

# Model checking

Software Analysis

Topic 6

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Carlo A. Furia

USI – Università della Svizzera Italiana

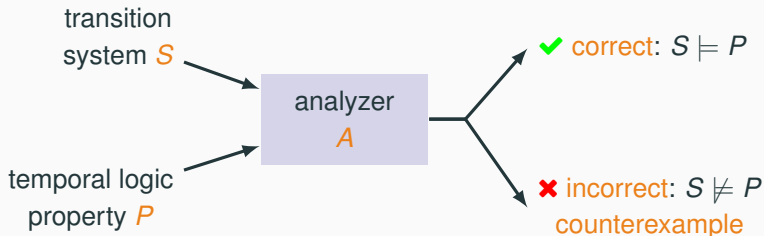
# Today's menu

Automata-based model checking

Software model checking

Real-time model checking

# Model checking: the very idea



**Model checking** is the algorithmic verification of finite-state systems:

- analyzes (finite-state) models formalized as transition systems
- verifies ordering and reachability properties expressed in temporal logic
- is completely automatic (“push button”)
- is sound and complete, as it targets fully decidable models
- when model checking fails, it returns a counterexample

## Model checking: this lecture

Model checking is a popular verification technique that has been applied in different ways. Due to its popularity, the name “model checking” has been sometimes used to describe techniques that deviate significantly from the original technique.

# Model checking: this lecture

Model checking is a popular verification technique that has been applied in different ways. Due to its popularity, the name “model checking” has been sometimes used to describe techniques that deviate significantly from the original technique.

In this lecture we have a look at three variants of **model checking**:

**model checking** in the **automata-based** framework – the most fundamental and conceptually elegant presentation of the ideas of model checking

**software** model checking combines model checking and **predicate abstraction** techniques to analyze real code using finite-state abstractions

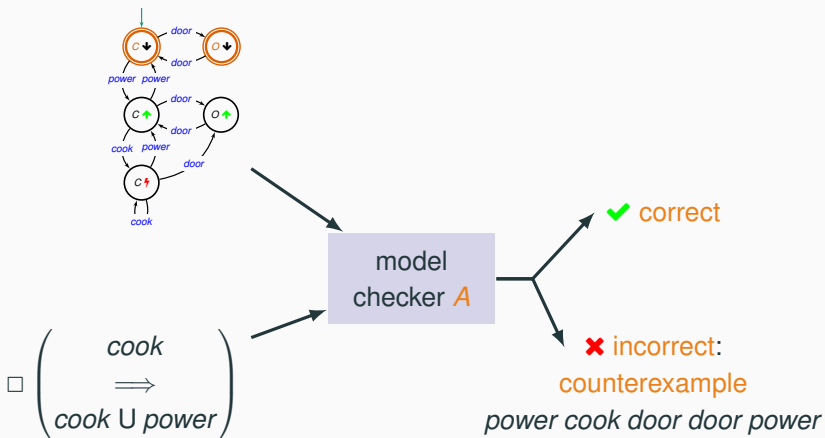
**real-time** model checking analyzes **quantitative time** models

# **Automata-based model checking**

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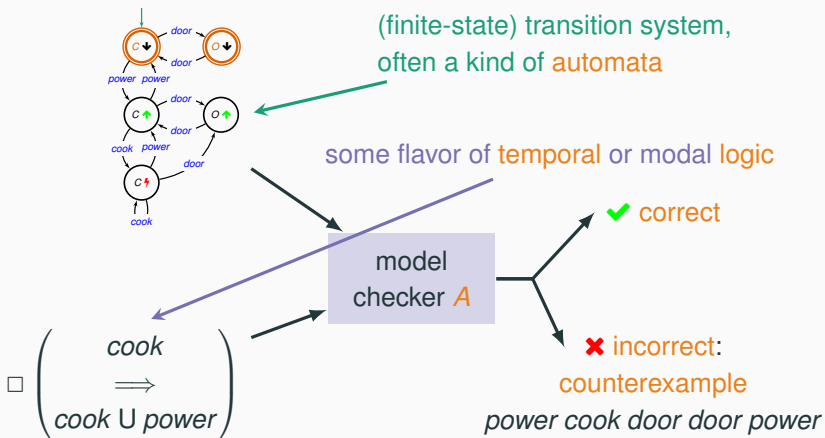
# Verification of finite-state systems

Model checking denotes a family of techniques for the  
algorithmic verification of finite-state systems  
with temporal-logic specifications



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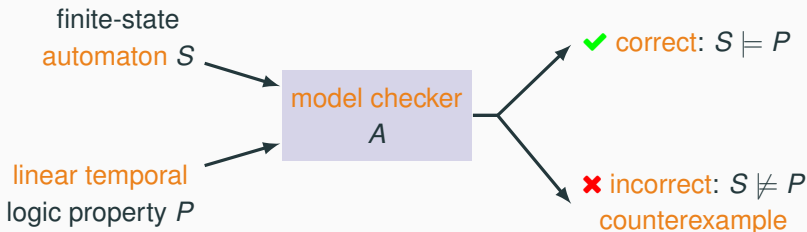


# Linear-time model checking

Linear-time model checking problem: given

- $S$ : a finite-state automaton (FSA)
- $P$ : a linear temporal logic (LTL) property

determine if every run of  $S$  satisfies  $P$ , or  
provide a counterexample: a run of  $S$  that violates  $P$

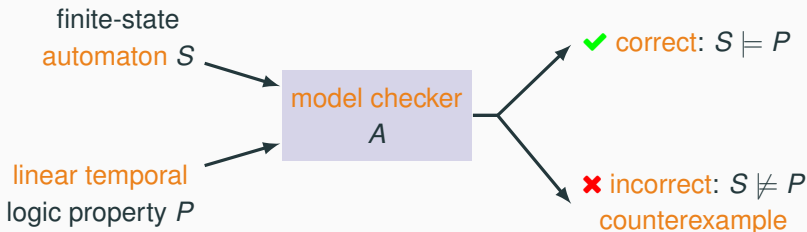


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We first describe **syntax** and **semantics** of FSAs and LTL, and then describe an **algorithm** for linear-time model checking.

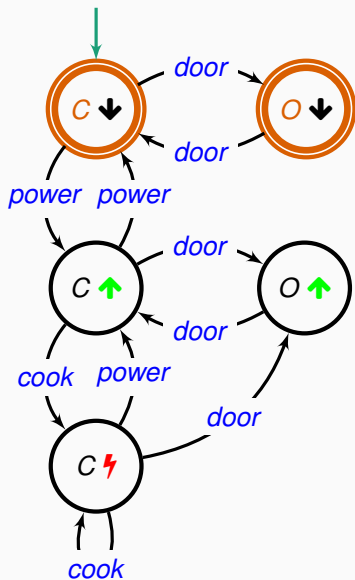
# Automata-based model checking

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## Finite-state automata

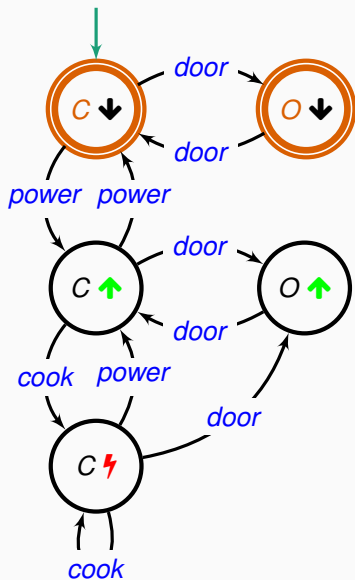
# Finite-state automata: example

We model the behavior of a **microwave oven** using an FSA.



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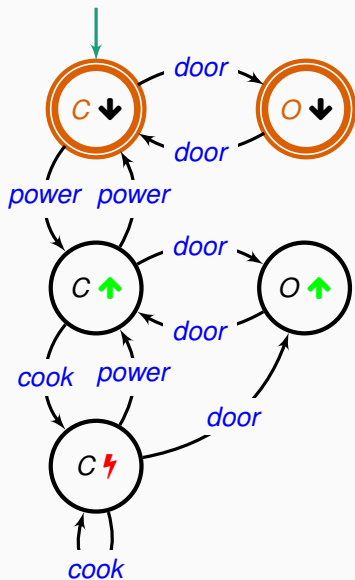
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The oven's door can be open *O* or closed *C*.

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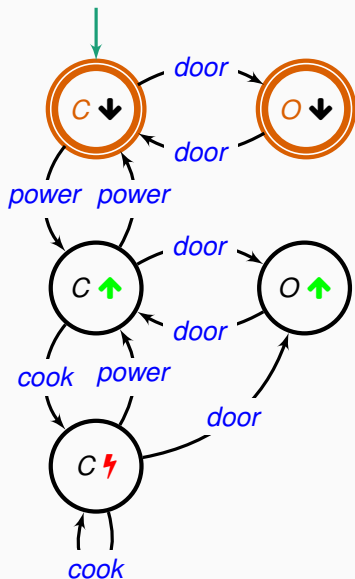


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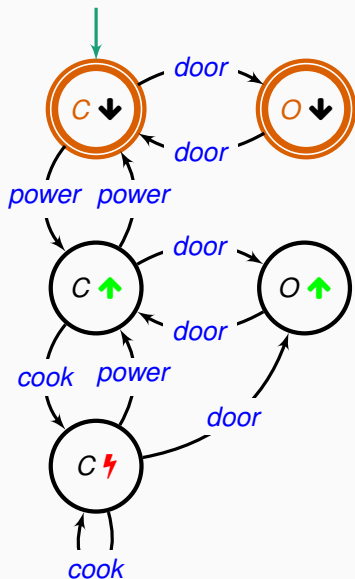
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When the oven is on and closed, pressing the cook button *cook* starts cooking **⚡**.

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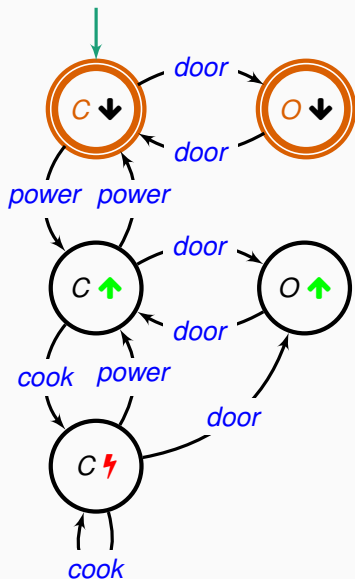
When the oven is on and closed, pressing the cook button *cook* starts cooking ⚡.

The oven is initially closed and off, and it must be eventually powered off.



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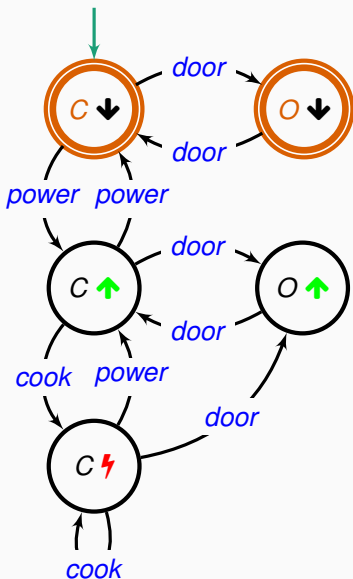
When the oven is on and closed, pressing the cook button *cook* starts cooking **⚡**.

The oven is initially closed and off, and it must be eventually powered off.

The *cook* button may only be pressed when the oven is closed and on or cooking.

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We model the behavior of a **microwave oven** using an FSA.



The oven's door can be open **O** or closed **C**.

The oven may be powered on **↑** or off **↓**.

When the oven is on and closed, pressing the cook button *cook* starts cooking **⚡**.

The oven is initially closed and off, and it must be eventually powered off.

The *cook* button may only be pressed when the oven is closed and on or cooking.

Pushing the *power* or *door* button while the oven is cooking immediately stops cooking.

# Finite-state automata: syntax

A **nondeterministic finite-state automaton (FSA)**  $A$  is a tuple  $\langle \Sigma, S, I, F, \rho \rangle$ :

- $\Sigma$ : finite nonempty input **alphabet**
- $S$ : finite nonempty set of **states**
- $I \subseteq S$ : set of **initial** states
- $F \subseteq S$ : set of **final (accepting)** states
- $\rho: S \times \Sigma \rightarrow \wp(S)$ : **transition** function

An FSA  $A$  is **deterministic** if, for all  $s, \sigma$ ,  $|\rho(s, \sigma)| \leq 1$ ; that is, there is at most **one outgoing** transition from any state  $s$  for each input symbol  $\sigma$ .

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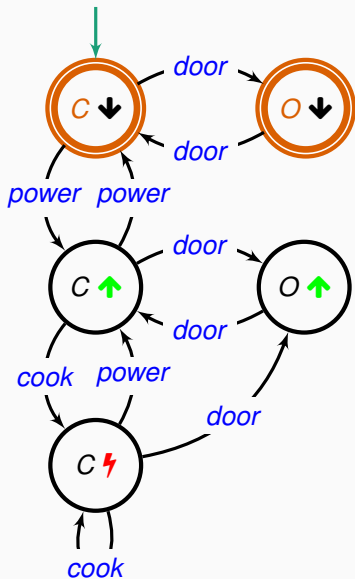
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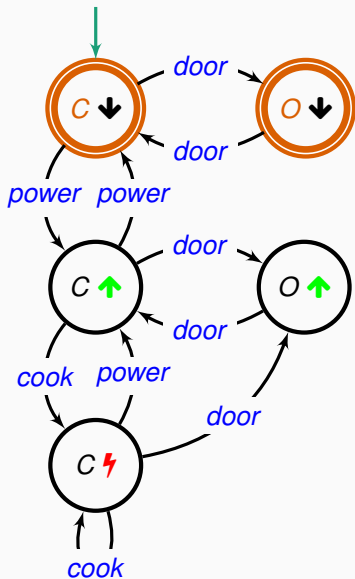
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We commonly represent FSAs with a **graph** whose nodes are states, transitions are edges, and input events decorate transitions, and initial and final states are marked.

# FSA syntax: example

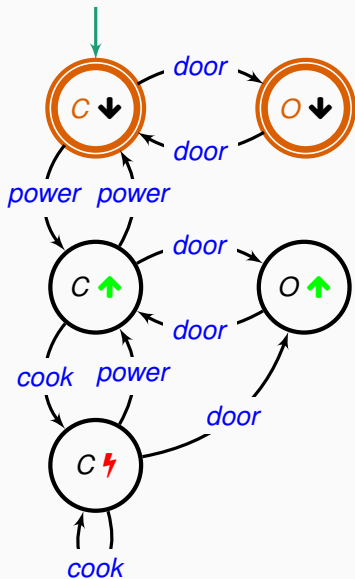


# FSA syntax: example



alphabet  $\Sigma = \{power, door, cook\}$

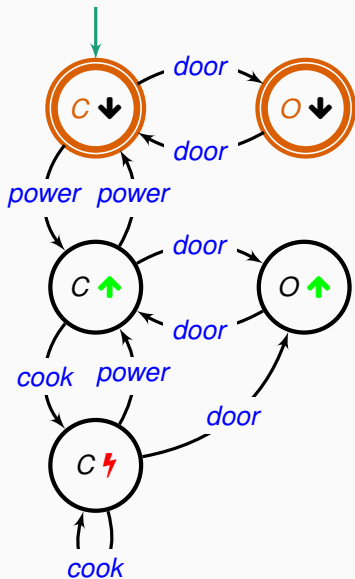
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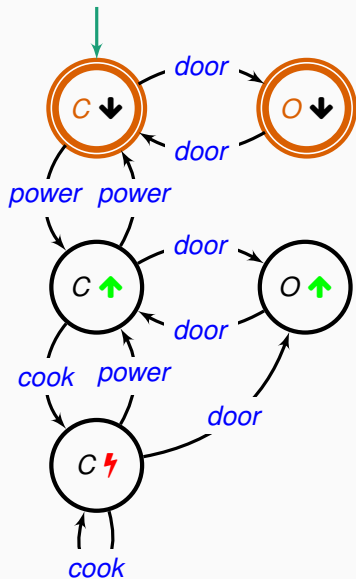
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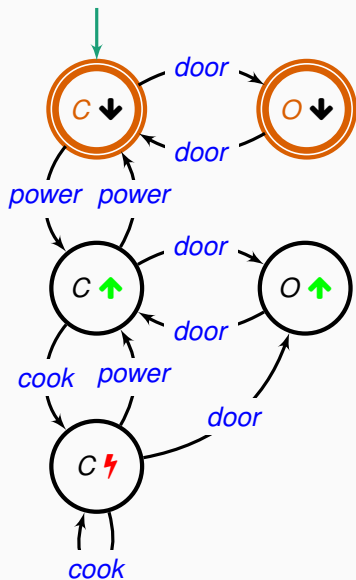
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final states  $F = \{C \downarrow, O \downarrow\}$

# FSA syntax: example



alphabet  $\Sigma = \{power, door, cook\}$

states  $S = \{C \blacktriangledown, C \uparrow, O \blacktriangledown, O \uparrow, C \text{⚡}\}$

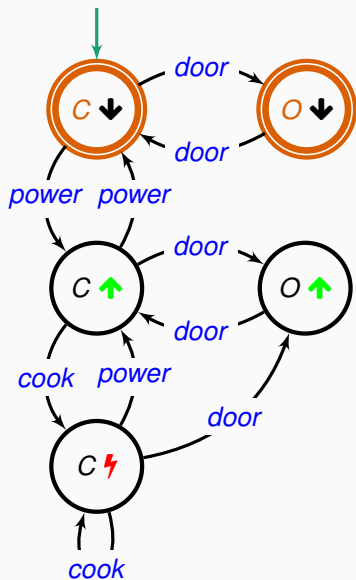
initial states  $I = \{C \blacktriangledown\}$

final states  $F = \{C \blacktriangledown, O \blacktriangledown\}$

transitions  $\rho$ :

- $\rho(C \blacktriangledown, power) = \{C \uparrow\}$
- $\rho(O \blacktriangledown, door) = \{C \blacktriangledown\}$
- $\rho(O \blacktriangledown, cook) = \{\}$
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- $\rho(O \downarrow, door) = \{C \downarrow\}$
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- ...

The automaton is **deterministic**

# Finite-state automata: runs

Let  $A = \langle \Sigma, S, I, \rho, F \rangle$  be an FSA.

An **input word** is an input sequence of any (finite) length:

$$w = w[1] w[2] \dots w[n] \in \Sigma^*$$

The **empty** word  $\epsilon$  has **zero length** ( $n = |\epsilon| = 0$ ).

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A **run** of  $A$  over  $w$  is a sequence of states

$$r = r[0] r[1] \dots r[n] \in S^* \quad \text{that:}$$

- **starts** from an **initial** state:  $r[0] \in I$
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A run  $r$  is **accepting** if it **ends** in a **final** state:  $r[n] \in F$ .

In this case we say that  $A$  **accepts**  $w$ .

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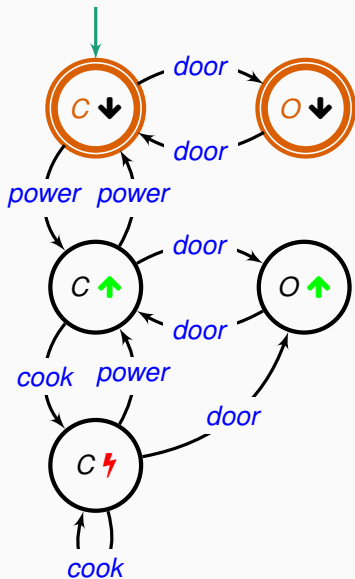
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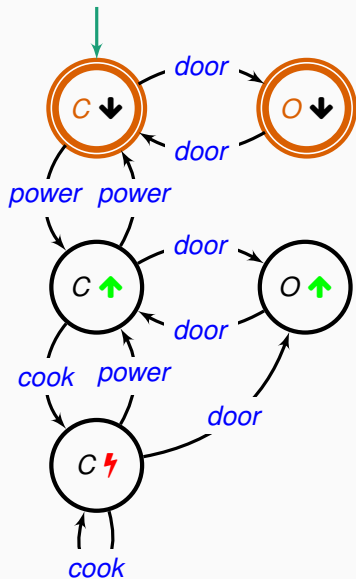
In practice, an accepting run is any **path on  $A$ 's directed graph** that starts in an initial state and ends in a final state.

# FSA runs: example



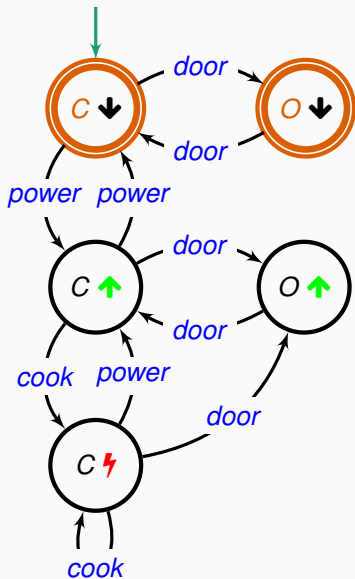


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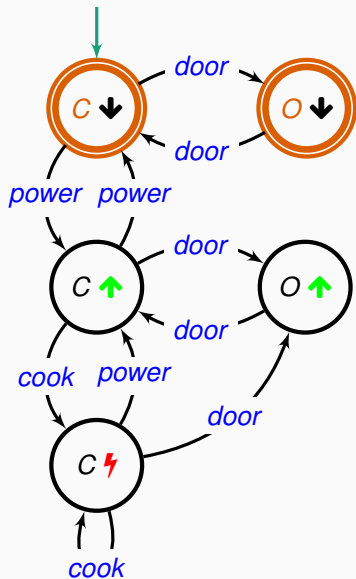
Run  $r = C \blacktriangledown C \blacktriangle C \textcolor{red}{\blacktriangleright} O \blacktriangle$

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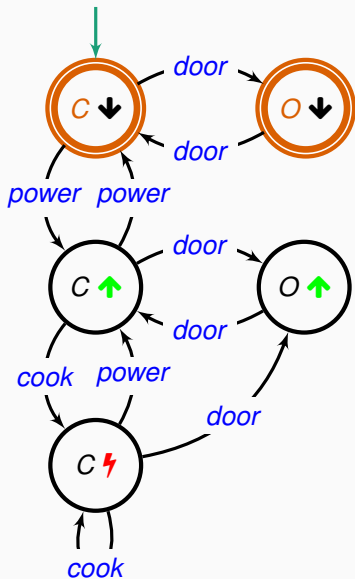
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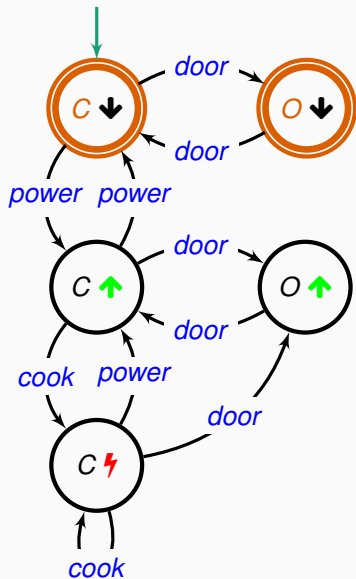
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Run  $r = C \downarrow C \uparrow C \text{⚡} O \uparrow$   
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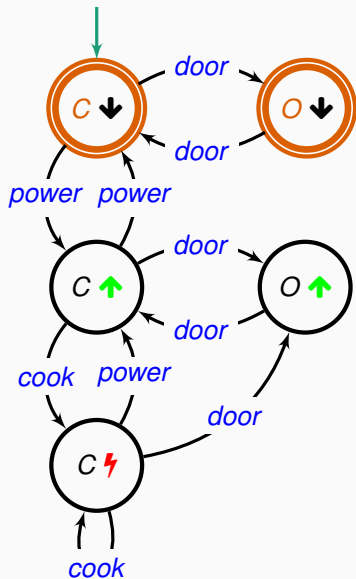
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is **accepting**.

