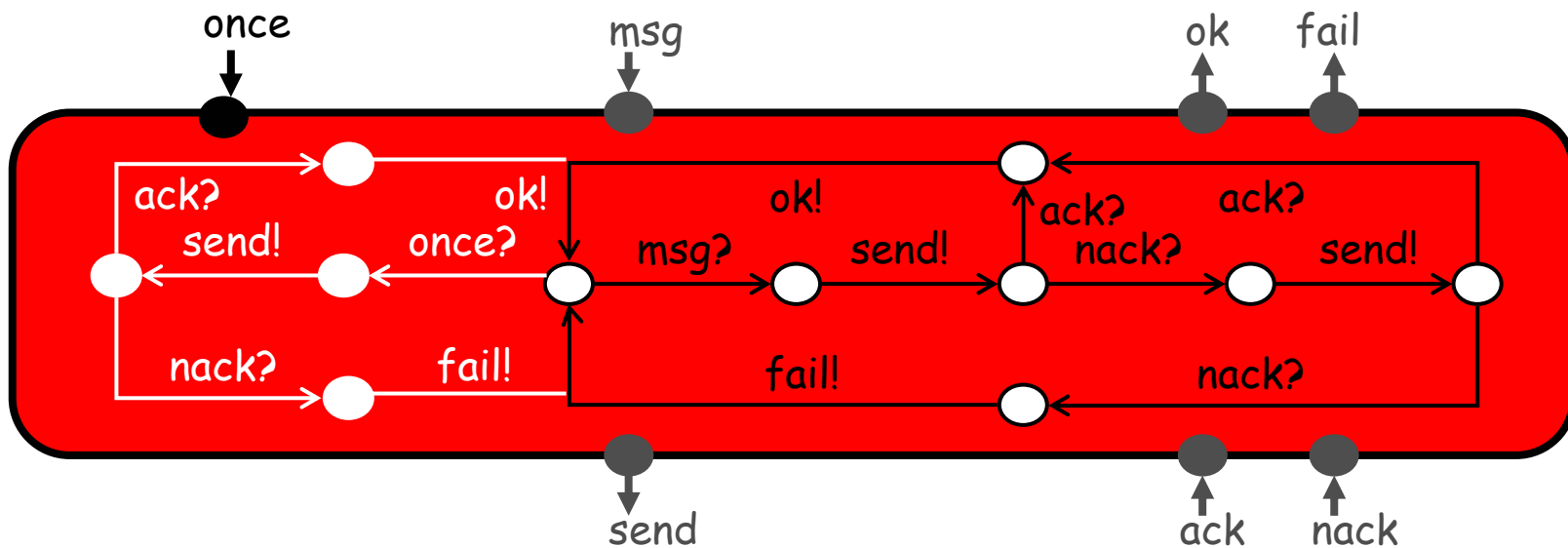


The Composite Interface





VI



We call a formalism with

-input-constraining composition

-contra-variant refinement

an **interface theory** [de Alfaro, H].