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The **language** of a finite-state automaton  $A = \langle \Sigma, S, I, F, \rho \rangle$  is the **set of input words** that  $A$  **accepts**:

$$\mathcal{L}(A) = \{w \in \Sigma^* \mid \text{there is an accepting run of } A \text{ over } w\}$$

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The emptiness problem is the fundamental **decision problem** of automata (and operational formalisms in general):

The **emptiness problem**: given an automaton  $A$ , determine if it accepts **any** words – that is if  $A$ 's language is **empty**.



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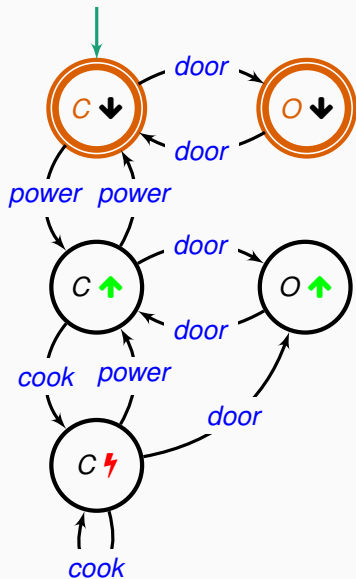
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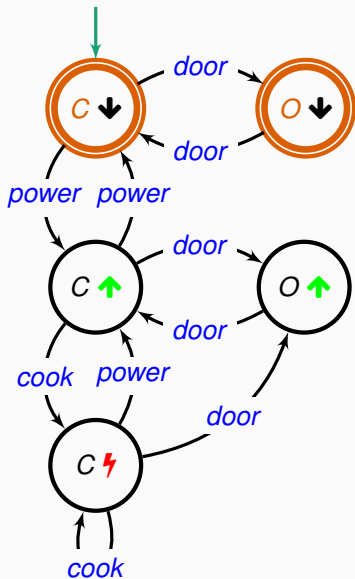
The **emptiness problem**: given an automaton  $A$ , determine if it accepts **any** words – that is if  $A$ 's language is **empty**.

Note that even though FSAs have a **finite** number of states, their languages may consist of an **infinite** number of words.

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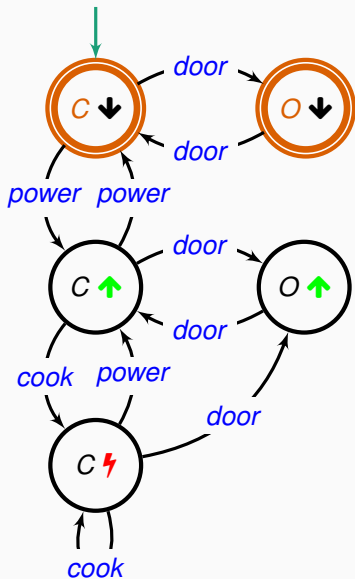


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Words in the automaton's language include:

- *power power*
- *power cook power door door power*
- $\epsilon$
- ...

# **Automata-based model checking**

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**Linear temporal logic**

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The power button is **never** pressed:

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The oven cooks **until** it is turned off:

$$cook \text{ U } power$$

The power button is **never** pressed:

$$\Box(\neg power)$$

# Linear temporal logic: syntax

Formulas of propositional **linear temporal logic (LTL)** are defined as:

$$\begin{aligned} F ::= & p \mid \neg F \mid F_1 \wedge F_2 \mid F_1 \vee F_2 \mid F_1 \implies F_2 && \text{(propositional connectives)} \\ & \mid \mathsf{X} F \mid \Box F \mid \Diamond F \mid F_1 \cup F_2 && \text{(temporal connectives)} \end{aligned}$$

where  $p \in \Pi$  is any **proposition** from a set  $\Pi$ .

# Linear temporal logic: syntax

next/in the  
next step

box/always/  
globally

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$\mid \mathbf{X} F \mid \mathbf{\Box} F \mid \mathbf{\Diamond} F \mid F_1 \mathbf{U} F_2$  (temporal connectives)

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**Temporal connectives**, also called temporal **operators** or **modal operators**, describe **when** their arguments are true.

## LTL syntax: examples

The examples we have seen before are LTL formulas over propositions in  $\Pi = \{door, power, cook\}$ .

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The door **will be** closed eventually after it is opened:

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The oven cooks **until** it is turned off:

$$cook \text{ U } \neg power$$

The power button is **never** pressed:

$$\Box(\neg power)$$

# Linear temporal logic: satisfaction relation

A word  $w = w[1] w[2] \dots w[n] \in \Pi^*$  satisfies LTL formula  $F$  at position (step)  $1 \leq k \leq n$ , written  $w, k \models F$ , iff:

$w, k \models p$	iff	$p = w[k]$
$w, k \models \neg F$	iff	$w, k \not\models F$
$w, k \models F_1 \wedge F_2$	iff	$w, k \models F_1$ and $w, k \models F_2$
$w, k \models F_1 \vee F_2$	iff	$w, k \models F_1$ or $w, k \models F_2$
$w, k \models F_1 \implies F_2$	iff	$w, k \models \neg F_1 \vee F_2$
$w, k \models \mathsf{X} F$	iff	$k < n$ and $w, k + 1 \models F$
$w, k \models \Box F$	iff	for all $k \leq h \leq n$ : $w, h \models F$
$w, k \models \Diamond F$	iff	for some $k \leq h \leq n$ : $w, h \models F$
$w, k \models F_1 \cup F_2$	iff	for some $k \leq h \leq n$ : $w, h \models F_2$ and, for all $k \leq j < h$ : $w, j \models F_1$

# LTL satisfaction: example

$w$     =     $door$      $door$      $door$      $power$   
               $w[1]$      $w[2]$      $w[3]$      $w[4]$

$w, 1 \models door$

$w, 2 \models door$

$w, 3 \models door$

$w, 4 \models power$

$w, 1 \models \neg power$

$w, 2 \models \neg cook$

$w, 4 \models \begin{matrix} door \\ \vee power \end{matrix}$

$w, 4 \models \begin{matrix} power \\ \wedge \neg door \end{matrix}$

$w, 1 \models X door$

$w, 2 \models X(\neg cook)$

$w, 3 \models X power$

$w, 4 \models \neg(X \top)$

$w, 1 \models \Box(\neg cook)$

$w, 2 \models \Box \left( \begin{matrix} door \\ \vee \top \end{matrix} \right)$

$w, 3 \models \neg \Box door$

$w, 4 \models \Box power$

$w, 1 \models \Diamond door$

$w, 2 \models \Diamond(\neg door)$

$w, 3 \models \Diamond X power$

$w, 4 \models \Diamond power$

$w, 1 \models door \cup power$

$w, 2 \models \top \cup power$

$w, 3 \models \perp \cup \top$

$w, 4 \models power \cup power$



# Linear temporal logic: semantics

A word  $w$  satisfies an LTL formula  $F$  if it satisfies it initially:

$$w \models F \quad \text{iff} \quad w, 1 \models F$$

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The language of a linear temporal logic formula  $F$  is the set of words that satisfy  $F$ :

$$\mathcal{L}(F) = \{w \in \Pi^* \mid w \models F\}$$

# LTL semantics: examples

Examples of words that **satisfy**:

$F$	$w_1 \models F$	$w_2 \models F$	$w_3 \models F$	$w_4 \not\models F$
$\Box p$				
$X q$				
$X(\Box p)$				
$\Box(X p)$				
$\Diamond p$				
$\Diamond(X p)$				
$p \cup q$				

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$\Box p$	$p$	$ppp$	$\epsilon$	$pqp$
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$X(\Box p)$				
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$Xq$	$pq$	$qqq$	$pqppq$	$qpqqq$
$X(\Box p)$	$qp$	$ppp$	$p$	$pqpp$
$\Box(Xp)$	$\epsilon$			$ppp$
$\Diamond p$	$p$	$qqqqpq$	$qpp$	$\epsilon$
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$\Box(Xp)$	$\epsilon$			$ppp$
$\Diamond p$	$p$	$qqqqpq$	$qpp$	$\epsilon$
$\Diamond(Xp)$	$pp$	$qqqpq$	$qp$	$pqq$
$p \cup q$	$ppq$	$q$	$ppppq$	$ppp$

# **Automata-based model checking**

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## **Model-checking algorithm**

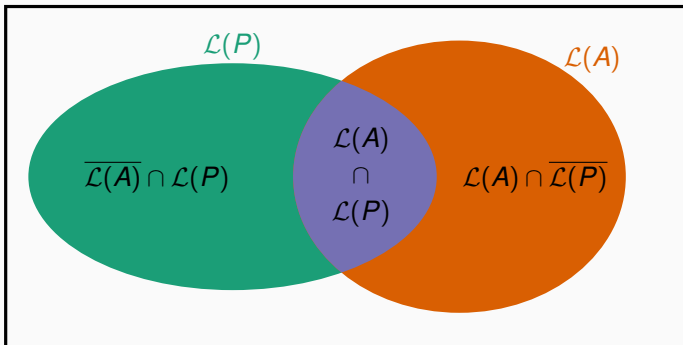
# Linear-time model checking, semantically

Linear-time model checking problem: given

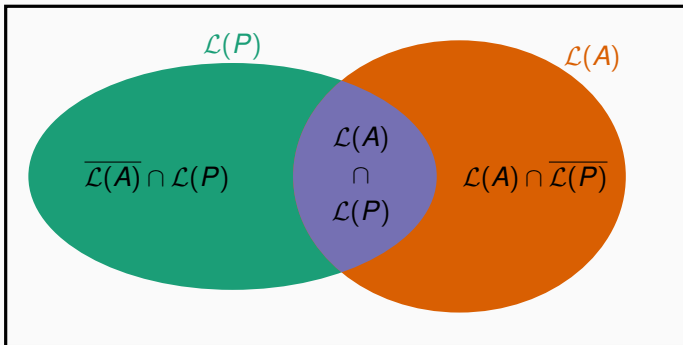
- $A$ : a finite-state automaton with alphabet  $\Sigma$
- $P$ : a linear temporal logic property over propositions in  $\Sigma$

determine if  $A \models P$ : every word accepted by  $A$  satisfies  $P$

$\Sigma^*$

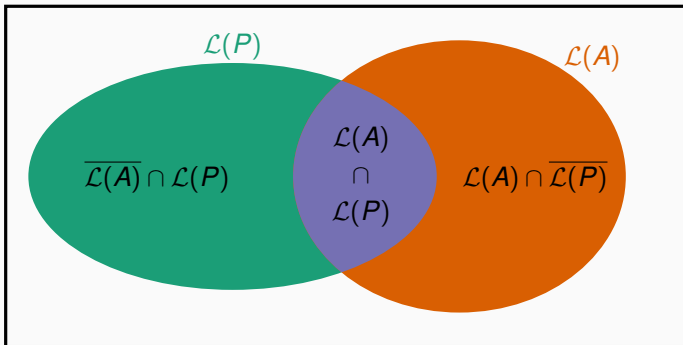


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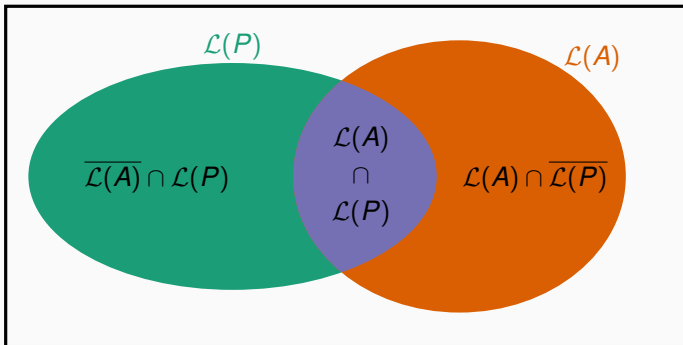
 $\Sigma^*$  $A \models P$ 

iff

 $w \models A \text{ implies } w \models P$

# Linear-time model checking, semantically

$\Sigma^*$



$A \models P$

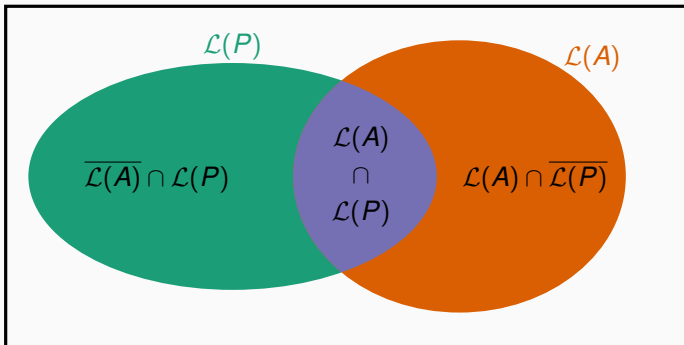
iff

$w \models A$  implies  $w \models P$

iff

$w \in \mathcal{L}(A)$  implies  $w \in \mathcal{L}(P)$

# Linear-time model checking, semantically

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iff

 $w \in \mathcal{L}(A) \text{ implies } w \in \mathcal{L}(P)$ 

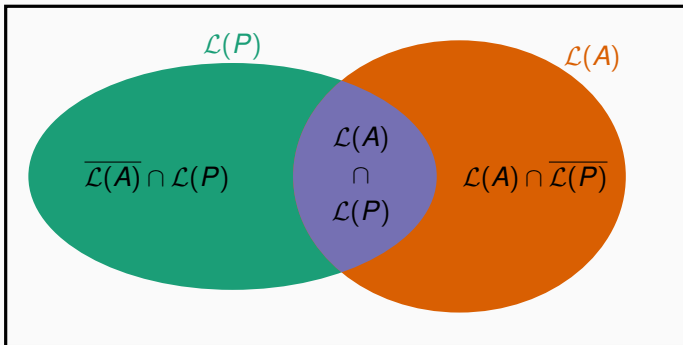
iff

 $\mathcal{L}(A) \subseteq \mathcal{L}(P)$



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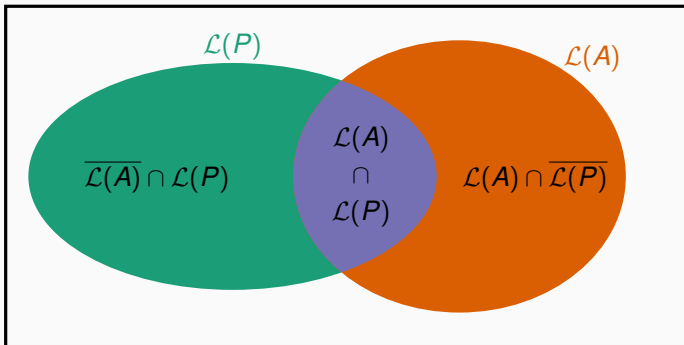
iff

$\mathcal{L}(A) \subseteq \mathcal{L}(P)$

iff

$\mathcal{L}(A) \cap \overline{\mathcal{L}(P)} = \emptyset$

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 $\mathcal{L}(A) \subseteq \mathcal{L}(P)$ 

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 $\mathcal{L}(A) \cap \overline{\mathcal{L}(P)} = \emptyset$ 

iff

 $\mathcal{L}(A) \cap \mathcal{L}(\neg P) = \emptyset$

# Model checking as emptiness checking

Linear-time model checking: given

- $A$ : a finite-state automaton with alphabet  $\Sigma$
- $P$ : a linear temporal logic property over propositions in  $\Sigma$

$\mathcal{L}(A) \cap \mathcal{L}(\neg P)$  is empty

every word accepted  
by  $A$  satisfies  $P$

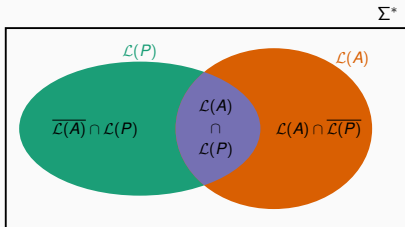
✓  $A \models P$

$\mathcal{L}(A) \cap \mathcal{L}(\neg P)$  is not empty

some word accepted by  $A$   
does not satisfy  $P$

✗  $A \not\models P$

every word in  $\mathcal{L}(A) \cap \mathcal{L}(\neg P)$   
is a counterexample



# Model-checking algorithm

Expressing the model-checking problem as **emptiness** checking suggests a **technique** for model checking which combines **three algorithms**:

**MONITOR:** given a temporal logic formula  $P$  build an automaton  $\mathcal{A}(P)$  such that  $\mathcal{L}(\mathcal{A}(P)) = \mathcal{L}(P)$

**INTERSECTION:** given automata  $A$  and  $B$ , build an automaton  $A \times B$  such that  $\mathcal{L}(A \times B) = \mathcal{L}(A) \cap \mathcal{L}(B)$

**EMPTINESS:** given an automaton  $A$  determine whether  $\mathcal{L}(A) = \emptyset$

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**EMPTINESS**: given an automaton  $A$  determine whether  $\mathcal{L}(A) = \emptyset$

Model-checking **algorithm**: given a finite-state automaton  $A$  and a linear temporal logic property  $P$ :

1. **monitor**: build  $\mathcal{A}(\neg P)$
2. **intersection**: build  $A \times \mathcal{A}(\neg P)$
3. **emptiness**: test whether  $\mathcal{L}(A \times \mathcal{A}(\neg P)) = \emptyset$   
$$\mathcal{L}(A \times \mathcal{A}(\neg P)) = \emptyset \quad \text{iff} \quad A \models P$$

# Monitors: from temporal logic to automata

Given an LTL formula  $P$  it is **always** possible to build a **monitor** of  $P$ : an FSA  $\mathcal{A}(P)$  that accepts precisely the words that satisfy  $P$ .

There are **various** algorithms to build monitors. We won't describe any of them in detail, but simply show their **general ideas** on some examples.

# Monitors: always

$P_1 = \Box p$      $p$  holds always

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# Monitors: always

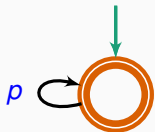
$$P_1 = \Box p \quad p \text{ holds always}$$



- the empty word satisfies  $P_1$

# Monitors: always

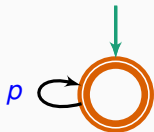
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- the empty word satisfies  $P_1$
- as long as  $p$  occurs, we accept

# Monitors: always

$$P_1 = \Box p \quad p \text{ holds always}$$



- the empty word satisfies  $P_1$
- as long as  $p$  occurs, we accept
- no other transition is possible

## Monitors: implication and next

$P_2 = \Box(p \implies X q)$     whenever  $p$  holds,  $q$  holds next

## Monitors: implication and next

$P_2 = \Box(p \Rightarrow X q)$     whenever  $p$  holds,  $q$  holds next



# Monitors: implication and next

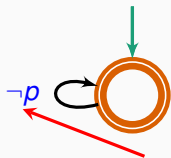
$P_2 = \Box(p \implies X q)$  whenever  $p$  holds,  $q$  holds next



- the empty word satisfies  $P_2$

# Monitors: implication and next

$P_2 = \Box(p \implies X q)$  whenever  $p$  holds,  $q$  holds next

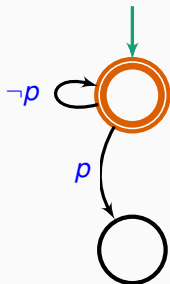


a transition for every proposition  
 $q \in \Pi$  such that  $q \neq p$

- the empty word satisfies  $P_2$
- as long as any proposition **other than**  $p$  occurs, the implication is trivially true

# Monitors: implication and next

$P_2 = \Box(p \implies X q)$  whenever  $p$  holds,  $q$  holds next

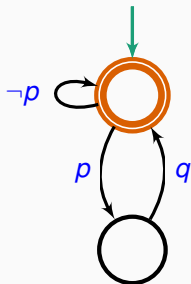


- the empty word satisfies  $P_2$
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- when  $p$  occurs, move to a different state (“**next**”) which is not accepting



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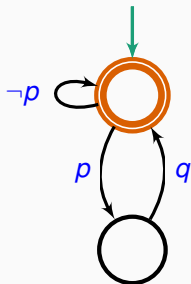
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## Monitors: until

$P_3 = \Box(p \implies \textcolor{brown}{X}(\textcolor{brown}{oUq}))$       whenever  $p$  holds,  $o$  holds until  $q$  holds later

# Monitors: until

$P_3 = \Box(p \implies X(oUq))$       whenever  $p$  holds,  $o$  holds until  $q$  holds later



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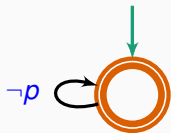
- the empty word satisfies  $P_3$



# Monitors: until

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whenever  $p$  holds,  $o$  holds until  $q$  holds later

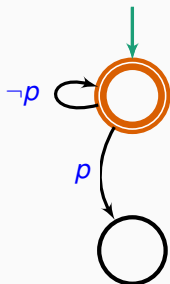


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# Monitors: until

$$P_3 = \Box(p \implies \textcolor{brown}{X}(o \cup q))$$

whenever  $p$  holds,  $o$  holds until  $q$  holds later

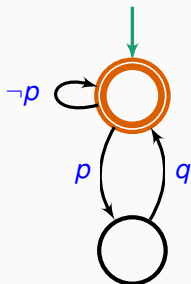


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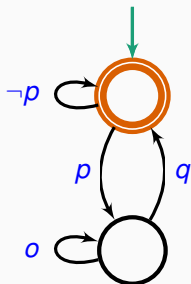
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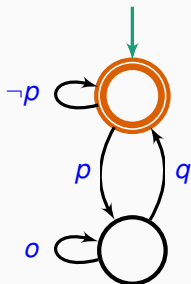


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## Monitors: eventually and until

$P_4 = \Diamond(p \wedge X(qUo))$       eventually  $p$  holds, and then  $q$  holds until  $o$  holds

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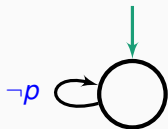


- the empty word does **not** satisfy  $P_4$

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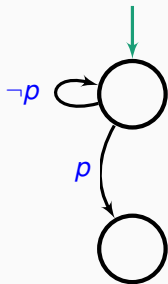


- the empty word does **not** satisfy  $P_4$
- as long as any proposition **other than**  $p$  occurs,  $P_4$  remains false

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$$P_4 = \diamond(p \wedge X(qUo))$$

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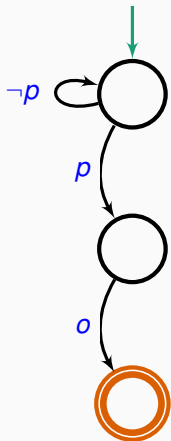


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eventually  $p$  holds, and then  $q$  holds until  $o$  holds



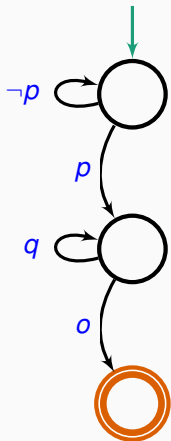
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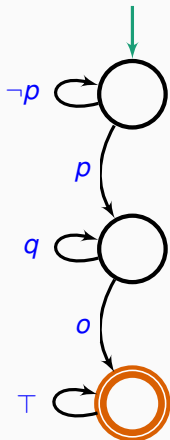


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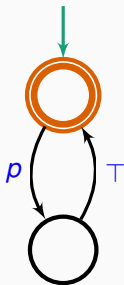
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- from in the new state,  $o$  must occur eventually, which finally leads to an **accepting** state
- until  $o$  occurs, only  $q$  **may** occur
- after we reach the accepting state, we never leave it –  $P_4$  remains **true forever**

# From automata to LTL?

While it is always possible to go from LTL to FSA, automata are strictly **more expressive**.

*ODD*: “ $p$  occurs in every **odd** time step 1, 3, 5, ...”

Automaton accepting  
language *ODD*:



There is no LTL formula that is satisfied  
**precisely** by words in *ODD*.

In particular, formula

$$p \wedge \Box(p \implies \Box \Box p)$$

is **too strong** because, once  $p$  occurs in  
an **even** time step, it will always occur  
afterwards.

# Automata of negated properties

The model-checking algorithm computes the monitor of a **negated** property  $\neg P$ . There are two ways of doing this, either of which may be easier depending on the specific  $P$ :

1. build the monitor  $\mathcal{A}(\neg P)$  of  $\neg P$  **directly**
2. build the monitor  $\mathcal{A}(P)$  of  $P$ , and then **complement** it, getting  $\overline{\mathcal{A}(P)} = \mathcal{A}(\neg P)$

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To build the monitor of the negated property **directly**, the following **equivalences** may be useful:

$$\neg \Diamond p \equiv \Box \neg p$$

$$\neg \Box p \equiv \Diamond \neg p$$

$$\neg X p \equiv X(\neg p) \vee \neg X \top$$

$$\Diamond p \equiv \top \cup p$$



holds in the last position of a word, or on the empty word

# Complementing automata

The **complement automaton**  $\bar{A}$  accepts the language  $\mathcal{L}(\bar{A})$  of all words that  $A$  rejects

To build the **complement** automaton from  $A$ :

1. build  $A_1$ : an automaton equivalent to  $A$  that is **deterministic** and with **total** transition function
2. build  $A_2$  by **switching accepting** and non-accepting states of  $A_1$
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To make a transition function  $\rho$  **total**:

1. add a special **error** state  $e \notin S$
2. add a **loop**  $\rho(e, \sigma) = \{e\}$  for every  $\sigma \in \Sigma$
3. for every  $s, \sigma$  such that  $\rho(s, \sigma) = \{\}$  (or undefined), define the transition  $\rho(s, \sigma) = \{e\}$

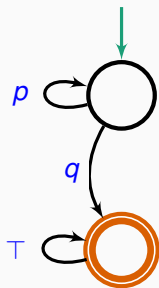
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Build the monitor of  $P_6 = \neg P_5 = \neg(p \cup q)$  by **complementing**  $\mathcal{A}(P_5)$ .  
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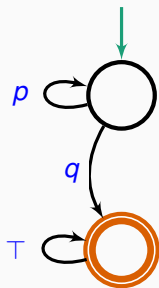
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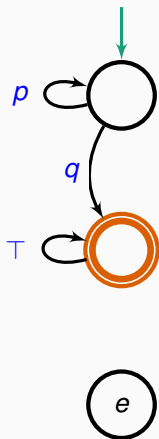
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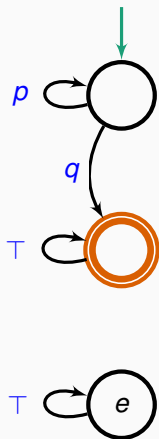
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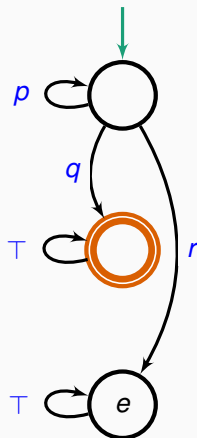
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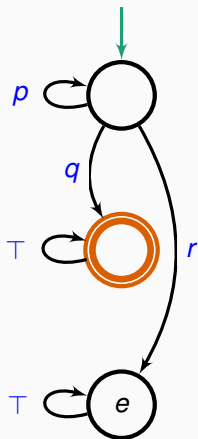
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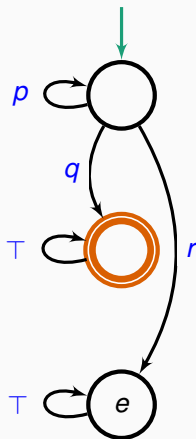
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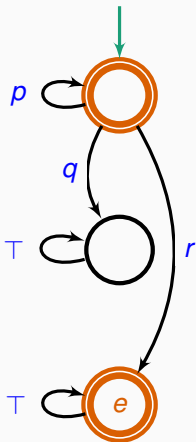
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# Monitor: running example

Let's build the monitors for two variants of properties that we would like to model check in the microwave running example.

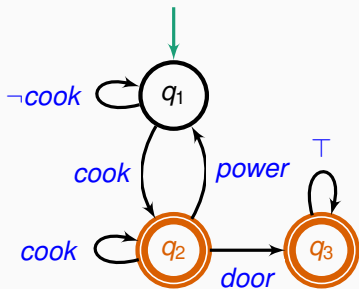
$$Q = \Box(\text{cook} \implies \text{cook} \cup \text{power})$$

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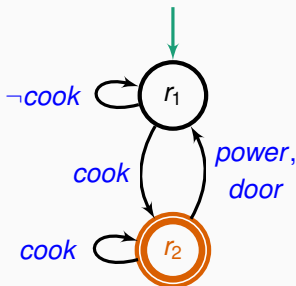
$$\neg Q = \Diamond(\text{cook} \wedge \neg(\text{cook} \cup \text{power}))$$

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Automaton  $\mathcal{A}(\neg Q)$  built  
by **complementing**  $\mathcal{A}(Q)$ :



Automaton  $\mathcal{A}(\neg R)$  built  
by **complementing**  $\mathcal{A}(R)$ :



## Intersection: running automata in parallel

An automaton  $C$  that accepts the intersection of two automata  $A$  and  $B$ 's languages runs  $A$  and  $B$  in parallel:

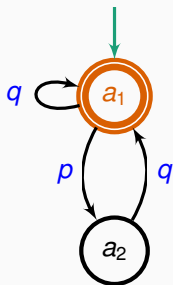
- starts from any combination of initial states of  $A$  and  $B$
- transitions only when both  $A$  and  $B$  have a transition for the current input
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# Intersection: running automata in parallel

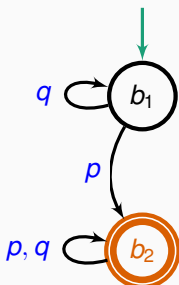
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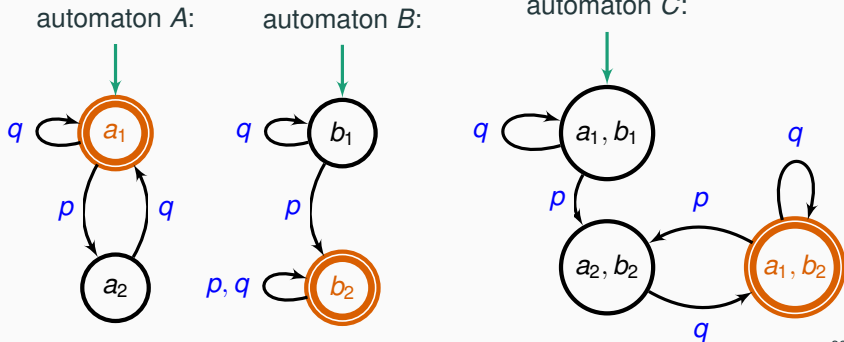
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# Product automaton construction

Given FSAs  $A = \langle \Sigma, S_A, I_A, F_A, \rho_A \rangle$  and  $B = \langle \Sigma, S_B, I_B, F_B, \rho_B \rangle$ , the **product automaton**  $A \times B = \langle \Sigma, S, I, F, \rho, \rangle$  is defined as:

$$S = S_A \times S_B$$

$$I = \{(a, b) \mid a \in I_A \text{ and } b \in I_B\}$$

$$F = \{(a, b) \mid a \in F_A \text{ and } b \in F_B\}$$

$$\rho((a, b), \sigma) = \{(a_2, b_2) \mid a_2 \in \rho_A(a, \sigma) \text{ and } b_2 \in \rho_B(b, \sigma)\}$$

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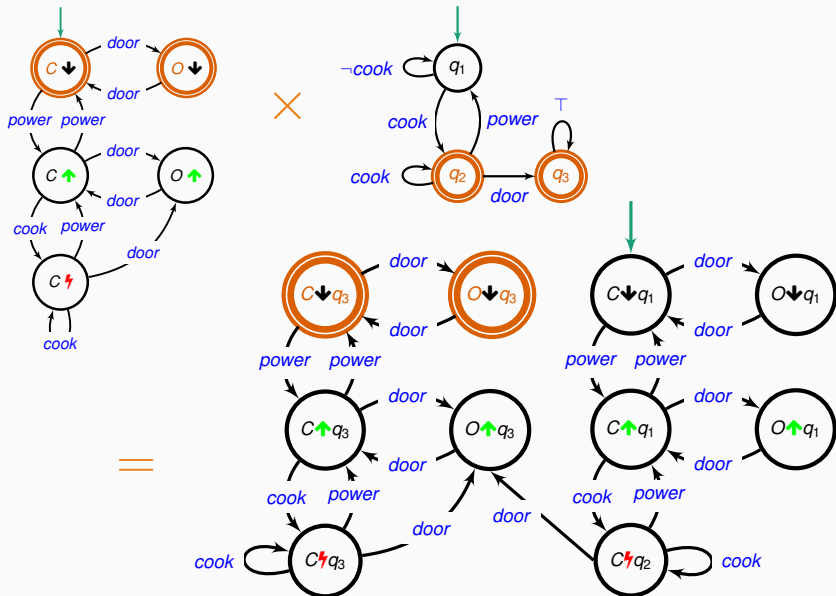
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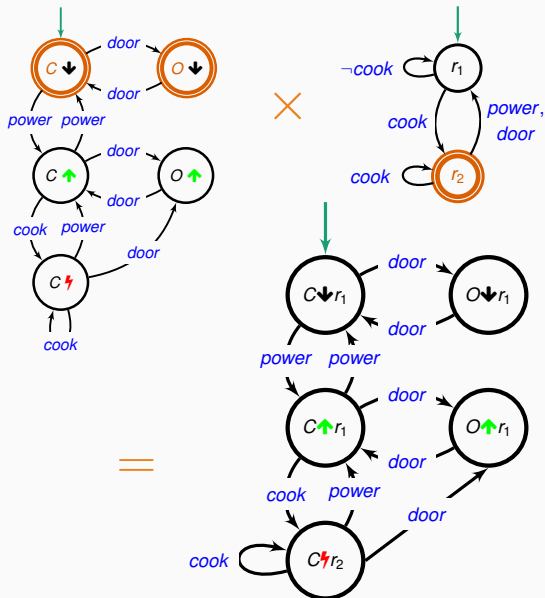
The **language** of the product automaton is the **intersection** of the intersected automata's languages:

$$\mathcal{L}(A \times B) = \mathcal{L}(A) \cap \mathcal{L}(B)$$

# Product automaton of microwave and $\mathcal{A}(\neg Q)$



# Product automaton of microwave and $\mathcal{A}(\neg R)$





## Emptiness: reachability on graph

An automaton  $A = \langle \Sigma, S, I, F, \rho \rangle$  accepts the **empty language** iff there is **no** final state  $f \in F$  that is **reachable** from any initial state  $i \in I$  on the directed **graph** representing  $A$ 's transitions

If we find a directed path from some  $i \in I$  to some  $f \in F$ , following it gives a **word** that is accepted by  $A$ .

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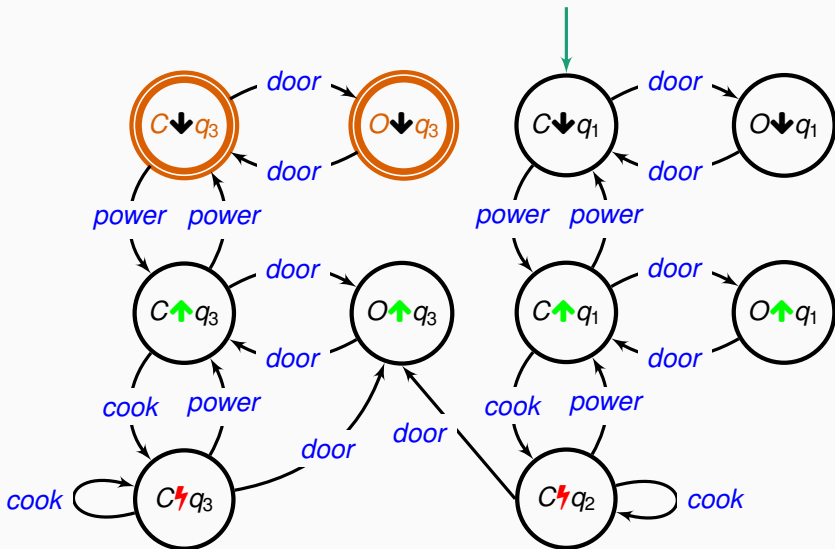
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In the overall model-checking algorithm, we check emptiness of

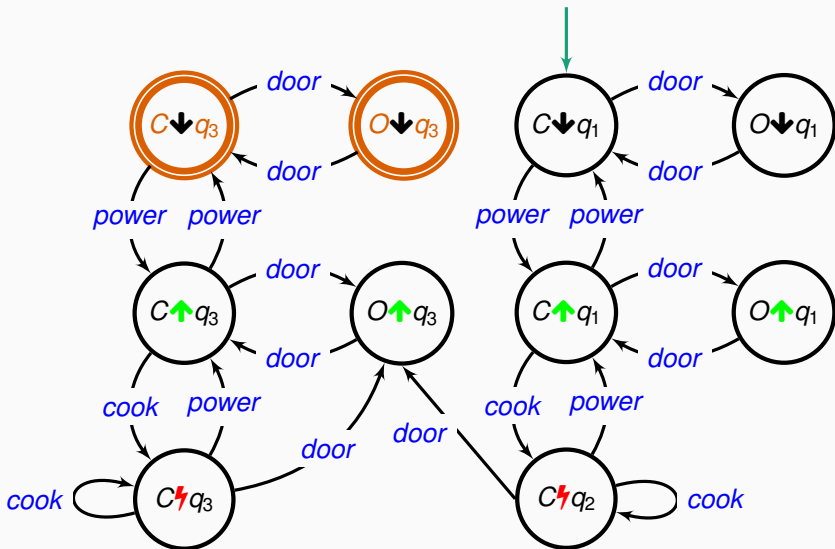
$$C = A \times \mathcal{A}(\neg P)$$

- if  $\mathcal{L}(C)$  is **empty**, we conclude  $A \models P$
- if  $\mathcal{L}(C)$  is **not** empty, any accepting path in  $C$  gives a **counterexample** word  $w$  such that  $w \models A$  and  $w \not\models P$

# Emptiness: microwave example with property $Q$



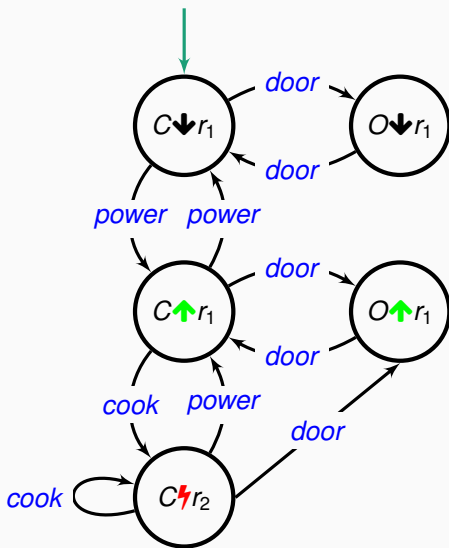
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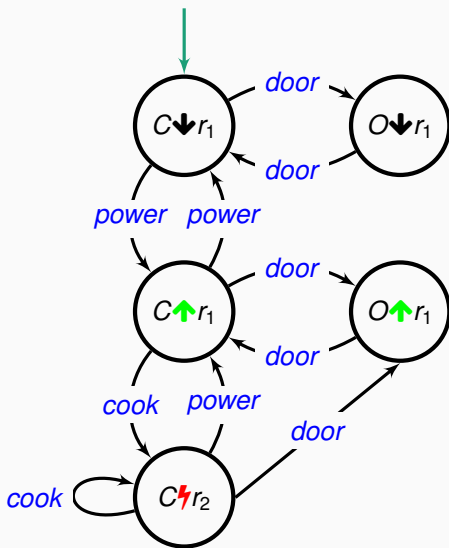
$A \not\models Q$

Counterexample: *power cook door door power*

## Emptiness: microwave example with property $R$



# Emptiness: microwave example with property $R$



$$A \models R$$

# Model checking: static or dynamic?

## Static:

- without executing the software
- based on symbolic constraints on states
- typically sound

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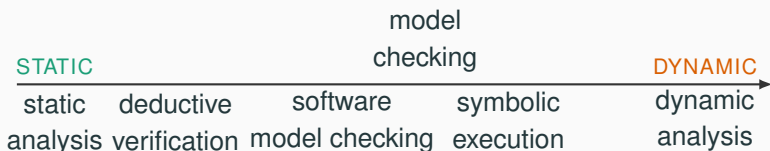
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# Soundness and completeness

Model checking is **sound** and **complete**:

**sound**: if  $\mathcal{L}(A \times \mathcal{A}(\neg P))$  is empty,  $A \models P$

**complete**: if  $\mathcal{L}(A \times \mathcal{A}(\neg P))$  is not empty,  $A \not\models P$   
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If the finite-state model  $A$  that we model check is an **abstraction** of a more complex **infinite-state** program, then the analysis of model-checking may be **not sound or complete** for the program. We will analyze this in detail when presenting software model checking.

# Complexity of the model-checking algorithm

Model-checking **algorithm**: given a finite-state automaton  $A$  and a linear temporal logic property  $P$ :

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The automata-based model checking algorithm to decide whether  $A \models F$  runs in time  **$O(|A| \cdot 2^{O(|F|)})$**



# Complexity of the model-checking problem

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(and space) with depth-first or breadth-first search,  
and is NLOGSPACE-complete

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**EMPTINESS** is **reachability**, which can be done in linear time (and space) with depth-first or breadth-first search, and is NLOGSPACE-complete

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Thus, the model-checking algorithm we have presented is worst-case optimal.

The automata-based model checking algorithm  
to decide whether  $A \models F$  runs in time  $O(|A| \cdot 2^{O(|F|)})$

# Model-checking complexity in practice

The automata-based model checking algorithm  
to decide whether  $A \models F$  runs in time  $O(|A| \cdot 2^{O(|F|)})$

In most cases, the LTL property  $F$  is **much smaller** than the system model  $A$ . Therefore, the exponential dependency is not a major problem in practice.

# Model-checking complexity in practice

On the contrary, the combinatorial growth of the **product** construction often leads to a **huge** graph to check that does not fit memory.

To ameliorate this, the model-checking algorithm can be performed **on the fly** by constructing the graph **incrementally** while checking emptiness:

- construct monitor in **depth-first** order
- while constructing the monitor, construct product in **depth-first** order
- while constructing the product, check reachability in **depth-first** order

This way, only the **necessary part** of the graph is built.

The on-the-fly algorithm does not change the worst-case complexity but it is likely to make model-checking **feasible in practice** – especially when a counterexample exists and can be found effectively.



# **Automata-based model checking**

---

**From programs to automata**

# Model checking for **software** analysis?

Finite-state models capture **programs** that use a **bounded** (finite and independent of input size) amount of memory.

How limiting is the restriction to **finite-state models**?

- Even when considering programs that are infinite state, a **finite-state** model can be useful to capture important **behavioral features** such as concurrency
- The **small scope hypothesis** suggests that model checking can be a very effective **bug finding** tool that provides high levels of assurance

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- The **small scope hypothesis** suggests that model checking can be a very effective **bug finding** tool that provides high levels of assurance

*Small scope hypothesis:*

*most bugs have small counterexamples*

*Daniel Jackson: Software Abstractions*



# Shared memory concurrency

Let's see how model checking can be applied to analyze properties of concurrent programs.

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The following concurrent program includes two processes that increment a counter variable in shared memory.

```
var counter: Integer // initially 0
shared memory

process t
var cnt: Integer
1 cnt := counter
2 counter := cnt + 1;
  ↑
  code

process u
var cnt: Integer ← local memory
cnt := counter      3
counter := cnt + 1  4
```

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The following **concurrent program** includes two **processes** that increment a counter variable in **shared memory**.

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```

---

	process t		process u
	var cnt: Integer		var cnt: Integer ← local memory
1	cnt := counter		cnt := counter 3
2	counter := cnt + 1;		counter := cnt + 1 4
	↑ code		

Each numbered line of code includes exactly one statement that can execute **atomically**.

Statements in different processes can be executed in **any relative order**.

# Finite-state models of concurrent processes

Analyzing the behavior of concurrent programs require to reason about a finite but very large number of different execution orders.

For example, consider the analysis question:

What is the value of `counter` after processes `t` and `u` terminate?

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What is the value of `counter` after processes `t` and `u` terminate?

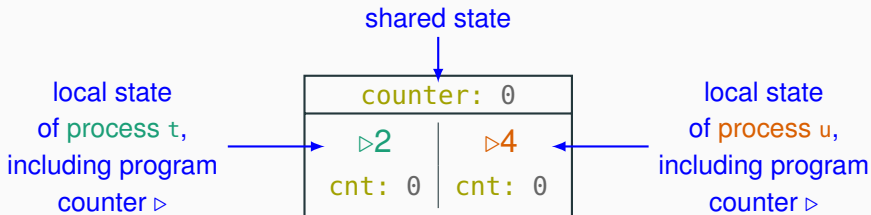
To analyze properties of concurrent programs, we formalize essential elements of their behavior using a state/transition model similar to finite-state automata.