We call a formalism with

- input-constraining composition
- contra-variant refinement

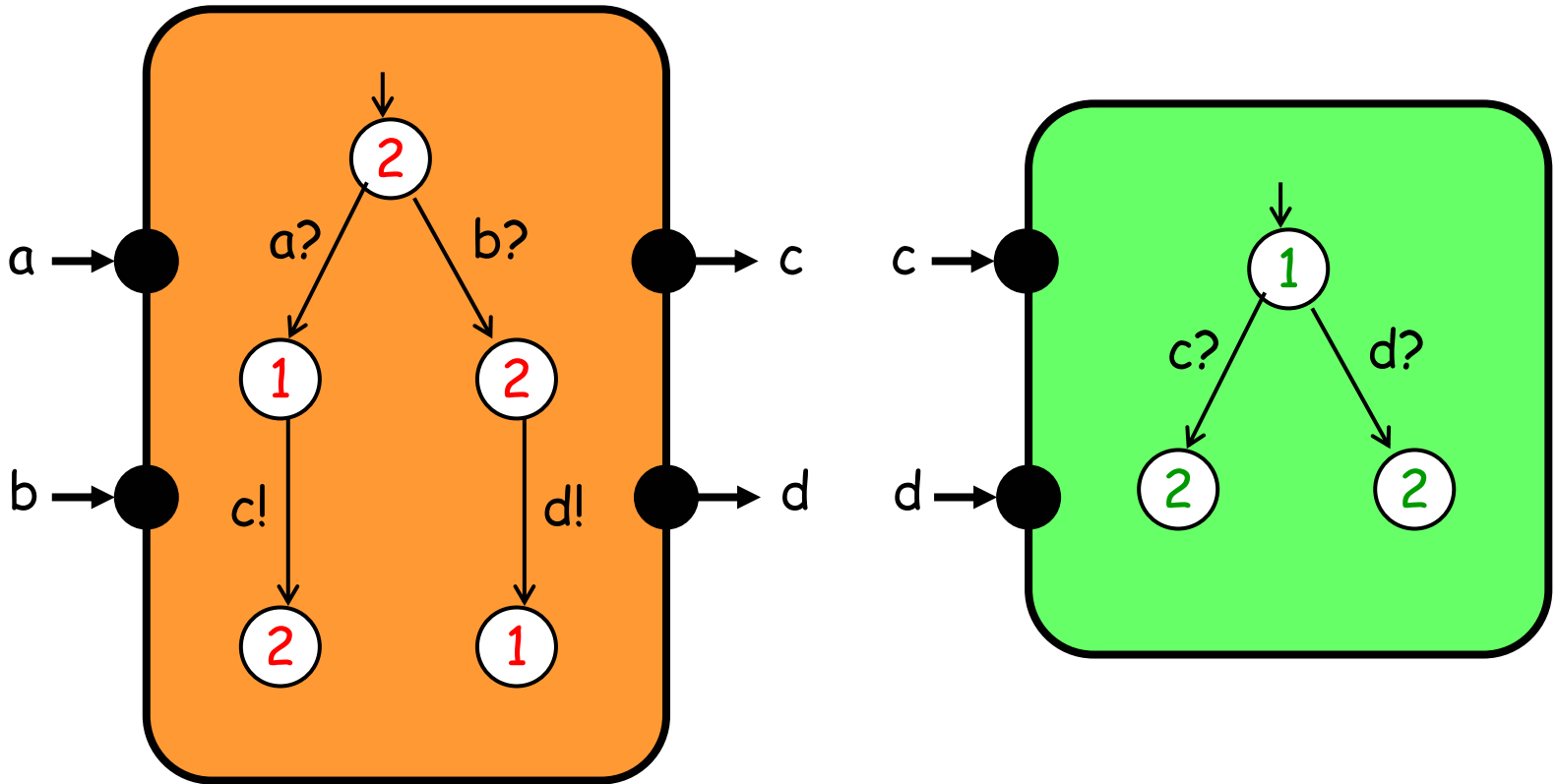
an **interface theory** [de Alfaro, H].

We have developed several interface theories, e.g. for

- message-passing components ("interface automata")
- synchronous hardware components [Chakrabarti, dA, H, Mang]
- possibly recursive software modules [C, dA, H, Jurdzinski, M]
- real-time components [dA, H, Stoelinga]
- resource-constrained components [C, dA, H, S]