- states in a model correspond to program states
- transitions connect states according to execution order



# Concurrent states

A **state** captures the shared and local states of a concurrent program:



```
var counter: Integer // initially 0
```

process t

```
var cnt: Integer
```

```
1 cnt := counter
2 counter := cnt + 1
```

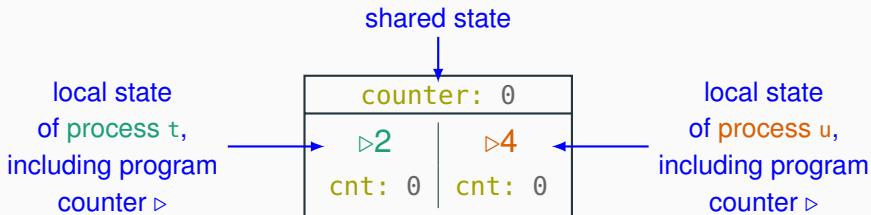
process u

```
var cnt: Integer
```

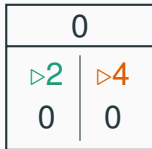
```
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```

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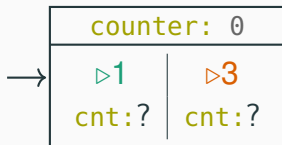


For simplicity, we may only keep a state's **essential information**:



# Initial state

As with automata, we mark the **initial state** with an incoming arrow:



```
var counter: Integer // initially 0
```

---

process t

```
var cnt: Integer
```

```
1 cnt := counter
2 counter := cnt + 1
```

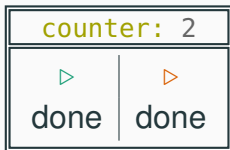
process u

```
var cnt: Integer
```

```
cnt := counter
counter := cnt + 1
```

# Final states

As with automata, the **final states** of a computation – when the program terminates – are marked with double-line edges:



```
var counter: Integer // initially 0
```

---

process t

```
var cnt: Integer
```

```
1 cnt := counter
2 counter := cnt + 1
```

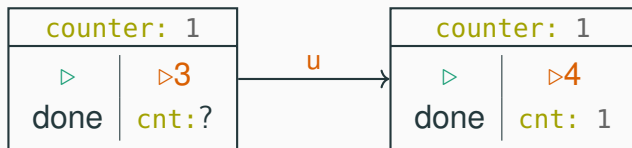
process u

```
var cnt: Integer
```

```
cnt := counter           3
counter := cnt + 1       4
```

# Transitions

A **transition** corresponds to the execution of one atomic instruction, and it is an arrow connecting two states (or a state to itself):



```
var counter: Integer // initially 0
```

process t

```
var cnt: Integer
```

```
1 cnt := counter
2 counter := cnt + 1
```

process u

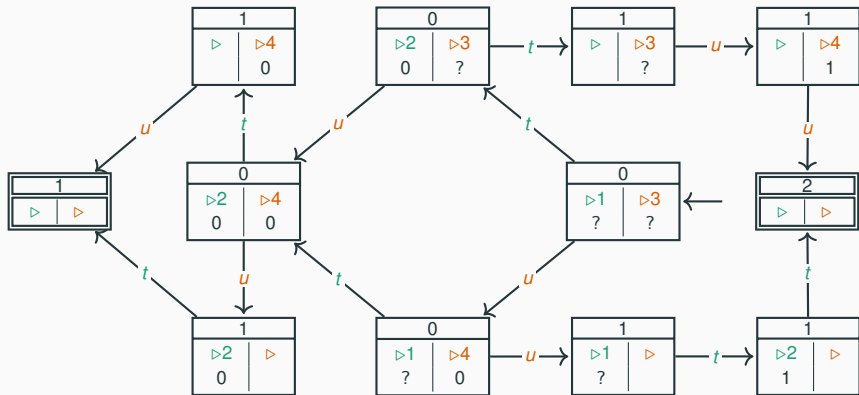
```
var cnt: Integer
```

```
cnt := counter
counter := cnt + 1
```

3  
4

# A complete state/transition model

The **complete** state/transition model for the concurrent counter example explicitly shows **all possible interleavings**:



The labels on transitions indicate which process executes, but that information is subsumed by the transition's pre- and post-state.

# Kripke structure

The state/transition model we have built is a **Kripke structure** (also known as state/transition diagram, or finite-state machine):

A **Kripke structure**  $K$  is a tuple  $\langle S, I, R, P, L \rangle$ :

- $S$ : finite nonempty set of **states**
- $I \subseteq S$ : set of **initial** states
- $R \subseteq S \times S$ : transition **relation**
- $P$ : a set of **propositions**
- $L: S \rightarrow \wp(P)$ : **labeling** function

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The **labeling** function assigns to each state  $s \in S$  the set  $L(s) \subseteq P$  of **propositions** that **hold** in  $s$ .

# From Kripke structures to finite-state automata

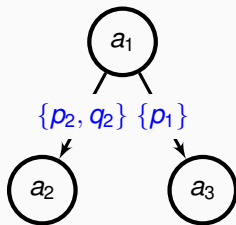
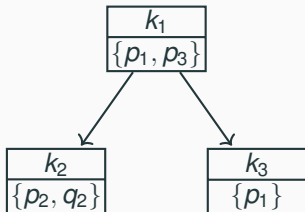
We can represent the computations of a Kripke structure  $K = \langle S_K, I_K, R_K, P_K, L_K \rangle$  with an automaton  $A = \langle \Sigma, S, I, F, \rho \rangle$ .

The basic idea is that we label each automata transition with a set of propositions corresponding to the propositions that label the state the transition enters.

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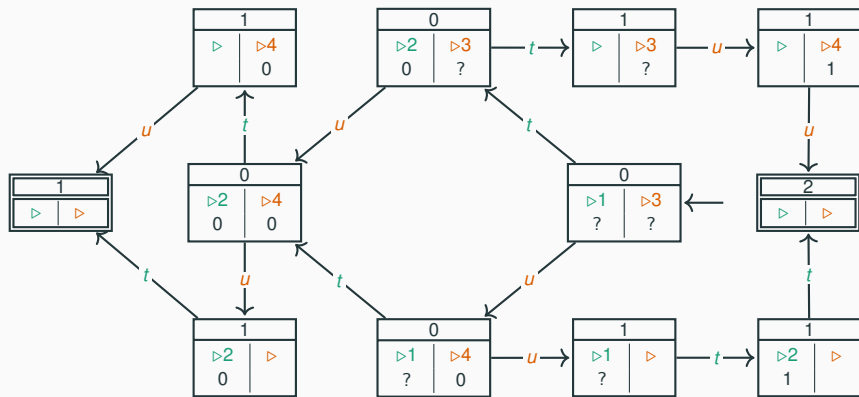
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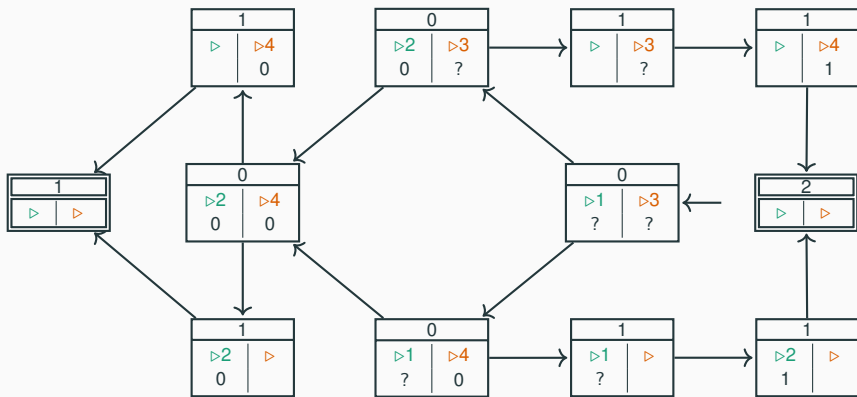
The basic idea is that we label each **automata transition** with a **set of propositions** corresponding to the propositions that label the state the transition **enters**.

- $\Sigma = \wp(P_K)$
- $S = S_K \cup \{s_0\}$   
 $s_0$  is an **initialization** state that leads to initial states of  $K$
- $I = \{s_0\}$
- $F = \{s \in S_K \mid \neg(s R s_2) \text{ for all } s_2 \text{ and } s \text{ is not an error state}\}$   
the final states are those without **outgoing** transitions
- $s \in \rho(s_0, L(s))$  iff  $s \in I_K$   
the initialization state leads to  $K$ 's **initial states** with an input corresponding to the proposition that hold there
- $s_2 \in \rho(s_1, L(s_2))$  iff  $s_1 R s_2$ : each transition takes an input corresponding to the proposition that hold in the state it **reaches**

# Kripke structure to finite-state automaton: example



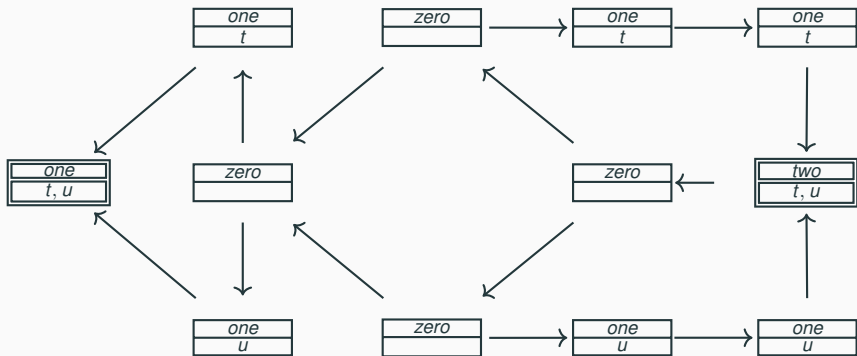
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We keep track of **predicates**:

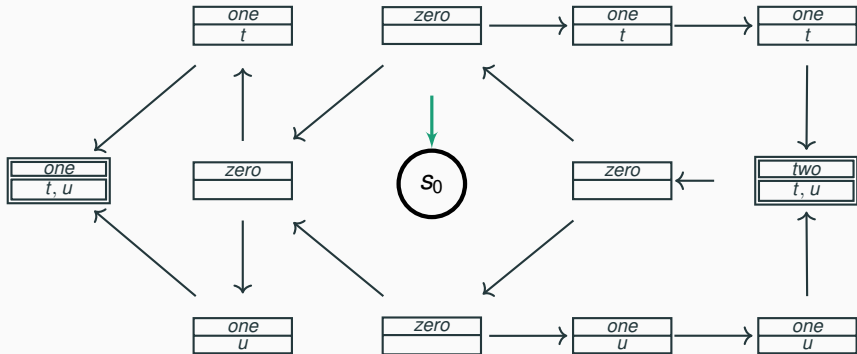
- *zero, one, two*: corresponding to the value of the **counter**
- *t, u*: process *t, u* has **terminated**



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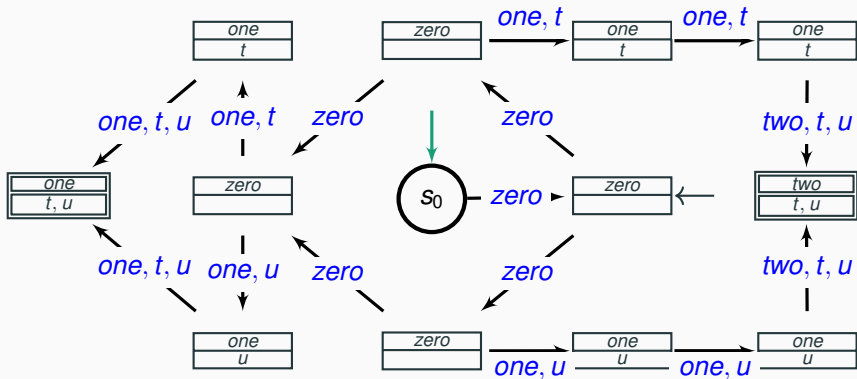




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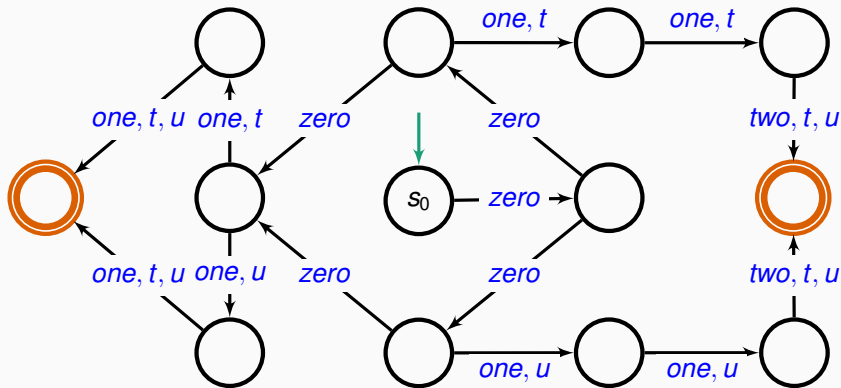
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# Model checking concurrent behavior

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$$Count \stackrel{?}{\models} Two$$

*Count*: the automaton built from the Kripke structure modeling the concurrent program

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**Count**: the automaton built from the Kripke structure modeling the concurrent program

**Two**: the LTL formula  $\Box(t \wedge u \implies two)$

Now LTL formulas are over  $\Pi = power(P)$ ; correspondingly, we interpret the **satisfaction** of a proposition at one time step as:

$$w, k \models p \qquad \text{iff} \qquad p \in w[k]$$

since each word symbol is a **set** of propositions in  $P$ .

# **Automata-based model checking**

---

**Model checking tools**

# Model checking with Spin

Model checking **tools** provide specialized languages to conveniently and concisely express **transition systems** conveniently as the composition of parallel **processes**.

**Spin** is widely used **explicit-state** model checking: it implements the automata-theoretic model checking algorithm we have seen, in the version **on the fly** and with many optimizations to support scalability.

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Spin inputs transition system models expressed in the **ProMeLa** (Process Meta Language) language, which describes transition systems and LTL properties.



# Shared counters in ProMeLa

Here is the **shared counter increment** example encoded as a transition system in ProMeLa.

```
// shared memory
```

```
int count = 0;
```

```
// boolean array
```

```
// to keep track of termination
```

```
bit done[3] = 0;
```

```
proctype inc_process() {
```

```
    int tmp;           // process-local variable
```

```
    tmp = count;
```

```
    count = tmp + 1;
```

```
    done[_pid] = 1;    // set termination bit
```

```
    // '_pid' gives the process id of the running process
```

```
}
```

```
// spawn two processes running the same code
```

```
init { run inc_process(); run inc_process(); }
```

```
// the spawned processes will have _pid == 1 and 2
```

```
// process 0 is the main entry point of the system
```

# LTL properties

We can embed the **LTL properties** we want to verify in a ProMeLa file.

We can encode predicates directly with C-style Boolean expression, or use pre-processor macros to define named predicates.

```
#define t (done[1] == 1)
```

```
#define u (done[2] == 1)
```

```
ltl Two {  
    [] (t && u -> count == 2)  
}
```

```
ltl OneOrTwo {  
    [] (t && u -> (count == 1 || count == 2))  
}
```

# Running Spin on the shared counter example

```
> spin -a unitCounter.pml      # create analyzer pan.c
> gcc -o mc_unitCounter pan.c  # compile analyzer
> ./mc_unitCounter -a -N Two   # run analyzer with property `Two'

pan:1: assertion violated
pan: wrote unitCounter.pml.trail
# this means: not (unitCounter |= Two)
> spin -k unitCounter.pml.trail unitCounter.pml # replay error trace

Process 2 reads 0
Process 1 reads 0
Process 2 writes 1
Process 2 terminates
Process 1 writes 1
Process 1 terminates
```

# Print statements

To make error traces easier to follow, we can embed **printf** statement that will be used when we replay an error trace.

To ensure a **printf** statement is printed when a process takes a certain step, we use **atomic** blocks, which are guaranteed to be **not interruptible**.

```
proctype inc_process() {
    int tmp;
    atomic {
        tmp = count;
        printf("Process_%d_reads_%d\n", _pid, tmp);
    }
    atomic{
        count = tmp + 1;
        printf("Process_%d_writes_%d\n", _pid, tmp + 1);
    }
    atomic {
        done[_pid] = 1;
        printf("Process_%d_terminates\n", _pid);
    }
}
```

# LTL properties on the command line

We can also provide LTL properties directly on the **command line** of Spin. In this case, the **given formula** is **directly** intersected with the automaton; thus we have to pass the **negated property**  $\neg P$  to verify  $P$

```
> spin -a -f ')' unitCounter.pml
> gcc -o mc_unitCounter pan.c
> ./mc_unitCounter -a      # verifies command-line property by default
```

[no errors]

```
# this means: unitCounter |= [](t && u -> (count >= 1))
```

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```

[no errors]

```
# this means: unitCounter |= [](t && u -> (count >= 1))
```

If you want a **more efficient** encoding of LTL properties, you can use a specialized tool that outputs ProMeLa code for  $\mathcal{A}(P)$ :

Oddoux and Gastin's `ltl2tgba`  
Web interface to LTL2BA

## A more interesting example

Here is a more challenging example, which is not easy to analyze without tools.

If each process **increments** the shared counter **10 times**, what is the **minimum** value of count after a complete run?

```
var counter: Integer // initially 0
```

---

process t

```
var cnt, k: Integer
```

```
1 k := 0
2 while (k < 10)
3   cnt := counter
4   counter := cnt + 1
5   k := k + 1
```

process u

```
var cnt, k: Integer
```

```
3 k := 0
4 while (k < 10)
5   cnt := counter
6   counter := cnt + 1
7   k := k + 1
```

# Running Spin on the multiple increment example

```
# compile one version of the analyzer for each bound k = 10, 9, ..., 2
> for k in $(seq 10 -1 2); do \
    spin -a -f '!'"[(t && u -> (count >= $k))" multiCounter.pml; \
    gcc -o "bound_$k" pan.c; done
> ./bound_10 -a | grep 'assertion violated'
# ...
> ./bound 2 -a | grep 'assertion violated'
> ./bound 1 -a | grep 'assertion violated'
```



# Model checking concurrent Java

Let us analyze using spin a more complex example of **concurrent code** using **Java** threads.

The following analysis is based on Hillel Wayne's, who analyzed, using the formal language TLA<sup>+</sup> and model checking, an **extreme programming** challenge posed by Tom Cargill.

```

class BoundedBuffer {
    synchronized void put(Object x)
        throws InterruptedException {
        while( occupied == buffer.length )
            wait();
        notify();
        ++occupied;
        putAt %= buffer.length;
        buffer[putAt++] = x;
    }

    synchronized Object take()
        throws InterruptedException {
        while( occupied == 0 )
            wait();
        notify();
        --occupied;
        takeAt %= buffer.length;
        return buffer[takeAt++];
    }

    private Object[] buffer = new Object[4];
    private int putAt=0, takeAt=0, occupied=0;
}

```

```
class BoundedBuffer {
```

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        ++occupied;
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    }
```

block running thread  
(waiting for condition)

unblock any one blocked thread  
(nondeterministically chosen)

```
    synchronized Object take()
        throws InterruptedException {
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        notify();
        --occupied;
        takeAt %= buffer.length;
        return buffer[takeAt++];
    }
```

monitor: threads run on shared object  
in mutual exclusion

```
    private Object[] buffer = new Object[4];
    private int putAt=0, takeAt=0, occupied=0;
}
```

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    private Object[] buffer = new Object[4];
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}

```

*Perhaps someone can show me how to modify [this] code to make it more testable, and then how to write a **test** that exposes its **bug**.*

Tom Cargill: Extreme Programming  
Challenge 14

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    private Object[] buffer = new Object[4];
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}

```

*Now that the unit tests have been written I see that there are **no bugs**. Sometimes unit tests surprise you by telling you that your code actually does work.*

Don Wells

# A ProMeLa model of Java monitors

```
// number of producer processes  
#define NPROD 1  
// number of consumer processes  
#define NCONS 2
```

# A ProMeLa model of Java monitors

```
// number of producer processes
#define NPROD 1
// number of consumer processes
#define NCONS 2

int lock = 0;           // pid of process which has the monitor's lock

bool blocked[P];       // set of blocked processes (blocked[pid] iff pid is blocked)
int nblocked = 0;      // number of blocked processes

int contains = 0;       // number of items in buffer
```

# A ProMeLa model of Java monitors

```
// number of producer processes
#define NPROD 1

// number of consumer processes
#define NCONS 2

int lock = 0;           // pid of process which has the monitor's lock

bool blocked[P];        // set of blocked processes (blocked[pid] iff pid is blocked)
int nblocked = 0;       // number of blocked processes

int contains = 0;       // number of items in buffer

proctype scheduler()
{
  int p;
  do
    // wait until no process is active on the monitor
    :: lock == 0 ->
      printf("Scheduler_running\n");
      assert (nblocked < P); // all blocked means deadlock!
      // nondeterministically select an unblocked process
  ...
}
```



# Spin verification of Java monitors

pan:1: assertion violated (nblocked<(1+2)) (at depth 339)

Simplified **error trace** (Spin processes correspond to Java threads):

1. Initially all processes  $P$  (producer) and  $C_1, C_2$  (consumers) are unblocked and the buffer (capacity 1) is empty
2.  $C_1$  finds empty buffer, blocks
3.  $C_2$  finds empty buffer, blocks
4.  $P$  writes to buffer, wakes up  $C_2$
5.  $P$  finds full buffer, blocks
6.  $C_2$  reads from buffer, wakes up  $C_1$
7.  $C_1$  finds empty buffer, blocks
8.  $C_2$  finds empty buffer, blocks
9. All processes are blocked!

## Detailed analysis using model checking

By modifying the model and running model checking again, we can evaluate different **design choices** and reason about **properties** that are nearly impossible to get right by **testing** and debugging.

- If we remove the violated assertion, we still get an **invalid end state** error, which means exactly that the system can **deadlock**.
- A deadlock seems to occur only when the number of process (producers and consumers) is **more than twice** the buffer's size.
- To avoid deadlocks it is enough to require add an **entry queue** to the ProMeLa model that processes have to go through every time they enter the monitor.

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Indeed, more recent versions of Java also offer a different, library-based implementation of monitors, which has nicer fairness properties (such as that `notify` wakes up the thread that has been waiting the longest).

# WRITING CONCURRENT SOFTWARE



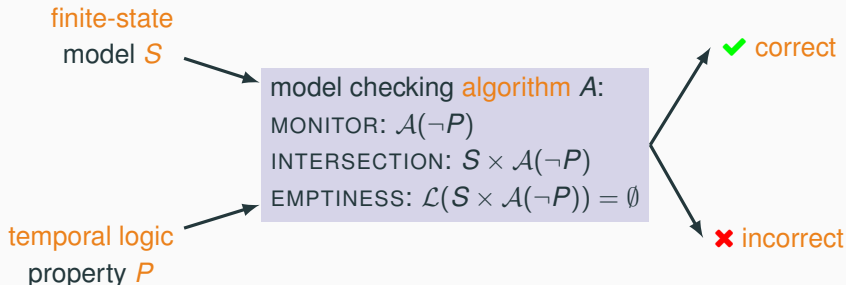
**ONLY USES TESTING    KNOWS MODEL CHECKING**

# **Automata-based model checking**

---

## **Variants of model checking**

# Flexible automata-based model checking



The **framework** of automata-based model checking is quite **flexible**, as it can accommodate several variants in:

- the kind of **automaton model** used to describe the **system**
- the flavor of **temporal logic** used to express **properties**
- the techniques used to implement the **checking algorithm**

Let's outline some interesting variants in each category.

# Variants of the model checking algorithm

There are two main families of **algorithms** to implement the monitor construction, intersection, and emptiness steps of model checking:

**explicit-state** algorithms perform the steps as algorithms **on graphs**, very similarly to how we have presented them (with practical optimizations such as applying them on the fly)

**symbolic** algorithms **encode** system model  $S$  and property  $P$  as symbolic **constraints**, and then rely on constraint solvers to perform the analysis

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**symbolic** algorithms **encode** system model  $S$  and property  $P$  as symbolic **constraints**, and then rely on constraint solvers to perform the analysis

The first symbolic model checking algorithms represented constraints using **binary decision diagrams (BDDs)**, a directed acyclic data structure that can concisely represent Boolean functions and operations such as conjunction and emptiness check.

Taking advantage of the spectacular advances in SAT-solving technology, **bounded model checking** is a different kind of **symbolic** algorithm that uses propositional logic to encode the model checking problem.



# Bounded model checking

Bounded model checking encodes automata and temporal logic as **propositional formulas**, where proposition represents the **state** a computation is in at each **step** in a run.

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We represent a run  $r[0] r[1] \dots r[n]$  using  $2 \cdot (n + 1)$  **propositions**  $a[k], b[k]$  for  $k = 0, \dots, n$ , where  $a[k]$  denotes that  $a \in r[k]$  holds at  $k$ .

The **runs** of **length** up to  $n$  are encoded as:

$$\text{runs}(K, n) = \text{init}(0) \wedge N(n) \wedge \bigwedge_{0 \leq k < n} (\neg N(k) \implies T(k, k + 1))$$

$\text{init}(k)$  is the indicator function of the set of **initial states**

$N(k)$  denotes that the run has length **up to**  $k$

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*Note: A blue arrow points from the expression  $\neg N(k) \Rightarrow T(k, k+1)$  to the text 'step k is taken unless the run has ended before k steps'.*

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# Encoding runs as propositional formulas: example



Let's encode all runs of length **up to bound 2**.

a transition is taken unless the run has ended before

$$runs(K, 2) = init(0) \wedge N(2) \wedge (\neg N(0) \Rightarrow T(0, 1)) \wedge (\neg N(1) \Rightarrow T(1, 2))$$

$$init(0) = \neg a[0] \wedge \neg b[0]$$

$$T(0, 1) = (a[1] \iff \neg a[0]) \wedge (b[1] \iff (a[0] \oplus b[0]))$$

$$T(1, 2) = (a[2] \iff \neg a[1]) \wedge (b[2] \iff (a[1] \oplus b[1]))$$

a switches value in every transition

b becomes true after a XOR b

$$N(k) = \bigvee_{0 \leq t \leq k} length(t) \quad \text{the run's length is encoded in binary using } \ell_1, \ell_0$$

$$length(0) = \neg \ell_1 \wedge \neg \ell_0 \quad length(1) = \neg \ell_1 \wedge \ell_0 \quad length(2) = \ell_1 \wedge \neg \ell_0$$

# Encoding monitors as propositional formulas

Using a similar formalization, we can also express **monitors** of LTL formulas as **propositional** formulas.

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For example, for property  $P = \diamond(a \wedge b)$ :

$$\text{runs}(P, n) = (a[0] \wedge b[0]) \vee \bigvee_{0 \leq t < n} (\neg N(t) \wedge a[t+1] \wedge b[t+1])$$

# Bounded model checking

In more general cases, the encoding has to be a bit more complex because  $runs(K, n)$  should encode runs of length up to  $n$  or with loops of length up to  $n$ .

In our example, however, there are no loops of length up to 2.

# Bounded model checking

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In our example, however, there are no loops of length up to 2.

Once we have defined the encoding, the model checking algorithm is straightforward to implement:

**monitor** constructs  $runs(\neg P, n) = \neg runs(P, n)$   
complementing the property is negation

**intersection** is conjunction:  $runs(K, n) \wedge \neg runs(P, n)$

**emptiness** is satisfiability:  $K \models P$  iff  $runs(K, n) \wedge \neg runs(P, n)$  is unsatisfiable



## Bounded model checking: completeness

Using a **finite bound** on run length does not affect **completeness** of bounded model checking.

Every finite-state model  $K$  has a computable **diameter**  $D_K$ :  
**all runs** of  $K$  have length up to  $D_K$  or a loop of length up to  $D_K$ .

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Therefore, bounded model checking works as follows:

```
k = 0
while k <= D_K:
    if SAT(runs(K, k) ∧ ¬runs(P, k)):
        return (K ⊈ P, counterexample)
    else:
        k = k + 1
return K ⊨ P # no counterexample beyond diameter
```

This algorithm has the additional advantage that it always returns counterexamples of **minimal length**.

## Different kinds of automata

Among the kinds of finite-state models used for model checking, we have seen **Kripke structures** and their correspondence to finite-state automata.

We will also see **timed automata** as an example of **infinite-state** model that can still be analyzed using the model checking approach since its emptiness problem is still decidable.

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A popular variant of finite-state automata are **Büchi** automata: finite-state automata that run over **infinite** words.


Infinite words are useful abstraction when **termination** is not an interesting event (such as in **reactive systems**) or is a potentially error (such as in **deadlocks**).

Since these are classic applications, most presentation of model checking use Büchi automata instead of finite-state automata.

# Büchi automata: semantics

Büchi automata have the same **syntax**  $B = \langle \Sigma, S, I, F, \rho \rangle$  as finite-state automata but are interpreted over **infinite words**:

An **infinite word** is an input sequence of unbounded length:


$$w = w[1] w[2] \dots w[n] \in \Sigma^\omega$$


infinite sequences of elements in  $\Sigma$

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A **run** of  $B$  over  $w$  is a sequence of states


$$r = r[0] r[1] \dots \in S^\omega$$

that **starts** from an **initial** state and **follows**  $B$ 's **transitions**.

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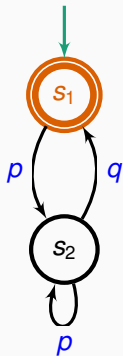
$$r = r[0] r[1] \dots \in S^\omega$$

that **starts** from an **initial** state and **follows**  $B$ 's **transitions**.

A run  $r$  of  $B$  is **accepting** if it traverses **infinitely often** a **final** state:

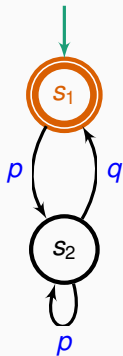
$$r[k] \in F \text{ for infinitely many value of } k.$$

## Büchi runs: example



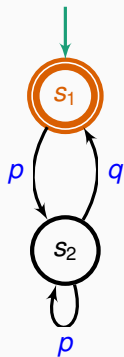


## Büchi runs: example



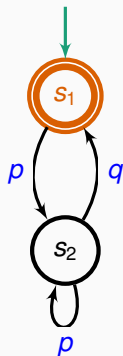
Run  $r = s_1 s_2 s_2 s_1 s_2 s_2 s_1 \dots$

## Büchi runs: example



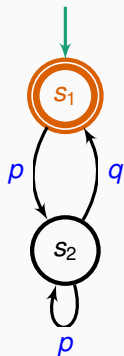
Run  $r = s_1 s_2 s_2 s_1 s_2 s_2 s_1 \dots$   
over  $w = ppqppq\dots$

## Büchi runs: example



Run  $r = s_1 s_2 s_2 s_1 s_2 s_2 s_1 \dots$   
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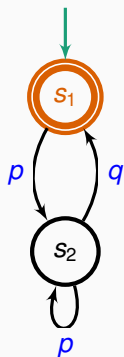
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Run  $r = s_1 s_2 s_2 s_2 s_2 \dots$

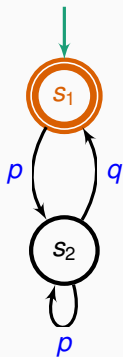
# Büchi runs: example



Run  $r = s_1 s_2 s_2 s_1 s_2 s_2 s_1 \dots$   
over  $w = ppqppq\dots$   
is accepting.

Run  $r = s_1 s_2 s_2 s_2 s_2 \dots$   
over  $w = pppp\dots$

## Büchi runs: example



Run  $r = s_1 s_2 s_2 s_1 s_2 s_2 s_1 \dots$   
over  $w = ppqppq\dots$   
is accepting.

Run  $r = s_1 s_2 s_2 s_s s_2 \dots$   
over  $w = pppp\dots$   
is not accepting.

# Büchi vs. finite state automata

There are a few technical differences in the kinds of **properties** Büchi rather than finite-state automata have:

FINITE-STATE AUTOMATA	BÜCHI AUTOMATA
closed under determinization	nondeterministic more expressive than deterministic
complementing is easy through determinization	complementing is hard (cannot always determinize)
intersection is easy	intersection is a bit trickier because of the acceptance condition
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Kripke structures can also be interpreted over infinite words. Then, all states of a Büchi automaton representing the infinite-length runs of a **Kripke** structure are **accepting** to ensure progress.



# Model checking Büchi automata

The automata-based model checking **algorithm** needs a few **adjustments** to work with Büchi automata.

**MONITORS** follow the semantics of LTL over **infinite words**, which is actually simpler because there is no **word length** to worry about. It is easier to **complement** the property rather than the monitor, due to the difficulties of complementing Büchi automata.

**INTERSECTION** uses a product construction that keeps track of the **acceptance condition** of  $A$  and of  $B$  simultaneously.

**EMPTINESS:**  $B$  is not empty iff there exists a path to a **cycle** containing an **accepting** state. This can be checked efficiently by finding the **maximal strongly connected components** of  $B$  and checking whether they contain an accepting state.

A **strongly connected component** is a subgraph of a directed graph where every node is reachable from any other node.

# Model checking Büchi automata

The automata-based model checking **algorithm** needs a few **adjustments** to work with Büchi automata.

Despite the technical differences, the **complexity** of model checking Büchi automata is essentially the same as the complexity of finite-state automata:

Model checking LTL properties  
of Büchi automata is **PSPACE**-complete.

Algorithms have complexity **linear** in the size of the automaton  
and **exponential** in the size of the property.

# Different kinds of temporal logic

LTL (linear-time temporal logic) is the most intuitive variant of temporal logic.

We will see an LTL extension called **MTL**, which adds **quantitative time constraints**.

The other main kind of temporal logic is **branching-time** temporal logic. Let's have a look at **CTL** (Computation Tree Logic) – a popular version of branching-time logic.

Formulas of propositional **CTL** are defined as:

$F ::= p \mid \neg F \mid F_1 \wedge F_2 \mid F_1 \vee F_2 \mid F_1 \implies F_2$  (propositional connectives)  
|  $\exists X F \mid \exists \Box F \mid \exists \Diamond F \mid F_1 \exists U F_2$  (existential connectives)  
|  $\forall X F \mid \forall \Box F \mid \forall \Diamond F \mid F_1 \forall U F_2$  (universal connectives)

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Each temporal connective has two **variants**:

**existential**: its arguments holds on **some** run from the current step

**universal**: its arguments holds on **all** runs from the current step

For this reason,  $\forall$  and  $\exists$  in branching-time formulas are called **path quantifiers**, since they quantify over **paths** (that is, runs).

## CTL: satisfaction relation

CTL formulas are interpreted directly on some form of transition system – typically a Kripke structure  $K = \langle S, I, R, P, L \rangle$  in the infinite-word interpretation.

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A structure  $K$  generates runs that satisfy CTL formula  $F$  at state  $s$ , written  $K, s \models F$ , iff:

$K, s \models p$	iff	$p \in L(s)$
$K, s \models \neg F$	iff	$K, s \not\models F$
$K, s \models F_1 \wedge F_2$	iff	$K, s \models F_1$ and $K, s \models F_2$
$K, s \models F_1 \vee F_2$	iff	$K, s \models F_1$ or $K, s \models F_2$

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$K, s \models \exists X F$       iff      for some  $s_1 : s R s_1$  and  $K, s_1 \models F$

$K, s \models \forall X F$       iff      for all  $s_1$  such that  $s R s_1 : K, s_1 \models F$

Intuitively:

$K, s \models \exists X F$       a state where  $F$  holds is reachable from  $s$  in one step

$K, s \models \forall X F$        $F$  holds in all states reachable from  $s$  in one step



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$K, s \models \exists \Box F$  iff for some run  $r = s s_1 s_2 s_3 \dots$  following  $R$ :  
for all states  $s_k \in r$ :  $K, s_k \models F$

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Intuitively:

$K, s \models \exists \Box F$  there is a run from  $s$  where  $F$  holds always

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for some state  $s_k \in r$ :  $K, s_k \models F$

Intuitively:

$K, s \models \exists \Diamond F$  there is a run from  $s$  where  $F$  holds eventually  
 $K, s \models \forall \Diamond F$   $F$  holds eventually in all runs from  $s$

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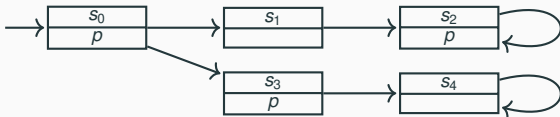
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for all  $s_h \in s s_1 s_{k-1}$ :  $K, s_h \models F_1$

$K, s \models F_1 \forall U F_2$  iff for all runs  $r = s s_1 s_2 s_3 \dots$  following  $R$ :  
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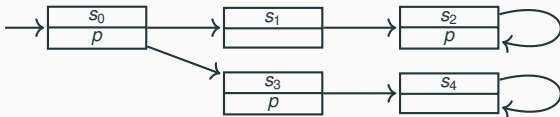
$K, s \models F_1 \exists U F_2$     there is a run from  $s$  where  $F_1 U F_2$  holds at  $s$   
 $K, s \models F_1 \forall U F_2$      $F_1 U F_2$  holds at  $s$  in all runs from  $s$

## CTL semantics: examples



$$K, s_0 \models \exists X(\neg p)$$

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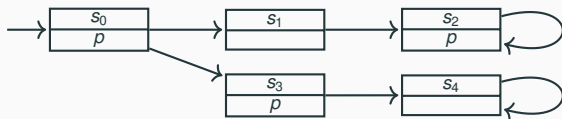


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# CTL semantics: examples



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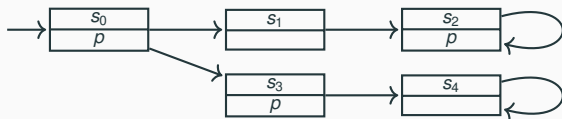


$K, s_0 \models \forall X(\neg p)$



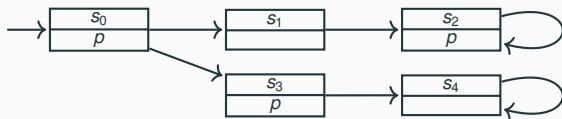
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# CTL semantics: examples



$K, s_0$	$\models$	$\exists X(\neg p)$	✓
$K, s_0$	$\models$	$\forall X(\neg p)$	✗
$K, s_2$	$\models$	$\forall X p$	✓
$K, s_0$	$\models$	$\exists \Diamond(\neg p)$	

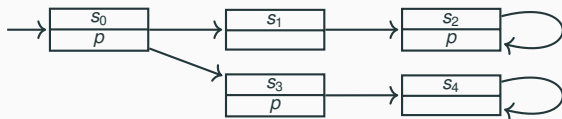
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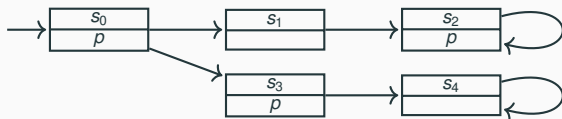


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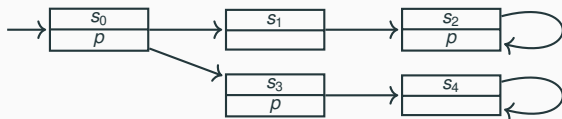
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# CTL semantics: examples



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# CTL semantics: examples



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$K, s_0$	$\models$	$\Diamond(p \wedge X p)$	✓

→  
LTL property

# LTL vs. CTL

There has been a long-standing debate about whether linear-time or branching-time logic is “better”:

**intuitiveness:** LTL formulas are generally easier to understand

**expressiveness:** LTL and CTL have **incomparable** expressive power

- $\forall \Diamond(p \wedge \forall X p)$  is inexpressible in LTL
- $\Diamond(p \wedge X p)$  is inexpressible in CTL

**complexity:** CTL model checking is **P**-complete

LTL model checking is **PSPACE**-complete

**algorithms:** CTL model checking can be solved in  $O(|A| \cdot |F|)$

LTL model checking can be solved in  $O(|A| \cdot 2^{|F|})$

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The complexity gap is not really an issue in practice:

- Formulas  $F$  are normally **small**
- CTL and LTL model checking tends to have similar performance for formulas that are **expressible** in both logics
- CTL's performance advantage vanishes when model checking **open** systems

# Linear temporal logic with past operators

LTL with past (PLTL) adds temporal connectives that refer to the past:

$$F ::= \Upsilon F \mid \Box F \mid \Diamond F \mid F_1 S F_2 \quad (\text{past temporal connectives})$$

# Linear temporal logic with past operators

yesterday/in the  
previous step

always in  
the past

LTL with past (**PLTL**) adds temporal connectives that refer to the past:

$F ::= YF \mid \Box F \mid \Diamond F \mid F_1 S F_2$  (past temporal connectives)

eventually/sometimes  
in the past

since

# PLTL: complexity and expressiveness

Some properties are **easier** to express using past operators.

For example: “every alarm is due to a fault”

LTL	PLTL
$\neg(\neg fault \cup (alarm \wedge \neg fault))$	$\square (alarm \implies \overleftarrow{\diamond} fault)$



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However, PLTL has the same **expressiveness** as LTL: every PLTL formula has an equivalent LTL formula

In some pathological cases, the LTL formula equivalent to some PLTL formula  $P$  has size **exponential** in  $|P|$ . That is, PLTL is exponentially more **succinct** than LTL.

# **Automata-based model checking**

---

## **History and tools**

# Model checking tools

Some notable **model checking** tools:

**Spin** is a versatile open-source **explicit-state** model checker. It was originally developed at Bell Labs in the 1980s, and has later been taken over to NASA's JPL by its inventor Gerard Holzmann.

**NuSMV** is a state-of-the-art **symbolic** model-checker based on SAT solving. It is a complete reimplementation of SMV, which was the first symbolic model checkers (based on BDDs).

**TLA<sup>+</sup>** is a **specification** language for systems and temporal properties based on logic, which extends Lamport's TLA (Temporal Logic of Actions). TLA<sup>+</sup>'s toolset includes a model checker as well as interactive provers.

The Alloy Analyzer is sometimes referred to as a model checker, but it does not directly apply the automata-based approach of the other tools.

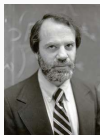
# A brief history of model checking

Most of the **ingredients** of model checking were introduced by **logicians** decades before they were used in computer science.

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Kripke structures	1963	Saul Kripke
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Büchi automata	1960	Julius Büchi
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Temporal (tense) logic	1957	Arthur Prior
------------------------	------	--------------







Temporal logic	1968	Hans Kamp
----------------	------	-----------



# A brief history of model checking

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	Swiss, ETH graduate		
Kripke structures	1963	Saul Kripke	
Büchi automata	1960	Julius Büchi	
Temporal (tense) logic	1957	Arthur Prior	
Temporal logic	1968	Hans Kamp	

# A brief history of model checking

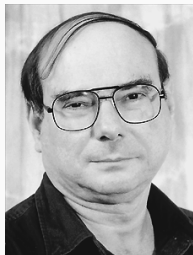
Model checking brought those abstract ideas to practical fruition.



# A brief history of model checking

Model checking brought those abstract ideas to practical fruition.

Around 1977, Amir Pnueli proposed to use temporal logic to express program properties.



# A brief history of model checking

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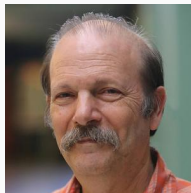
In 1981, the first **model checking** techniques were invented by **Clarke** and **Emerson**, and, independently, Queille and **Sifakis**.



# A brief history of model checking

Model checking brought those abstract ideas to practical fruition.

The automata-theoretic framework was introduced by Vardi and Wolper around 1986.



# A brief history of model checking

Model checking brought those abstract ideas to practical fruition.

Scalable model checking tools first appeared to the general public at the beginning of the 1990s.

For example, Holzmann's Spin  
and McMillan's SMV.



# Software model checking

---

# Model checking software?

In its most general meaning, **software model checking** denotes techniques that apply **model checking** to **real software** (actual code).



Model checking denotes a family of techniques for the algorithmic verification of **finite-state systems** with temporal-logic specifications.

# Model checking software?

In its most general meaning, **software model checking** denotes techniques that apply **model checking** to **real software** (actual code).



Model checking denotes a family of techniques for the algorithmic verification of **finite-state systems** with temporal-logic specifications.

Most software is **not finite** state:

- arbitrary size **inputs**
- dynamic **memory** management
- unbounded **recursion**

# Automatic software model checking with CEGAR

We have seen that finite-state models can capture **important properties** of real software – such as concurrent behavior – and can explore **bounded behavior** to systematically find bugs.



# Automatic software model checking with CEGAR

We have seen that finite-state models can capture **important properties** of real software – such as concurrent behavior – and can explore **bounded behavior** to systematically find bugs.

In this lecture, however, we present a specific approach to software model checking that is:

- fully **automatic** – in particular, it builds **finite-state models** of software automatically
- analyzes **arbitrary (executable) assertions** of **generic software behavior** – it is not limited to concurrency or other aspects

**CEGAR** (CounterExample-Guided Abstraction Refinement) is a framework to automatically **model check** real **software**.

## High level view of CEGAR model checking

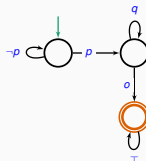
## CONCRETE PROGRAM

```

    __m512i_invariant_m = __m512i_loadu(invariant_ptr, INCREMENT_BY_SIZE);
    // add invariant (assumed to be of type Bgr)
    // to every sample while loop in the first procedure
    if (invariant_type == Bgr)
    {
        __m512i_invariant = __m512i_loadu(invariant_ptr, INCREMENT_BY_SIZE);
        // build sample while with invariant
        create_loop_invariant_m(
            loop_invariant_ptr, INCREMENT_BY_SIZE, invariant_ptr,
            loop_invariant_ptr, INCREMENT_BY_SIZE, INCREMENT_BY_SIZE);
        loop_invariant_ptr += INCREMENT_BY_SIZE;
        // build sample while loop in procedure
        create_loop_invariant_m(
            loop_invariant_ptr, INCREMENT_BY_SIZE, INCREMENT_BY_SIZE,
            loop_invariant_ptr, INCREMENT_BY_SIZE, INCREMENT_BY_SIZE);
    }
    // for every while loop in procedure
    for (procedure_start = 0; procedure_start < 1000; procedure_start++)
    {
        // build procedure after
        if (procedure_ptr[procedure_start].type == W0000000000)
        {
            // build loop completion while
            create_loop_completion_while(
                loop_invariant_ptr, loop_invariant_ptr, loop_invariant_ptr);
            // add loop_invariant as first child
            loop_invariant_ptr = loop_invariant_ptr + INCREMENT_BY_SIZE;
        }
    }
}

```

## PREDICATE ABSTRACTION



concrete

abstract

## High level view of CEGAR model checking

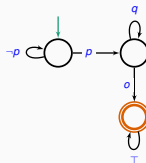
## CONCRETE PROGRAM

[illegible]

can execute (test)

concrete

## PREDICATE ABSTRACTION



can model check

abstract

## High level view of CEGAR model checking

## CONCRETE PROGRAM

```

def __init__(self, h, n_channels, log, invariant):
    """
    -- invariant (assumed) to be of type bool
    -- to every sample while loop in the first procedure

    invariant: bool = True
    """
    self.invariant_type = bool

    log_invariant = PRODUCTION_FLAG
    log_invariant_start = PRODUCTION_FLAG

    -- build complex while with invariant
    while log_invariant and log_invariant_start:
        log_invariant and self.__invariant__(PRODUCTION_FLAG, self, with_value
                                           PRODUCTION_FLAG, "invariant")

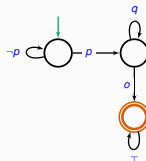
    log_invariant and self.__invariant__(PRODUCTION_FLAG, self, with_value
                                        LOGGING_LEVEL, "LOGGING_LEVEL")

    -- for every while loop in procedure
    for procedure.start (start):
        self.procedure after

    if procedure (new child) is type == WOLGEM then
        -- build new complex while
        create log_invariant_start (log_invariant)
        -- add __invariant__ as first child
        log_invariant and self.__invariant__(PRODUCTION_FLAG, self, child)

```

## PREDICATE ABSTRACTION



## 1. model check abstraction

concrete

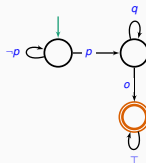
abstract

## High level view of CEGAR model checking

## CONCRETE PROGRAM

[illegible]

## PREDICATE ABSTRACTION



✗  
counterexample

## 1. model check abstraction

concrete

abstract

## High level view of CEGAR model checking

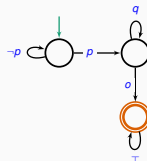
## CONCRETE PROGRAM

[illegible]

2. execute counterexample

✗ counterexample

## PREDICATE ABSTRACTION



## 1. model check abstraction

concrete

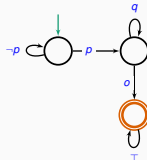
abstract

## High level view of CEGAR model checking

counterexample  
not executable

## CONCRETE PROGRAM

## PREDICATE ABSTRACTION

[illegible]

2. execute  
counterexample

counterexample

# 1. model check abstraction

concrete

abstract



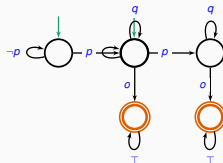


## High level view of CEGAR model checking

## CONCRETE PROGRAM

[illegible]

## PREDICATE ABSTRACTION



concrete

abstract

## High level view of CEGAR model checking

The loop continues **until**:

## CONCRETE PROGRAM

```

    __m128i_invariant_m_half_lo, __m128i_invariant_m_half_hi);
    // -- add invariant (assumed to be of type Bgr)
    // -- To every sample while loop in the first procedure
    register
    __m128i_invariant_type = Bgr;

    loop_invariant = PRODUCTION_GLASS;
    loop_invariant_star = PRODUCTION_GLASS;

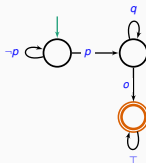
    // -- build sample while with invariant
    create_loop_invariant_and_star(
        loop_invariant_and_star, create(
            PRODUCTION_GLASS) with value
            loop_invariant_star, loop_invariant_star);

    loop_invariant_and_star_child = loop_invariant;
    loop_invariant_and_star_child_star = create(
        PRODUCTION_GLASS) with value
        loop_invariant_star, loop_invariant_star);

    // -- for every while loop in procedure
    for procedure_start (start)
    while procedure_after
    if procedure (new child_id()) type = WOLGMS then
        // -- build new sample while with invariant
        create_loop_invariant_and_star(
            loop_invariant_and_star, create(
                loop_invariant_star, loop_invariant_star) as first child
            loop_invariant_star, loop_invariant_star);
    end if procedure (new child_id()) type = WOLGMS then

```

## PREDICATE ABSTRACTION



start over  
on new abstraction

concrete

abstract

# High level view of CEGAR model checking

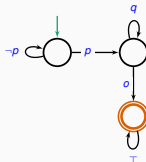
The loop continues **until**:

- the program is verified **correct**

## CONCRETE PROGRAM

```
while_invariant to all while loop do invariant: PREDICATE_CLASS do
  all invariant checked to be of type Expr()
  -- for every single while loop in the first procedure
  loop_invariant_type = Expr
loop
  loop_invariant: PREDICATE_CLASS
  loop_invariant_class: PREDICATE_CLASS
  -- build complete class with invariant
  create loop_invariant make (loop_invariant)
  loop_invariant_and_child: create (PREDICATE_CLASS) make with value
    (loop_invariant, "invariant")
  loop_invariant_and_child: create (PREDICATE_CLASS) make with value
    (loop_invariant, "T")
  -- for every while loop in procedure
  loop procedure start (loop)
  until procedure after
  if procedure (loop_invariant()) type = nil then
    -- build the complete class
    create loop_invariant_after make (loop_invariant)
    -- add all children to the child
    loop_invariant_after_and_child: procedure (loop_invariant())
```

## PREDICATE ABSTRACTION



✓  
program is correct

1. model check  
abstraction

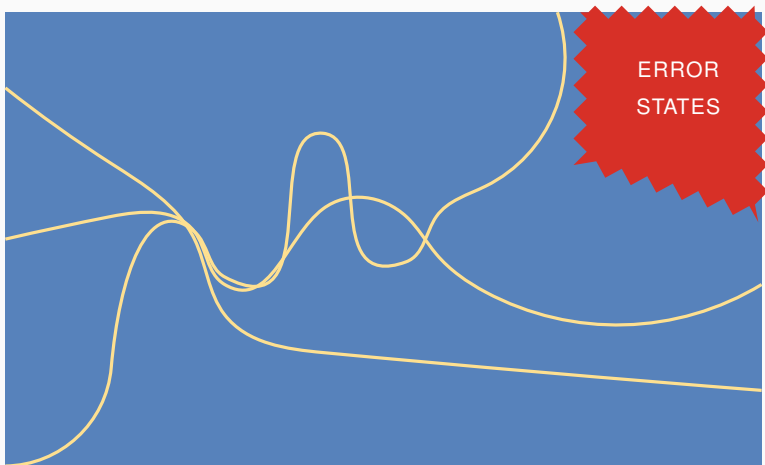
concrete

abstract



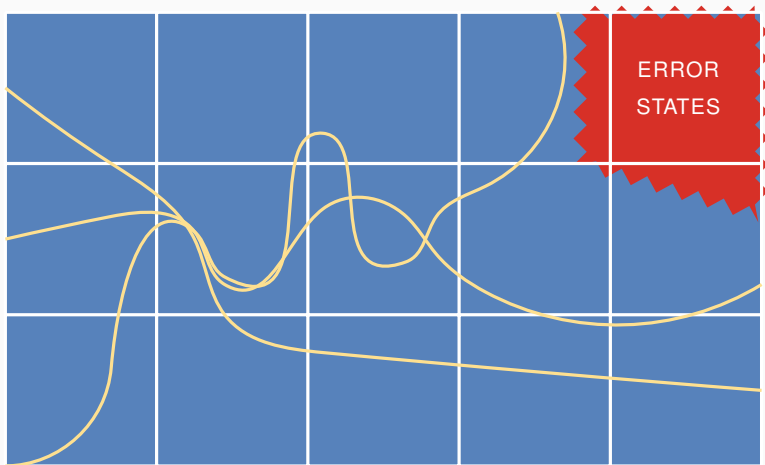


# CEGAR model checking: overview



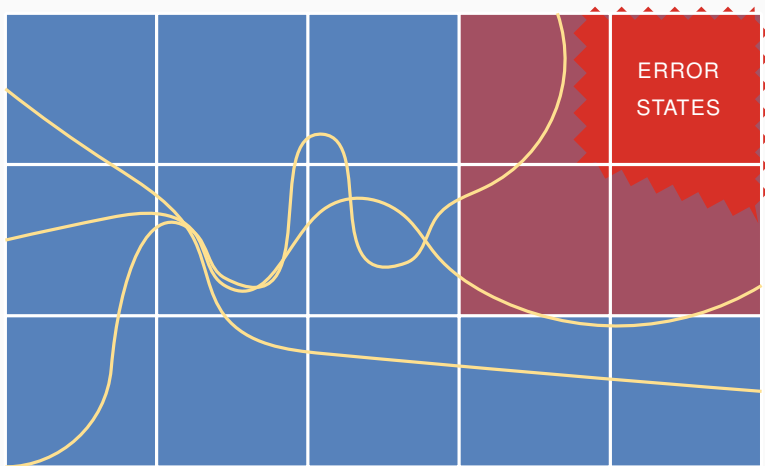
As usual, we want to analyze the **behavior** of a program, checking whether any runs enter **error states**.

# CEGAR model checking: overview



Instead of analyzing the program's infinite-state behavior, we build a **finite-state** abstraction of the program.

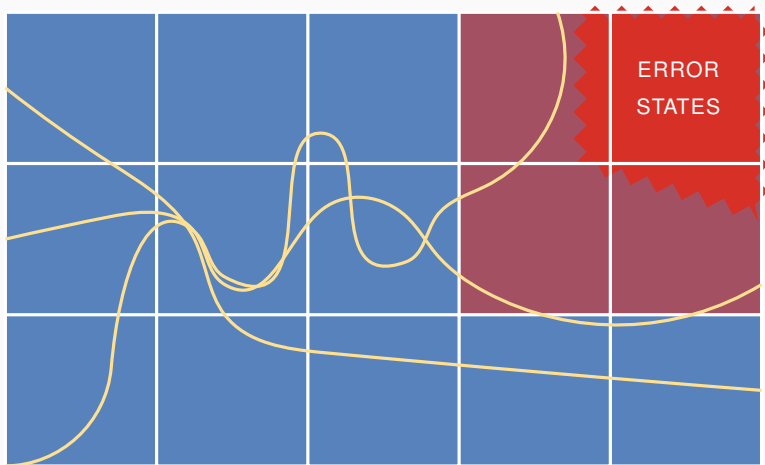
# CEGAR model checking: overview



The finite-state abstraction is an **over approximation** of the program's behavior, and hence it is **sound** but introduces **imprecision**.

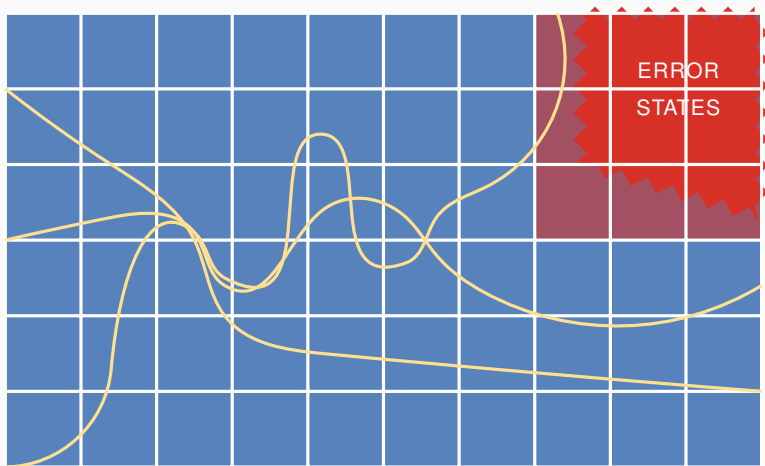


# CEGAR model checking: overview



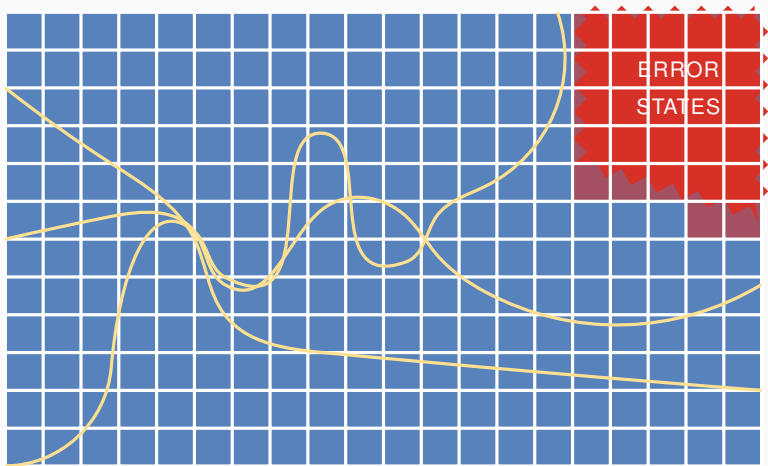
Because of imprecision, some **counterexamples** produced by model checking the finite-state abstraction may be **spurious** – that is **false positives** due to the abstraction's imprecision.

# CEGAR model checking: overview



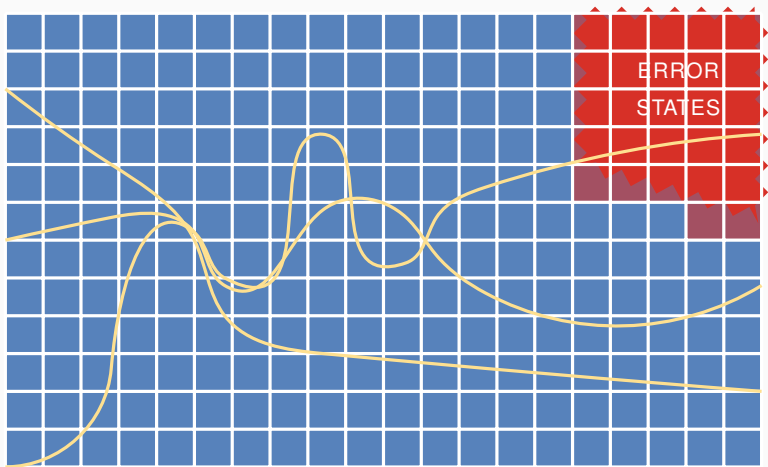
If we detect a spurious counterexample, we can **refine** the abstraction by making it **more precise** – but still finite state.

# CEGAR model checking: overview



We can continue to refine until we **verify** the abstraction is **correct**.

# CEGAR model checking: overview



Or until we find a **true counterexample** which reveals a real bug.

# CEGAR software model checking

CEGAR model checking of software combines three program analysis techniques:

**PREDICATE ABSTRACTION** to build finite-state abstractions of programs that over approximate real behavior

**SPURIOUS COUNTEREXAMPLE detection** to determine whether an abstract counterexample is spurious or executable on the real program

**PREDICATE DISCOVERY** to refine the abstraction making it more precise

# Software model checking

---

## Predicate abstraction

# Nondeterminism

To encode **over approximations**, we sometimes use **nondeterministic value** `?` as expressions in our programs.

Using **value** `?` corresponds to introducing **several computations** – one for every value that `v` may take according to its type.

```
var b: Boolean
b := ?
```

corresponds to **two computations**:

- one where `b := true` executes
- another one where `b := false` executes

```
var v: Integer
if ?
```

```
    v := 3
```

```
else
```

```
    v := 10
```

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- one where `v` is assigned value 3
- another one where `v` is assigned value 10

Nondeterminism of this kind immediately translates to nondeterminism in a **finite-state transition system** or **automaton**.



# Predicate sets

A predicate set  $B$  is a set of propositional symbols

$$B = \{b_1, b_2, \dots, b_m\}$$

where proposition  $b_k$  corresponds to a quantifier-free predicate over program variables.

Example:  $PE = \{(x \geq 0)^{\text{pos}}, (y = x)^{\text{eq}}\}$  introduces symbol `pos` for predicate  $x \geq 0$ , and symbol `eq` for predicate  $y = x$ .

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When referring to a generic predicate set  $\{b_1, b_2, \dots, b_m\}$  we **overload** the notation so that  $b_k$  may indicate a propositional symbol or a predicate – which one it is will be always clear from the context.

# Boolean programs

A **Boolean program** is a program that only uses **Boolean variables**.

A **Boolean program over** predicate set  $B$  is a program that **only uses Boolean variables** whose names are the propositional symbols in  $B$ .

In this presentation, a program is any **Helium** program – limited to integer and Boolean variables for simplicity.

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In this presentation, a program is any **Helium** program – limited to integer and Boolean variables for simplicity.

Example Boolean program over  $PE = \{(x \geq 0)^{\text{pos}}, (y = x)^{\text{eq}}\}$ :

```
var pos, eq: Boolean
if pos
  eq := true
else
  eq := ?
```

# Abstract and concrete runs

A **run** is a **sequence of states**  $s[0] s[1] s[2] \dots s[n]$  that a complete program's execution goes through.

A **program**  $H$  defines a set  $\mathcal{L}(H)$  of **runs** – all possible executions of  $H$  according to its operational semantics.

- If  $H$  is a **Boolean program** its runs are called **abstract runs** and are sequences of abstract states
- If  $H$  is a **concrete program** its runs are called **concrete runs** and are sequences of concrete states

```
var pos, eq: Boolean
```

```
if pos
```

```
  eq := true
```

```
else
```

```
  eq := ?
```

```
var x, y: Integer
```

```
if x ≥ 0
```

```
  y := x
```

```
else
```

```
  y := -x
```

Some abstract runs:

$a_1 = (\text{pos}, \text{eq}) (\text{pos}, \text{eq})$

$a_2 = (\neg \text{pos}, \text{eq}) (\neg \text{pos}, \text{eq})$

Some concrete runs:

$c_1 = (x = 3, y = 0) (x = 3, y = 3)$

$c_2 = (x = -1, y = -1) (x = -1, y = 1)$

# Abstraction of concrete runs

The **abstraction**  $A(c, B)$  of a concrete **run**  $c = c[0] c[1] \dots c[n]$  with respect to predicates  $B$  is the **abstract run** over  $B$  defined as:

$$B(c[0]) B(c[1]) \dots B(c[n])$$

where  $B(c[k])$  is the **Boolean value** of predicates in  $B$  in state  $c[k]$ .

Example: given  $PE = \{(x \geq 0)^{\text{pos}}, (y = x)^{\text{eq}}\}$ :

$$c_1 = (x = 3, y = 0) (x = 3, y = 3) \quad A(c_1, PE) = (\text{pos}, \neg \text{eq}) (\text{pos}, \text{eq})$$

$$c_2 = (x = -1, y = -1) (x = -1, y = 1) \quad A(c_2, PE) = (\neg \text{pos}, \text{eq}) (\neg \text{pos}, \neg \text{eq})$$

# Abstraction of concrete programs

Thus, a Boolean program  $A$  over  $B$  also defines a set of **concrete runs**  $\mathcal{C}(A)$ : all runs of every program over **variables in**  $\mathcal{V}(B)$  whose abstractions with respect to  $B$  are in  $\mathcal{L}(A)$ :

$$\mathcal{C}(A) = \bigcup \{ \mathcal{L}(C) \mid A(C, B) \in \mathcal{L}(A) \}$$

The Boolean program over  $PE = \{(x \geq 0)^{\text{pos}}, (y = x)^{\text{eq}}\}$  on the left defines the concrete runs of the nondeterministic program over  $\mathcal{V}(PE) = \{x, y\}$  on the right:

```
var pos, eq: Boolean
if pos
  eq := true
else
  eq := ?
```

```
var x, y: Integer
if ?
  x, y := ?, ?
  assume x ≥ 0
  assume x = y
else
  x, y := ?, ?
  assume x < 0
```

# Predicate abstraction: definition

The **predicate abstraction**  $PA(C, B)$  of program  $C$  with predicates  $B$  is a Boolean program over  $P$  which is the **strongest over-approximation** of  $C$  with respect to  $B$ .

**over-approximation:**  $\mathcal{L}(C) \subseteq \mathcal{L}(PA(C, B))$

**strongest:** for every other Boolean program  $A$  over  $B$  that over-approximates  $C$ , it is  $\mathcal{L}(PA(C, B)) \subseteq \mathcal{L}(A)$

The predicate abstraction is the **most precise** approximation of  $C$  given the information that is captured by the **predicates**  $B$ .



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The predicate abstraction is the **most precise** approximation of  $C$  given the information that is captured by the **predicates**  $B$ .

The Boolean program on the left is the predicate abstraction of the program on the right with  $PE = \{(x \geq 0)^{\text{pos}}, (y = x)^{\text{eq}}\}$ .

```
var pos, eq: Boolean
```

```
if pos
```

```
  eq := true
```

```
else
```

```
  eq := ?
```

```
var x, y: Integer
```

```
if x ≥ 0
```

```
  y := x
```

```
else
```

```
  y := -3
```

# Predicate abstraction: informal overview

To **over approximate** a statement:

- compute what follows with **certainty** from the abstract state by computing the **weakest under-approximation** of the concrete state
- update the abstract state **accordingly**
- in all other cases, use **nondeterministic** values – corresponding to “don’t know” – to have an **over-approximation**

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- update the abstract state **accordingly**
- in all other cases, use **nondeterministic** values – corresponding to “don’t know” – to have an **over-approximation**

To **over approximate** a program: build the over approximation of each statement, while maintaining the same **program structure**.

```
var pos, eq: Boolean
if pos
  eq := true
else
  eq := ?
```

```
var x, y: Integer
if x ≥ 0
  y := x
else
  y := -3
```

# Under approximation of concrete predicates

$\mathcal{U}(b, B)$  is the **weakest** Boolean expression over predicates  $B$  which is **at least as strong** as  $b$  (that is, an **under approximation**):

**under-approximation:**  $\mathcal{U}(b, B) \implies b$

**weakest:** for every Boolean expression  $a$  over  $B$  such that  $a \implies b$ , it is  $a \implies \mathcal{U}(p, B)$

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**weakest:** for every Boolean expression  $a$  over  $B$  such that  $a \implies b$ , it is  $a \implies \mathcal{U}(p, B)$

To build  $\mathcal{U}(b, B)$ :

1. start from the **strongest** under-approximation:  $\perp$
2. **weaken** the under-approximation by adding **conjunctions** of **predicates** (negated or unnegated), as long as the expression still implies  $b$
3. stop after trying **all possible** conjunctions

## Under approximations: example

$$B = \left\{ (x = 1)^p, (x = 2)^q, (x \leq 3)^r \right\}$$

$$\mathcal{U}(x = 1, B) =$$

## Under approximations: example

$$B = \left\{ (x = 1)^p, (x = 2)^q, (x \leq 3)^r \right\}$$

$$\mathcal{U}(x = 1, B) = p$$

$$\mathcal{U}(x = 0, B) =$$

## Under approximations: example

$$B = \left\{ (x = 1)^p, (x = 2)^q, (x \leq 3)^r \right\}$$

$$\mathcal{U}(x = 1, B) = p$$

$$\mathcal{U}(x = 0, B) = \perp$$

$$\mathcal{U}(x \leq 2, B) =$$



## Under approximations: example

$$B = \left\{ (x = 1)^p, (x = 2)^q, (x \leq 3)^r \right\}$$

$$\mathcal{U}(x = 1, B) = p$$

$$\mathcal{U}(x = 0, B) = \perp$$

$$\mathcal{U}(x \leq 2, B) = p \vee q$$

$$\mathcal{U}(x \neq 0, B) =$$

## Under approximations: example

$$B = \left\{ (x = 1)^p, (x = 2)^q, (x \leq 3)^r \right\}$$

$$\mathcal{U}(x = 1, B) = p$$

$$\mathcal{U}(x = 0, B) = \perp$$

$$\mathcal{U}(x \leq 2, B) = p \vee q$$

$$\mathcal{U}(x \neq 0, B) = p \vee q \vee \neg r$$

These examples show that, in general,  $\mathcal{U}(\neg p, B) \neq \neg \mathcal{U}(p, B)$ .

## Under approximations: uniqueness

When some predicates in  $B$  **imply** some other predicates,  $\mathcal{U}(b, B)$  may not be **syntactically unique** – that is, there may be different **equivalent ways** of expressing it.

Example:

$$B = \left\{ (x < 2)^p, (x \leq 2)^q \right\}$$

$$\begin{aligned} \mathcal{U}(x \leq 3, B) &= p \vee q && \text{which is equivalent to} \\ &= q \end{aligned}$$

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Example:

$$B = \left\{ (x < 2)^p, (x \leq 2)^q \right\}$$

$$\begin{aligned} \mathcal{U}(x \leq 3, B) &= p \vee q && \text{which is equivalent to} \\ &= q \end{aligned}$$

Using any equivalent variant is correct. However, we will be able to **simplify** the construction of predicate abstraction when the predicate set  $B$  does not have any predicates that imply some other.

# Conditional Boolean expressions

To express program transformations involved in predicate abstraction, we will use this notation for **conditional Boolean expressions**:

$$(c:d:e)$$

which is a **shorthand** for:

```
if c then
  true
else (
  if d then
    false
  else
    e
)
```

if  $c$  then the expression evaluates to  $true$

if  $d$  (and  $\neg c$ ) then the expression evaluates to  $false$

otherwise  $(\neg c \wedge \neg d)$  the expression evaluates to  $e$

# Predicate abstraction: assignments

The predicate abstraction  $PA(s, B)$  of an **assignment**

$$s: \quad v := E$$

with predicates

$$B: \quad \{b_1, b_2, \dots, b_m\}$$

is the parallel conditional assignment

$$b_1, b_2, \dots, b_m := (c_1 : d_1 : ?), (c_2 : d_2 : ?), \dots, (c_m : d_m : ?)$$

where the  $k$ th component is defined as

$$c_k = \mathcal{U}(\overleftarrow{b}_k, B)$$

$$d_k = \mathcal{U}(\overleftarrow{\neg b}_k, B)$$

and  $\overleftarrow{b}$  is the **weakest precondition**  $\mathbf{wp}(v := E, b)$  of predicate  $b$  through the concrete assignment we are abstracting.