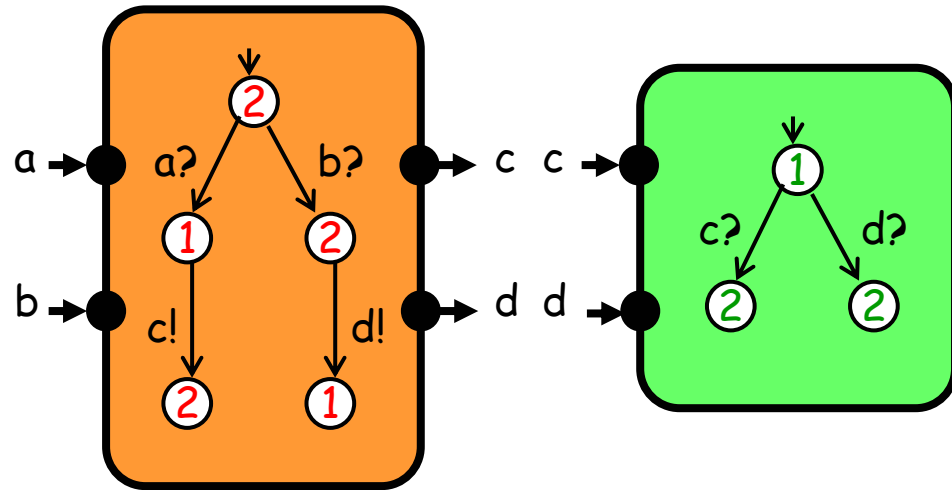
They have been implemented in the **CHIC** tool.

Resource Interfaces

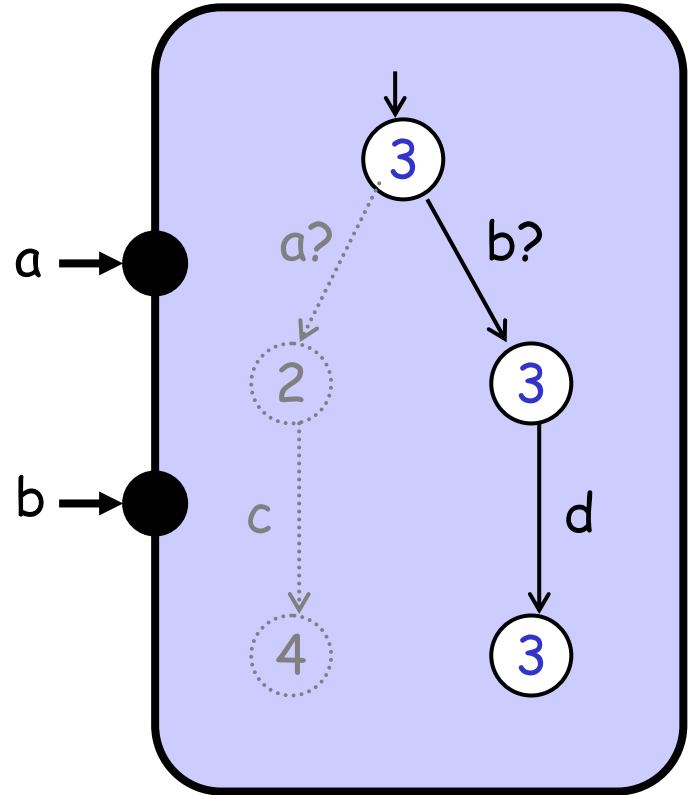


Available peak power: 3

Resource Interfaces



Available peak power: 3



The composite interface.