# Abstraction of assignments: example

$$PA(z := x, B) \quad \text{where} \quad B = \{(x > y)^p, (z \geq x)^q, (z \geq y)^r\}$$

The **predicate abstraction** is  $\mathcal{U}(x)$  is a shorthand for  $\mathcal{U}(x, B)$ :

$$p, q, r := (\mathcal{U}(\overleftarrow{p}) : \mathcal{U}(\overleftarrow{p}) : ?), (\mathcal{U}(\overleftarrow{q}) : \mathcal{U}(\overleftarrow{q}) : ?), (\mathcal{U}(\overleftarrow{r}) : \mathcal{U}(\overleftarrow{r}) : ?)$$

$x$	$\overleftarrow{x}$	$\mathcal{U}(\overleftarrow{x}, B)$
$p$	$x > y$	$p$
$\neg p$	$x \leq y$	$\neg p$
$q$	$x \geq x$	$\top$
$\neg q$	$x < x$	$\perp$
$r$	$x \geq y$	$p$
$\neg r$	$x < y$	$\perp$

$$p, q, r := (p : \neg p : ?), (\text{true} : \text{false} : ?), (p : \text{false} : ?)$$

# Abstraction of assignments: intuition

$PA(z := x, B)$  where  $B = \{(x > y)^p, (z \geq x)^q, (z \geq y)^r\}$  is:

$p, q, r := (p : \neg p : ?), (\text{true} : \text{false} : ?), (p : \text{false} : ?)$

- $p$  keeps its **current value** – since an assignment to  $z$  does not affect the value of  $x > y$
- $q$  becomes **true unconditionally** – since  $z = x$  implies that  $z \geq x$
- $r$  becomes **true if  $p$** , otherwise it is **nondeterministically** assigned – since  $z = x > y$  implies  $z \geq y$ , but if  $z = x \leq y$  we cannot conclude that  $z < y$



# Abstraction of assignments: intuition

$PA(z := x, B)$  where  $B = \{(x > y)^p, (z \geq x)^q, (z \geq y)^r\}$  is:

$p, q, r := (p : \neg p : ?), (\text{true} : \text{false} : ?), (p : \text{false} : ?)$

$p$  keeps its **current value** – since an assignment to  $z$  does not affect the value of  $x > y$

$q$  becomes **true unconditionally** – since  $z = x$  implies that  $z \geq x$

$r$  becomes **true if  $p$** , otherwise it is **nondeterministically** assigned – since  $z = x > y$  implies  $z \geq y$ , but if  $z = x \leq y$  we cannot conclude that  $z < y$

The conditional assignment can be rewritten into the equivalent, simpler form:

$q, r := \text{true}, (p : \text{false} : ?)$

# Abstraction of assignments: intuition

In general, the *k*th predicate  $b_k \in B$  in  $PA(v := E, B)$  is updated to:

$$b_k := (\mathcal{U}(\overleftarrow{b_k}, B) : \mathcal{U}(\overleftarrow{\neg b_k}, B) : ?)$$

- $\overleftarrow{b_k}$  is the **exact pre-condition** that implies  $b_k$  after executing  $v := E$
- $\mathcal{U}(\overleftarrow{b_k}, B)$  is its **best sound** approximation expressible using predicates in  $B$
- if  $\mathcal{U}(\overleftarrow{b_k}, B)$  is true, we are sure that  **$b_k$  is true** after executing the assignment
- a similar reasoning shows that if  $\mathcal{U}(\overleftarrow{\neg b_k}, B)$  is true, we are sure that  **$b_k$  is false** after executing the assignment
- in all **other cases**, we over-approximate by assigning a **nondeterministic value** to  $b_k$

# Abstraction of assignments: example

$PA(y := x, B)$  where  $B = \{(x = 1)^p, (y = 1)^q, (x > y)^r\}$

The **predicate abstraction** is  $\mathcal{U}(x)$  is a shorthand for  $\mathcal{U}(x, B)$ :

$p, q, r := (\mathcal{U}(\overleftarrow{p}) : \mathcal{U}(\overleftarrow{\neg p}) : ?), (\mathcal{U}(\overleftarrow{q}) : \mathcal{U}(\overleftarrow{\neg q}) : ?), (\mathcal{U}(\overleftarrow{r}) : \mathcal{U}(\overleftarrow{\neg r}) : ?)$

$x$	$\overleftarrow{x}$	$\mathcal{U}(\overleftarrow{x}, B)$
$p$	$x = 1$	$p$
$\neg p$	$x \neq 1$	$\neg p$
$q$	$x = 1$	$p$
$\neg q$	$x \neq 1$	$\neg p$
$r$	$x > x$	$\perp$
$\neg r$	$x \leq x$	$\top$

$p, q, r := (p : \neg p : ?), (p : \neg p : ?), (\text{false} : \text{true} : ?)$

equivalent to  $q, r := p, \text{false}$ .

# Predicate abstraction: assertions

The predicate abstraction  $PA(s, B)$  of an **assert**

$s: \text{assert } E$

with predicates

$B: \{b_1, b_2, \dots, b_m\}$

is the assertion

$\text{assert } \mathcal{U}(E, B)$

# Predicate abstraction: assertions

The predicate abstraction  $PA(s, B)$  of an **assert**

$s: \text{ assert } E$

with predicates

$B: \{b_1, b_2, \dots, b_m\}$

is the assertion

$\text{assert } \mathcal{U}(E, B)$

This is an **over-approximation** because its correctness **implies** the original program's correctness, but there may be **spurious** failures:

- if  $\mathcal{U}(E, B)$  holds in every abstract run, then  $E$  holds in every concrete run
- if  $\mathcal{U}(E, B)$  fails in some abstract run, then  $E$  may or may not hold in every concrete run

# Predicate abstraction: assumptions

The predicate abstraction  $PA(s, B)$  of an **assumption**

$s: \text{assume } E$

with predicates

$B: \{b_1, b_2, \dots, b_m\}$

is an assumption followed by a parallel conditional assignment

$\text{assume } \neg \mathcal{U}(\neg E, B)$

$b_1, b_2, \dots, b_m := (c_1 : d_1 : ?), (c_2 : d_2 : ?), \dots, (c_m : d_m : ?)$

where the  $k$ th component is defined as

$$c_k = \mathcal{U}(\overleftarrow{b_k}, B)$$

$$d_k = \mathcal{U}(\overleftarrow{\neg b_k}, B)$$

and  $\overleftarrow{b}$  is the **weakest precondition**  $\text{wp}(\text{assume } E, b)$  of predicate  $b$  through the concrete assumption we are abstracting.

# Abstraction of assumptions: example

$PA(\text{assume } x \leq 2, B)$  where  $B = \{(x = 1)^p, (x = 2)^q, (x \leq 3)^r\}$

The **predicate abstraction** is  $\mathcal{U}(x)$  is a shorthand for  $\mathcal{U}(x, B)$ :

$\text{assume } \neg \mathcal{U}(\neg(x \leq 2))$

$p, q, r := (\mathcal{U}(\overleftarrow{p}) : \mathcal{U}(\overleftarrow{\neg p}) : ?), (\mathcal{U}(\overleftarrow{q}) : \mathcal{U}(\overleftarrow{\neg q}) : ?), (\mathcal{U}(\overleftarrow{r}) : \mathcal{U}(\overleftarrow{\neg r}) : ?)$

$x$	$\overleftarrow{x}$	$\mathcal{U}(\overleftarrow{x}, B)$
	$x \leq 2$	$p \vee q$
	$x > 2$	$\neg r$
$p$	$x \leq 2 \implies x = 1$	$p \vee \neg r$
$\neg p$	$x \leq 2 \implies x \neq 1$	$\neg p \vee \neg r$
$q$	$x \leq 2 \implies x = 2$	$q \vee \neg r$
$\neg q$	$x \leq 2 \implies x \neq 2$	$\neg q \vee \neg r$
$r$	$x \leq 2 \implies x \leq 3$	$\top$
$\neg r$	$x \leq 2 \implies x > 3$	$\neg r$

$\text{assume } r$

followed by the parallel conditional assignment

# Abstraction of assumptions: intuition

$PA(\text{assume } x \leq 2, B)$  where  $B = \{(x = 1)^p, (x = 2)^q, (x \leq 3)^r\}$  is:

$\text{assume } r$

$p, q, r := (p \vee \neg r : \neg p \vee \neg r : ?), (q \vee \neg r : \neg q \vee \neg r : ?), \text{true}$

The double negation  $\text{assume } \neg \mathcal{U}(\neg E, B)$  builds, by duality, an **over**-approximation on the state **from** the **under**-approximation  $\mathcal{U}$ :

- $\text{assume } \mathcal{U}(x \leq 2, B) = \text{assume } p \vee q$ : if we can prove the abstract program correct under the **stronger assumption**  $x = 1 \vee x = 2$ , the concrete program may still misbehave under the **assumption**  $x \leq 2 \wedge x \neq 1 \wedge x \neq 2 = x < 1$
- $\text{assume } \neg \mathcal{U}(x > 2, B) = \text{assume } r$ : whenever we can prove the abstract program correct under the **weaker assumption**  $x \leq 3$ , the concrete program is also correct under the **stronger assumption**  $x \leq 2$

The parallel conditional assignment **propagates** the assumed state to **all predicates** correctly.



# Predicate abstraction of assumptions: simplification

$PA(\text{assume } E, B)$  where  $B = \{b_1, b_2, \dots, b_m\}$  is  
 $\text{assume } \neg \mathcal{U}(\neg E, B) \quad \dots, b_k, \dots := \dots, (\mathcal{U}(\overleftarrow{b_k}) : \mathcal{U}(\overleftarrow{\neg b_k}) : ?), \dots$

If  $E \implies b_k$  and  $E \implies b_k$  are **not** unconditionally **valid**:

$$\mathcal{U}(\overleftarrow{b_k}) = \neg \mathcal{U}(\neg E) \implies b_k \quad \mathcal{U}(\overleftarrow{\neg b_k}) = \neg \mathcal{U}(\neg E) \implies \neg b_k$$

which, under the assumption  $\text{assume } \neg \mathcal{U}(\neg E)$ , simplify to

$$\mathcal{U}(\overleftarrow{b_k}) = b_k \quad \mathcal{U}(\overleftarrow{\neg b_k}) = \neg b_k$$

Thus  $b_k := (b_k : \neg b_k : ?)$  which has no effect on  $b_k$

For example:

$$\begin{aligned} \mathcal{U}(\overleftarrow{b_k}) &= \mathcal{U}(E \implies b_k) = \mathcal{U}(\neg E \vee b_k) && \text{definition of implication} \\ &= \mathcal{U}(\neg E) \vee \mathcal{U}(b_k) && \text{provable from the definition of } \mathcal{U} \\ &= \neg \mathcal{U}(\neg E) \implies \mathcal{U}(b_k) && \text{definition of implication} \end{aligned}$$

# Predicate abstraction of assumptions: simplified rule

The predicate abstraction  $PA(s, B)$  of an assumption

$$s: \text{assume } E$$

with predicates

$$B: \{b_1, b_2, \dots, b_m\}$$

is an assumption

$$\text{assume } \neg \mathcal{U}(\neg E, B) \wedge \bigwedge_{k \in J^+} b_k \wedge \bigwedge_{k \in J^-} \neg b_k$$

where  $J^+$  and  $J^-$  are defined as:

$$J^+ = \{k \mid E \implies p_k \text{ is valid}\} \quad J^- = \{k \mid E \implies \neg p_k \text{ is valid}\}$$

# Predicate abstraction: conditionals

The predicate abstraction  $PA(s, B)$  of a **conditional**

$$s: \text{ if } (C) \ T \ \text{else} \ E$$

with predicates

$$B: \quad \{b_1, b_2, \dots, b_m\}$$

is a **nondeterministic conditional**:

$$\text{ if } (?) \ PA(\{\text{assume } C ; T\}, B) \ \text{else} \ PA(\{\text{assume } \neg C ; E\}, B)$$

that recursively applies predicate abstraction to the **then** and **else** branches under the right conditions.

# Predicate abstraction: loops

The predicate abstraction  $PA(s, B)$  of a **loop**

$s: \text{while } (C) L$

with predicates

$B: \{b_1, b_2, \dots, b_m\}$

is a **nondeterministic loop**:

$\left\{ \text{while } (?) \text{ } PA(\{\text{assume } C ; L\}, B) \right\} ; PA(\text{assume } \neg C, B)$

that recursively applies predicate abstraction to the **body**  $L$  and to the **exit** condition of the loop.

# Predicate abstraction: procedure definitions

The predicate abstraction  $PA(s, B)$  of a **procedure definition**

$s$ : **procedure**  $proc$ : **require**  $P$  **ensure**  $Q$   $R$

with predicates

$B$ :  $\{b_1, b_2, \dots, b_m\}$

is a **program**:

$\text{var } b_1, b_2, \dots, b_m$ : **Boolean** ;  $PA \left( \begin{array}{c} \text{assume } P \\ R \\ \text{assert } Q \end{array}, B \right)$

# Predicate abstraction: procedure definitions

The predicate abstraction  $PA(s, B)$  of a **procedure definition**

$s$ :    **procedure**  $proc$ : **require**  $P$  **ensure**  $Q$   $R$

with predicates

$B$ :     $\{b_1, b_2, \dots, b_m\}$

is a **program**:

$\text{var } b_1, b_2, \dots, b_m$ : **Boolean** ;  $PA \left( \begin{array}{c} \text{assume } P \\ R \\ \text{assert } Q \end{array}, B \right)$

In our examples we always choose  $B$  so that precondition  $P$  and postcondition  $Q$  can be **expressed exactly**.

# Predicate abstraction: complete program example

$$B = \{(x > y)^p, (\max \geq x)^q, (\max \geq y)^r\}$$

**procedure** max

(x, y: **Integer**): (max: **Integer**)

**ensure** max  $\geq$  x  $\wedge$  max  $\geq$  y

**if** x > y

max := x

**else**

max := y

# Predicate abstraction: complete program example

$$B = \{(x > y)^p, (\max \geq x)^q, (\max \geq y)^r\}$$

**procedure** max

(x, y: **Integer**): (max: **Integer**)

**ensure** max  $\geq$  x  $\wedge$  max  $\geq$  y

**if** x > y

max := x

**else**

max := y

**var** p, q, r: **Boolean**

**if** x > y

max := x

**else**

max := y

**assert** max  $\geq$  x  $\wedge$  max  $\geq$  y



# Predicate abstraction: complete program example

$$B = \{(x > y)^p, (\max \geq x)^q, (\max \geq y)^r\}$$

**procedure** max

(x, y: **Integer**): (max: **Integer**)

**ensure** max  $\geq$  x  $\wedge$  max  $\geq$  y

**if** x > y

max := x

**else**

max := y

**var** p, q, r: **Boolean**

**if** ?

assume x > y

max := x

**else**

assume x  $\leq$  y

max := y

**assert** max  $\geq$  x  $\wedge$  max  $\geq$  y

# Predicate abstraction: complete program example

$$B = \{(x > y)^p, (\max \geq x)^q, (\max \geq y)^r\}$$

**procedure** max

(x, y: **Integer**): (max: **Integer**)

**ensure** max  $\geq$  x  $\wedge$  max  $\geq$  y

**if** x > y

max := x

**else**

max := y

**var** p, q, r: **Boolean**

**if** ?

assume p

q, r := true, (p:false:?)

**else**

assume  $\neg$  p

max := y

**assert** q  $\wedge$  r

# Predicate abstraction: complete program example

$$B = \{(x > y)^p, (\max \geq x)^q, (\max \geq y)^r\}$$

**procedure** max

(x, y: **Integer**): (max: **Integer**)

**ensure** max  $\geq$  x  $\wedge$  max  $\geq$  y

**if** x > y

max := x

**else**

max := y

**var** p, q, r: **Boolean**

**if** ?

assume p

q, r := **true**, (p:false:?)

**else**

assume  $\neg$ p

q, r := ( $\neg$ p:false:?), **true**

**assert** q  $\wedge$  r

# Predicate abstraction: complete program example

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**if** ?

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**else**

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q, r := true, true

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# Predicate abstraction: complete program example

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**procedure** max

(x, y: **Integer**): (max: **Integer**)

**ensure** max  $\geq$  x  $\wedge$  max  $\geq$  y

**if** x > y

max := x

**else**

max := y

**procedure** PA\_max

(p, q, r: **Boolean**):

**ensure** q  $\wedge$  r

**if** p

q, r := true, true

**else**

q, r := true, true

# Model checking Boolean programs

Every Boolean program  $P = \ell_1, \dots, \ell_n$  over variables  $B = \{b_1, \dots, b_m\}$  defines a **finite-state transition** system  $\langle S, I, R, B, L \rangle$ , which can be **model checked**.

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$S$ :  $\mathbb{B}^k \times \{1, \dots, n+1, \text{error}\}$

The **states** are the set of all **Boolean values** of variables in  $B$ , plus a **program counter** indicating the next statement in  $P$  to be executed (or  $n+1$  if the program has terminated). There is also a distinct **error** location.

$I$ :  $\mathbb{B}^k \times \{1\}$

The **initial** states are all those pointing to the first statement  $\ell_1$  in program  $P$ .

$L$ :  $L((p_1, \dots, p_m, k)) = k$

The **labeling** function just projects the program location of each state.

# Model checking Boolean programs

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$R$ :  $(p_1, \dots, p_m, k) R (q_1, \dots, q_m, h)$  iff one of the following holds:

- $\ell_k$ :  $p_x := E$ ,  $e = E[b_1, \dots, b_m \mapsto p_1, \dots, p_m]$ ,  $q_1, \dots, q_m = p_1, \dots, p_m[p_x \mapsto e]$ , and  $h = k + 1$  is the next statement
- $\ell_k$ :  $, e = E[b_1, \dots, b_m \mapsto p_1, \dots, p_m]$ ,  $q_1, \dots, q_m = p_1, \dots, p_m[p_x \mapsto e]$ , and  $h = k + 1$  is the next statement
- $\ell_k$ : **assert**  $E$ ,  $E[b_1, \dots, b_m \mapsto p_1, \dots, p_m]$  evaluates to true in the pre-state,  $q_1, \dots, q_m = p_1, \dots, p_m[p_x \mapsto e]$ , and  $h = k + 1$  is the next statement
- $\ell_k$ : **assert**  $E$ ,  $E[b_1, \dots, b_m \mapsto p_1, \dots, p_m]$  evaluates to false in the pre-state, and  $h = \text{error}$  the error location
- $\ell_k$ : **assume**  $E$ ,  $E[b_1, \dots, b_m \mapsto q_1, \dots, q_m]$  evaluates to true in the post-state, and  $h = k + 1$  is the next statement
- $\ell_k$ : **if**  $C \{T\}^t$  **else**  $\{E\}^e$ , condition  $C[b_1, \dots, b_m \mapsto p_1, \dots, p_m]$  evaluates to true in the pre-state,  $q_1, \dots, q_m = p_1, \dots, p_m$ , and  $h = t$  is the then branch's location
- $\ell_k$ : **if**  $C \{T\}^t$  **else**  $\{E\}^e$ , condition  $C[b_1, \dots, b_m \mapsto p_1, \dots, p_m]$  evaluates to false in the pre-state,  $q_1, \dots, q_m = p_1, \dots, p_m$ , and  $h = e$  is the else branch's location
- $\ell_k$ : **while**  $C \{L\}^i$ ;  $\ell_o$ , condition  $C[b_1, \dots, b_m \mapsto p_1, \dots, p_m]$  evaluates to true in the pre-state,  $q_1, \dots, q_m = p_1, \dots, p_m$ , and  $h = i$  is the loop body's first statement
- $\ell_k$ : **while**  $C \{L\}^i$ ;  $\ell_o$ , condition  $C[b_1, \dots, b_m \mapsto p_1, \dots, p_m]$  evaluates to false in the pre-state,  $q_1, \dots, q_m = p_1, \dots, p_m$ , and  $h = o$  is the first statement after the loop



# Model checking Boolean programs

Every Boolean program  $P = \ell_1, \dots, \ell_n$  over variables  $B = \{b_1, \dots, b_m\}$  defines a **finite-state transition** system  $\langle S, I, R, B, L \rangle$ , which can be **model checked**.

With this model, a Boolean program is **correct** iff there are **no assertion violations**, iff model checking **verifies**  $P \models \Box(\neg \text{error})$ .

# Software model checking

---

## Spurious counterexample detection

# Verification of Boolean programs

The **predicate abstraction**  $PA(C, B)$  of program  $C$  with predicates  $B$  is a Boolean program over  $P$  which is the **strongest over-approximation** of  $C$  with respect to  $B$ .

$PA(C, B)$  ✓

$C$  is correct

$PA(C, B)$  ✗

model checking may return:

- a **real** counterexample (true positive) if  $C$  is **not** correct
- a **spurious** counterexample (false positive) if  $C$  is **correct**

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- a **spurious** counterexample (false positive) if  $C$  is **correct**

We present a technique to detect if an **abstract counterexample** is **spurious** (not executable) or **real** (executable).

# Run concretization

Every statement in a **concrete** program  $C$  correspond to **one** (possibly compound) **statement** in its predicate **abstraction**  $PA(C, B)$ . Thus, given an **abstract run** we can always build a sequence of **concrete statements** of the concrete program the abstract program emulated.

ABSTRACT RUN	CONCRETE RUN
$a_0$	
$A_1$	$C_1$
$a_1$	
$A_2$	$C_2$
$\vdots$	$\vdots$
$A_N$	$C_N$
$a_N$	

- $a_0 a_1 \dots a_N$  is the abstract run's sequence of **abstract states**
- $A_1 \dots A_N$  are the run's **abstract statements** executed
- $C_1 \dots C_N$  are the corresponding **concrete statements**

# Run concretization: example

**procedure** max

(x, y: **Integer**): (max: **Integer**)  $QR = \{(\max \geq x)^q, (\max \geq y)^r\}$

**ensure** max  $\geq$  x  $\wedge$  max  $\geq$  y

**if** x > y

max := x

**else**

max := y

**procedure** QR\_max

(q, r: **Boolean**):

**ensure** q  $\wedge$  r

**if** ?

q, r := **true**, ?

**else**

q, r := ?, **true**

ABSTRACT RUN  $A^1$

CONCRETE RUN  $C^1$

$q \wedge \neg r$

**if** ?

$q \wedge \neg r$

q, r := **true**, ?

$q \wedge \neg r$

**if** x > y

max := x

# Run concretization: example

**procedure** max

(x, y: **Integer**): (max: **Integer**)  $QR = \{(\max \geq x)^q, (\max \geq y)^r\}$

**ensure** max  $\geq$  x  $\wedge$  max  $\geq$  y

**if** x > y

max := x

**else**

max := y

**procedure** QR\_max

(q, r: **Boolean**):

**ensure** q  $\wedge$  r

**if** ?

q, r := **true**, ?

**else**

q, r := ?, **true**

ABSTRACT RUN  $A^2$

CONCRETE RUN  $C^2$

$q \wedge \neg r$

**if** ?

$q \wedge \neg r$

q, r := ?, **true**

$q \wedge r$

**if** x > y

max := y

# Run feasibility constraints

To determine if a concrete run is **feasible** (executable):

- transform every concrete **branch** statement  $C$  (**if**  $E$ , **while**  $E$ , **assume**  $E$ , and **assert**  $E$ ) into a **feasibility constraint**  $C'$ : **assert**  $E$
- compute the **weakest precondition** of  $a_n$  through each concrete statement  $C_k$

The concrete run  $C$  obtained from an abstract run  $A$  is **feasible** iff, for every  $k = 0, \dots, N$ ,  $a_k \wedge c_k$  is **satisfiable**.