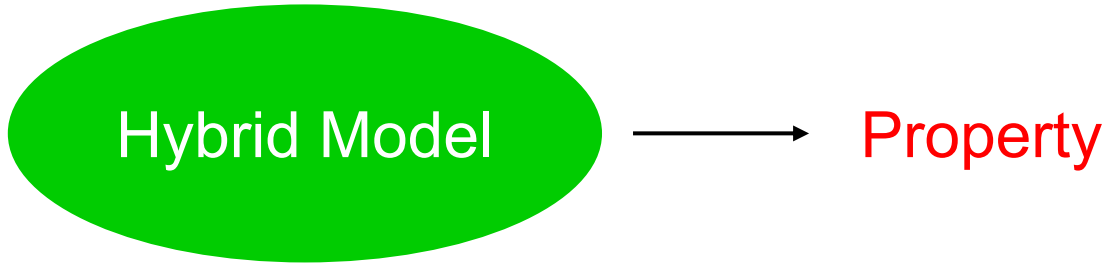
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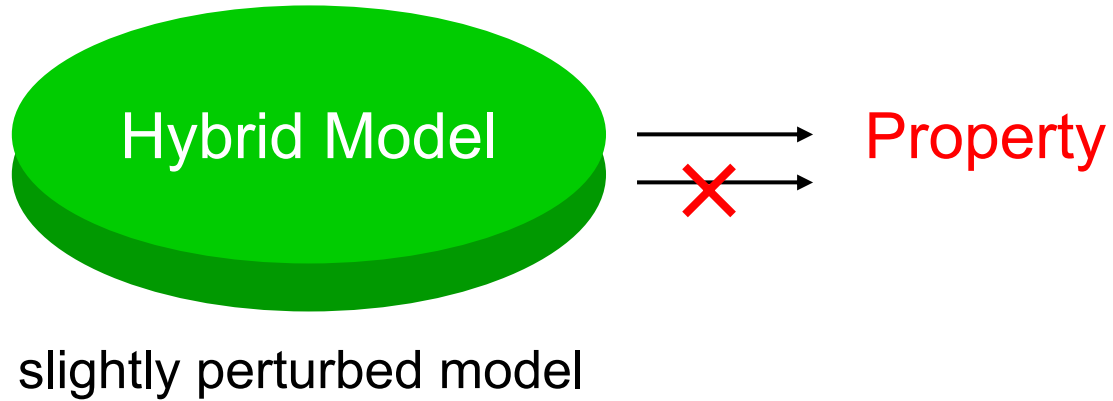
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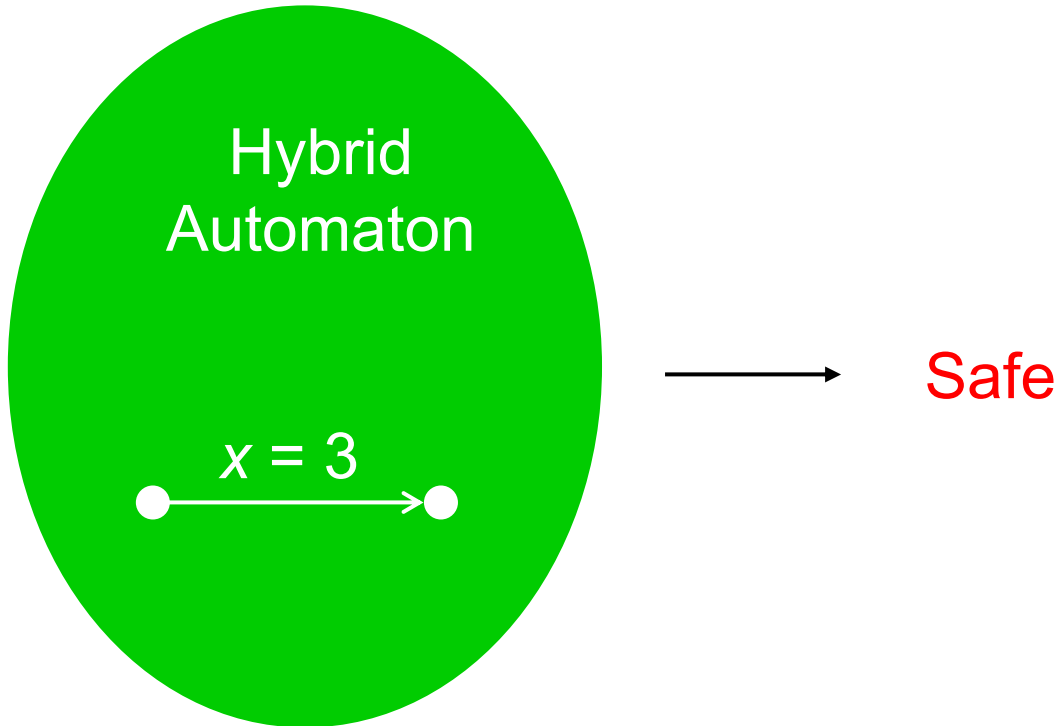
The Problem



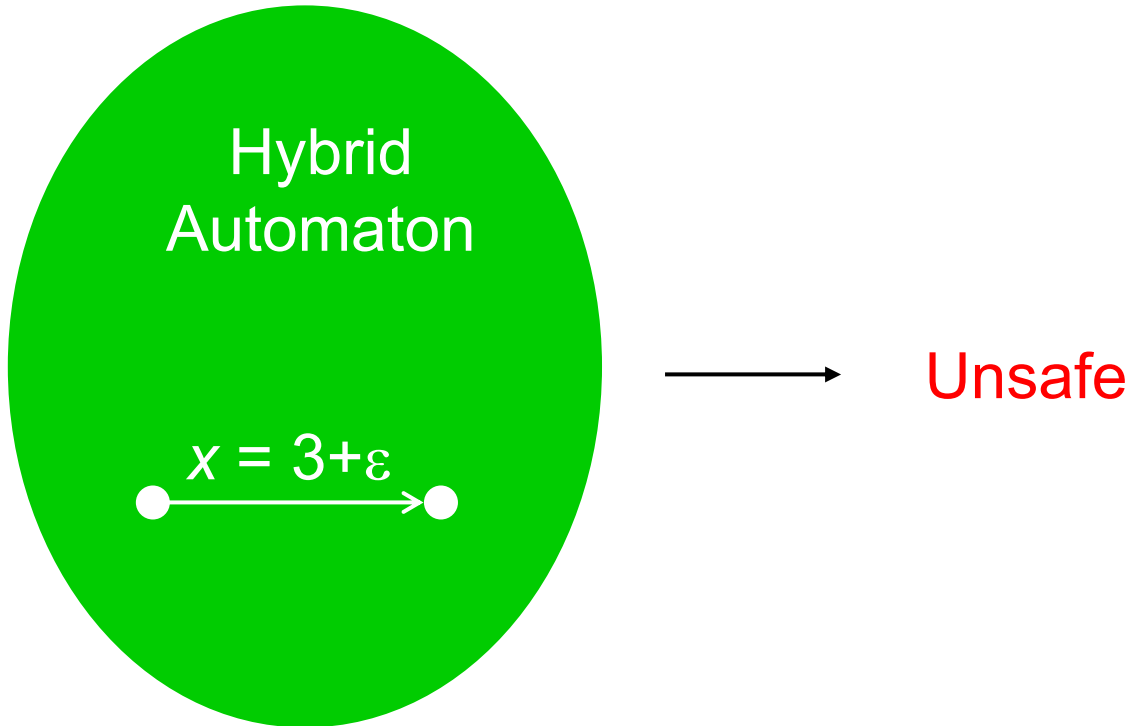
The Problem



The Problem



The Problem



The Proposed Solution

$\text{value}(\text{Model}, \text{Property}): \text{States} \rightarrow \{0, 1\}$



$\text{value}(\text{Model}, \text{Property}): \text{States} \rightarrow [0, 1]$

Discounting the Future

$\text{value}(\text{Model}, \text{Property}): \text{States} \rightarrow \{0, 1\}$

$$\text{value}(m, \diamond T) = \mu X. (T \vee \text{pre}(X))$$



$\text{discountedValue}(\text{Model}, \text{Property}): \text{States} \rightarrow [0, 1]$

$$\text{discountedValue}(m, \diamond T) = \mu X. \max(T, \lambda \cdot \text{pre}(X))$$

↑
discount factor $0 < \lambda < 1$

Robustness Theorem:

If $\text{discountedBisimilarity}(m_1, m_2) > 1 - \varepsilon$,
then $|\text{discountedValue}(m_1, \rho) - \text{discountedValue}(m_2, \rho)| < f(\varepsilon)$.

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then $|\text{discountedValue}(m_1, \rho) - \text{discountedValue}(m_2, \rho)| < f(\varepsilon)$.

Further Advantages of Discounting:

- approximability** because of geometric convergence
(avoids non-termination of fixpoint iteration)
- applies also to **probabilistic** systems and to **games**
(enables control)

Our Research Explores Three Paradigms

In **modeling**, use discounted quantitative measures.

ICALP '03 [de Alfaro, H, Majumdar]

In **composition**, treat inputs and outputs contra-variantly.

www.eecs.berkeley.edu/~tah/chic

In **implementation**, preserve logical execution times.

www.eecs.berkeley.edu/~tah/giotto