ABSTRACT RUN	CONCRETE RUN
$a_0$	$c_0 = \mathbf{wp}(C_1, c_1)$
$A_1$	$C'_1$
$a_1$	$c_1 = \mathbf{wp}(C_2, c_2)$
$A_2$	$C'_2$
$\vdots$	$\vdots$
$a_{N-1}$	$c_{N-1} = \mathbf{wp}(C_N, c_N)$
$A_N$	$C'_N$
$a_N$	$c_N = a_N$

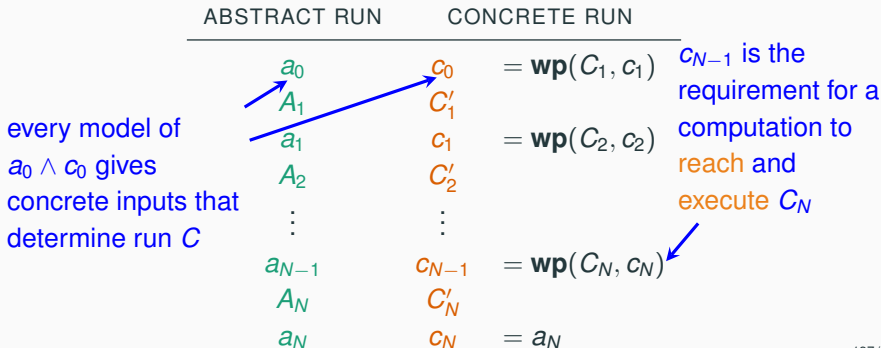


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The concrete run  $C$  obtained from an abstract run  $A$  is **feasible** iff, for every  $k = 0, \dots, N$ ,  $a_k \wedge c_k$  is **satisfiable**.



# Spurious run detection

An abstract run  $A$  is **spurious** iff  
there exists  $k = 0, \dots, N$  such that  $a_k \wedge c_k$  is **unsatisfiable**.

Spurious run detection can be **automated**:

- computing the weakest precondition along a run does not involve loops, and hence can be automated
- an SMT solver can decide satisfiability of  $a_k \wedge c_k$  (if it uses a decidable fragment)

ABSTRACT RUN	CONCRETE RUN	
$a_0$	$c_0$	$= \mathbf{wp}(C_1, c_1)$
$A_1$	$C'_1$	
$a_1$	$c_1$	$= \mathbf{wp}(C_2, c_2)$
$A_2$	$C'_2$	
$\vdots$	$\vdots$	
$a_{N-1}$	$c_{N-1}$	$= \mathbf{wp}(C_N, c_N)$
$A_N$	$C'_N$	
$a_N$	$c_N$	$= a_N$

## Spurious run detection: example

An abstract run  $A$  is **spurious** iff  
there exists  $k = 0, \dots, N$  such that  $a_k \wedge c_k$  is **unsatisfiable**.

$$QR = \{(\max \geq x)^q, (\max \geq y)^r\}$$

## Spurious run detection: example

An abstract run  $A$  is **spurious** iff  
there exists  $k = 0, \dots, N$  such that  $a_k \wedge c_k$  is **unsatisfiable**.

$$QR = \{(\max \geq x)^q, (\max \geq y)^r\}$$

ABSTRACT RUN  $A^1$

CONCRETE RUN  $C^1$

$q \wedge \neg r$

$x > y \wedge x \geq x \wedge x < y$

**if** ?

**assert**  $x > y$

$q \wedge \neg r$

$x \geq x \wedge x < y$

$q, r := \text{true}, ?$

$\max := x$

$q \wedge \neg r$

$\max \geq x \wedge \max < y$

$x > y \wedge x \geq x \wedge x < y$  is **unsatisfiable**,  
and hence the abstract run is **spurious**.

There is **no concrete** run that is abstracted by the abstract run  $A^1$ .

## Spurious run detection: example

An abstract run  $A$  is **spurious** iff  
there exists  $k = 0, \dots, N$  such that  $a_k \wedge c_k$  is **unsatisfiable**.

$$QR = \{(\max \geq x)^q, (\max \geq y)^r\}$$

ABSTRACT RUN $A^2$	CONCRETE RUN $C^2$
$q \wedge \neg r$	$x \leq y \wedge y \geq x \wedge y \geq y$
<b>if</b> ?	<b>assert</b> $x \leq y$
$q \wedge \neg r$	$y \geq x \wedge y \geq y$
$q, r := ?, \text{true}$	$\max := y$
$q \wedge r$	$\max \geq x \wedge \max \geq y$

The abstract run is **feasible**. Every **input**  $x, y$  that satisfies  $x \leq y \equiv x \leq y \wedge y \geq x \wedge y \geq y$  determines a concrete run that is abstracted by the abstract run  $A^2$ .

# Concrete counterexample detection: example

Is procedure `negpow` **correct** with respect to its **specification**?

```
procedure negpow
  (x, y: Integer): (pow: Integer)
require x < 0 ∧ y > 0
ensure pow > 0
  pow := 1
  while y > 0
    pow := pow * x
    y := y - 1
```

# Concrete counterexample detection: example

Is procedure `negpow` **correct** with respect to its **specification**?

Let's analyze it using a **predicate abstraction** over predicates:

$$NP = \{(x < 0)^p, (y > 0)^q, (pow > 0)^r\}$$

**procedure** `negpow`

(`x`, `y`: **Integer**): (`pow`: **Integer**)

**require** `x < 0`  $\wedge$  `y > 0`

**ensure** `pow > 0`

`pow` := 1

**while** `y > 0`

`pow` := `pow` \* `x`

`y` := `y` - 1

**procedure** `NP_negpow`

(`p`, `q`, `r`: **Boolean**):

**require** `p`  $\wedge$  `q`

**ensure** `r`

`r` := **true**

**while** `q`

`r` := (**false**:`p`  $\wedge$  `r`:?)

`q` := ?

# Concrete counterexample detection: example

Model checking NP\_negpow finds the abstract counterexample  $A^e$ :

ABSTRACT RUN  $A^e$

CONCRETE RUN  $C^e$

---

```
 $p \wedge q \wedge r$   
r := true  
 $p \wedge q \wedge r$   
while q  
   $p \wedge q \wedge r$   
r := (false:p  $\wedge$  r: ?)  
   $p \wedge q \wedge \neg r$   
  q := ?  
   $p \wedge \neg q \wedge \neg r$   
  while q  
     $p \wedge \neg q \wedge \neg r$ 
```

```
pow := 1  
  
assert y > 0  
  
pow := pow * x  
  
y := y - 1  
  
assert y ≤ 0
```

$$NP = \{(x < 0)^q, (y > 0)^q, (\text{pow} > 0)^r\}$$



# Concrete counterexample detection: example

Model checking NP\_negpow finds the abstract counterexample  $A^e$ :

ABSTRACT RUN  $A^e$

CONCRETE RUN  $C^e$

$p \wedge q \wedge r$	$y > 0 \wedge x < 0 \wedge y \leq 1 \wedge x \leq 0$
$r := \text{true}$	$\text{pow} := 1$
$p \wedge q \wedge r$	$y > 0 \wedge x < 0 \wedge y \leq 1 \wedge \text{pow} * x \leq 0$
<b>while</b> $q$	<b>assert</b> $y > 0$
$p \wedge q \wedge r$	$x < 0 \wedge y \leq 1 \wedge \text{pow} * x \leq 0$
$r := (\text{false}; p \wedge r; ?)$	$\text{pow} := \text{pow} * x$
$p \wedge q \wedge \neg r$	$x < 0 \wedge y \leq 1 \wedge \text{pow} \leq 0$
$q := ?$	$y := y - 1$
$p \wedge \neg q \wedge \neg r$	$x < 0 \wedge y \leq 0 \wedge \text{pow} \leq 0$
<b>while</b> $q$	<b>assert</b> $y \leq 0$
$p \wedge \neg q \wedge \neg r$	$x < 0 \wedge y \leq 0 \wedge \text{pow} \leq 0$

$$NP = \{(x < 0)^q, (y > 0)^q, (\text{pow} > 0)^r\}$$

# Concrete counterexample detection: example

Model checking NP\_negpow finds the abstract counterexample  $A^e$ :

ABSTRACT RUN  $A^e$

CONCRETE RUN  $C^e$

$p \wedge q \wedge r$	$y > 0 \wedge x < 0 \wedge y \leq 1 \wedge x \leq 0$
$r := \text{true}$	$\text{pow} := 1$
$p \wedge q \wedge r$	$y > 0 \wedge x < 0 \wedge y \leq 1 \wedge \text{pow} * x \leq 0$
<b>while</b> $q$	<b>assert</b> $y > 0$
$p \wedge q \wedge r$	$x < 0 \wedge y \leq 1 \wedge \text{pow} * x \leq 0$
$r := (\text{false}; p \wedge r; ?)$	$\text{pow} := \text{pow} * x$
$p \wedge q \wedge \neg r$	$x < 0 \wedge y \leq 1 \wedge \text{pow} \leq 0$
$q := ?$	$y := y - 1$
$p \wedge \neg q \wedge \neg r$	$x < 0 \wedge y \leq 0 \wedge \text{pow} \leq 0$
<b>while</b> $q$	<b>assert</b> $y \leq 0$
$p \wedge \neg q \wedge \neg r$	$x < 0 \wedge y \leq 0 \wedge \text{pow} \leq 0$

$$NP = \{(x < 0)^q, (y > 0)^q, (\text{pow} > 0)^r\}$$

The abstract counterexample is **feasible**. It reveals a real **error**, triggered for every input that satisfies  $x < 0 \wedge y = 1$ .

# Software model checking

---

## Predicate discovery

A **spurious counterexample** indicates that the predicate abstraction  $PA(C, B)$  is **too imprecise** to conclusively **verify**  $C$ .

In these case, we want to **refine** the abstraction – getting a **more precise** one.

A **spurious counterexample** indicates that the predicate abstraction  $PA(C, B)$  is **too imprecise** to conclusively **verify**  $C$ .

In these case, we want to **refine** the abstraction – getting a **more precise** one.

**Predicate discovery** is a technique to find a **new** predicate  $b \notin B$  such that  $PA(C, B \cup \{b\})$  **refines**  $PA(C, B)$ .

Predicate discovery starts from a **spurious counterexample**  $A$  of  $PA(C, B)$  and finds  $b$  such that  $A$  is **not feasible** in  $PA(C, B \cup \{b\})$ .

# Syntactic predicate discovery

We look for a predicate  $b$  that:

- holds in the **concrete** run
- is **not traced** by any predicates in  $B$
- makes the abstract state **unsatisfiable**

ABSTRACT RUN	CONCRETE RUN	
$a_0$	$c_0$	$= \mathbf{wp}(C_1, c_1)$
$A_1$	$C'_1$	
$a_1$	$c_1$	$= \mathbf{wp}(C_2, c_2)$
$A_2$	$C'_2$	
$\vdots$	$\vdots$	
$a_{N-1}$	$c_{N-1}$	$= \mathbf{wp}(C_N, c_N)$
$A_N$	$C'_N$	
$a_N$	$c_N$	$= a_N$

# Syntactic predicate discovery

We look for a predicate  $b$  that:

- holds in the **concrete** run
- is **not traced** by any predicates in  $B$
- makes the abstract state **unsatisfiable**

$\mathcal{P}(F)$  denotes the **set** of all elementary **predicates** in a formula  $F$ .

ABSTRACT RUN

CONCRETE RUN

---

$a_0$	$c_0$	$= \mathbf{wp}(C_1, c_1)$
$A_1$	$C'_1$	
$a_1$	$c_1$	$= \mathbf{wp}(C_2, c_2)$
$A_2$	$C'_2$	
$\vdots$	$\vdots$	
$a_{N-1}$	$c_{N-1}$	$= \mathbf{wp}(C_N, c_N)$
$A_N$	$C'_N$	
$a_N$	$c_N$	$= a_N$

# Syntactic predicate discovery

$\mathcal{P}(F)$  denotes the **set** of all elementary **predicates** in a formula  $F$ .

For any  $k$  such that  $c_k \wedge a_k$  is **unsatisfiable**,  
 $B' = \mathcal{P}(c_k) \setminus \mathcal{P}(a_k)$  gives **new predicates**. Any  $b \in B'$  that is not  
equivalent to  $\mathcal{U}(b, B)$  **refines** the abstraction.

ABSTRACT RUN

CONCRETE RUN

---

$a_0$	$c_0$	$= \mathbf{wp}(C_1, c_1)$
$A_1$	$C'_1$	
$a_1$	$c_1$	$= \mathbf{wp}(C_2, c_2)$
$A_2$	$C'_2$	
$\vdots$	$\vdots$	
$a_{N-1}$	$c_{N-1}$	$= \mathbf{wp}(C_N, c_N)$
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 $B' = \mathcal{P}(c_k) \setminus \mathcal{P}(a_k)$  gives **new predicates**. Any  $b \in B'$  that is not  
 equivalent to  $\mathcal{U}(b, B)$  **refines** the abstraction.

ABSTRACT RUN	CONCRETE RUN	NEW PREDICATES
$a_0$	$c_0 = \mathbf{wp}(C_1, c_1)$	$\mathcal{P}(c_0) \setminus \mathcal{P}(a_0)$
$A_1$	$C'_1$	
$a_1$	$c_1 = \mathbf{wp}(C_2, c_2)$	$\mathcal{P}(c_1) \setminus \mathcal{P}(a_1)$
$A_2$	$C'_2$	
$\vdots$	$\vdots$	
$a_{N-1}$	$c_{N-1} = \mathbf{wp}(C_N, c_N)$	$\mathcal{P}(c_{N-1}) \setminus \mathcal{P}(a_{N-1})$
$A_N$	$C'_N$	
$a_N$	$c_N = a_N$	$\mathcal{P}(c_N) \setminus \mathcal{P}(a_N)$

## Predicate discovery: example

$$QR = \{(\max \geq x)^q, (\max \geq y)^r\}$$

ABSTRACT RUN $A^1$	CONCRETE RUN $C^1$
$q \wedge \neg r$	$x > y \wedge x \geq x \wedge x < y$
<b>if</b> ?	<b>assert</b> $x > y$
$q \wedge \neg r$	$x \geq x \wedge x < y$
$q, r := \text{true}, ?$	$\max := x$
$q \wedge \neg r$	$\max \geq x \wedge \max < y$
$x > y \wedge x \geq x \wedge x < y$ is <b>unsatisfiable</b> , and hence the abstract run is <b>spurious</b> .	

# Predicate discovery: example

$$QR = \{(\max \geq x)^q, (\max \geq y)^r\}$$

ABSTRACT RUN $A^1$	CONCRETE RUN $C^1$	$c_k \wedge a_k$ SAT?
$q \wedge \neg r$	$x > y \wedge x \geq x \wedge x < y$	✗
<b>if</b> ?	<b>assert</b> $x > y$	
$q \wedge \neg r$	$x \geq x \wedge x < y$	✓
$q, r := \text{true}, ?$	$\max := x$	
$q \wedge \neg r$	$\max \geq x \wedge \max < y$	✓

# Predicate discovery: example

$$QR = \{(\max \geq x)^q, (\max \geq y)^r\}$$

ABSTRACT RUN $A^1$	CONCRETE RUN $C^1$	$c_k \wedge a_k$ SAT?	$\mathcal{P}(c_k) \setminus \mathcal{P}(a_k)$
$q \wedge \neg r$	$x > y \wedge x \geq x \wedge x < y$	✗	$\{x > y\}$
if ?	assert $x > y$		
$q \wedge \neg r$	$x \geq x \wedge x < y$	✓	$\{x < y\}$
$q, r := \text{true}, ?$	$\max := x$		
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We use  $x > y$  for refinement since it contradicts the abstract state, is a **new predicate**, and

$$\mathcal{U}(x > y, QR) = \neg q \wedge r \not\Leftarrow x > y.$$

As we already know, the refined abstraction  $PA(\max, QR \cup \{x > y\})$  passes verification without spurious counterexamples.

# Software model checking

---

## Software model checking tools

# A short demo of CPAchecker

CPAchecker is a software model checker for C code, which includes numerous analysis techniques based on automatic abstraction – including specialized variants of the CEGAR model checking algorithm we have seen.

Let's run some of the examples we have seen before using CPAchecker.

# A short demo of CPAchecker: examples

By default, CPAchecker analyzes programs looking for paths that reach a special program label `ERROR`. We can encode assertions as reachability as usual.

```
int max(int x, int y)
{
    int max;
    if (x > y)
        max = x;
    else
        max = y;
    if (max >= x && max >= y)
        return max;
    else
        ERROR: return max;
}
```

```
int negpow(int x, int y)
{
    int pow;
    if (!(x < 0 && y > 0))
        return;
    pow = 1;
    while (y > 0) {
        pow *= x;
        y--;
    }
    if (pow > 0)
        return pow;
    else
        ERROR: return pow;
}
```



# A short demo of CPAchecker: analysis and results

The script `cpa` provided with the Docker image runs CPAchecker using the analysis `predicateAnalysis-PredAbsRefiner-SBE`, which is a form of CEGAR analysis.

```
> cpa max.c
```

The tool produces a detailed output and error report in subdirectory `output`. The Docker image includes the main HTML reports for each example:

1. `max.c` (correctly verified): `max.ok.html`
2. `max.c` (error after assigning `max = y` in both branches):  
`max.error.html`
3. `negpow.c` (real error): `negpow.error.html`
4. `maxa.c` (analysis breaks down – even if we inline the function call): `maxa.error.html`

# A short demo of CPAchecker: unsupported features

- Function `maxa` computes the maximum `max` of input array `a`
- Function `is_max` checks that `max` is indeed an upper bound on the values in `a`

The checking loop, which is executed a variable number of times depending on the size `n` of `a`, cannot be **abstracted effectively** using CPAchecker's analysis.

```
int maxa(int a[], int n)
```

```
{
    int max;
    if (n <= 0)
        return;
    max = a[0];
    for (int k = 1; k < n; k++) {
        if (a[k] > max)
            max = a[k];
    }
    is_max(a, n, max);
}
```

```
int is_max(int a[], int n, int m)
```

```
{
    for (int k = 0; k < n; k++) {
        if (a[k] > m)
            goto ERROR;
    }
    return 1;
ERROR:
    return 0;
}
```

## A short demo of CPAchecker: unsupported features

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The checking loop, which is executed a variable number of times depending on the size `n` of `a`, cannot be **abstracted effectively** using CPAchecker's analysis.

CPAchecker can **scale** to much larger programs than this small example, but only if the **properties** we want to check are **simple** and can be effectively captured with a reasonable number of **predicates**.

# Is software model checking model checking?

Model checking proper is only **one ingredient** of CEGAR software model checking:

**PREDICATE ABSTRACTION** uses **static analysis** (expressible as abstract interpretation)

**SPURIOUS COUNTEREXAMPLE** detection uses **deductive verification**

**PREDICATE DISCOVERY** is based on **symbolic execution** of traces

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*The term “software model checker” is probably a misnomer [...] We retain the term solely to reflect historical development.*

*Jhala & Majumdar, 2009*



# Software model checking tools

Microsoft Research's **SLAM** (Ball, Rajamani, et al.) was the first complete CEGAR model checker.

UC Berkeley's **BLAST** (Jhala, Majumdar, Henzinger, et al.) was the first open-source CEGAR model checker, which introduced several improvements to scalability.

Software model checkers that are currently maintained include:

**CPAchecker** is an extensible configurable **verifier** for C code, with a modular architecture that includes CEGAR-style verification algorithms

**CBMC** is a verifier for C/C++ using bounded model checking, which is part of a toolset that includes tools for Java (JBMC) and for predicate abstraction (SatAbs)

**ESBMC** is a bounded model checker for C/C++ supporting the analysis of multi-threaded programs

**JPF** (Java PathFinder) is a model checker for Java source code supporting concurrency and abstraction

# Real-time model checking

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# Real-time systems

**Real-time** model checking applies model checking techniques to finite-state models of **real-time** systems.



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A **real-time specification** is a specification that includes **quantitative (exact) timing** information.

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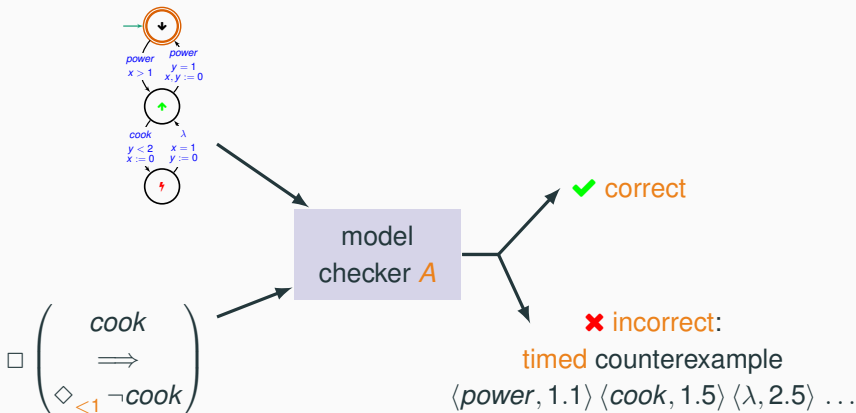
A **real-time system** is a system whose **correctness** is defined by a real-time specification.

Many **embedded computing** devices are real-time systems:

- a car's braking system (ABS)
- a multi-media player (or smart TV)
- Boeing 737 MAX's Maneuvering Characteristics Augmentation System
- ...

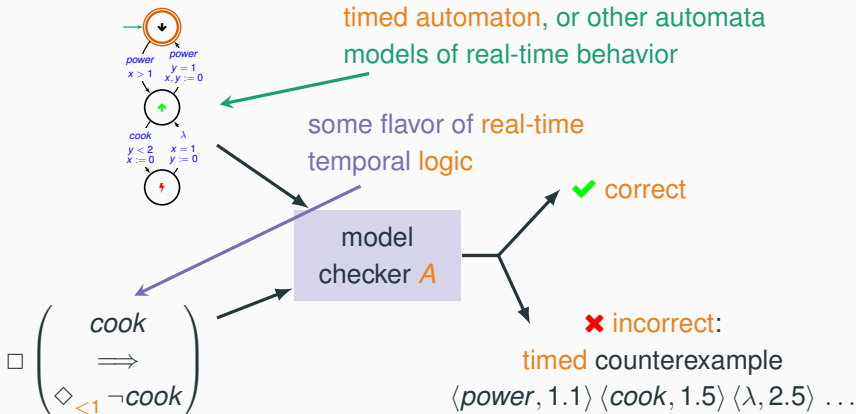
# Model-checking real-time systems

**Real-time** model checking verifies **timed automata** models  
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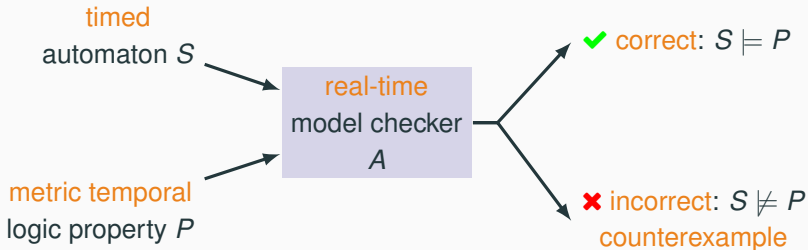


# Real-time model checking of timed automata and MTL

Real-time model checking problem: given

- $S$ : a **timed automaton** (TA)
- $P$ : a **metric temporal logic** (MTL) property

determine if **every run of  $S$  satisfies  $P$** , or  
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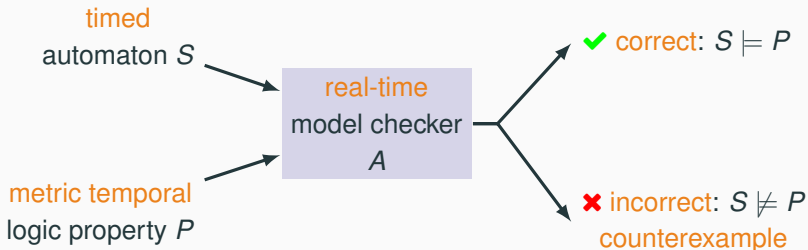


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We first describe **syntax** and **semantics** of TAs and MTL, and then describe an **algorithm** for real-time model checking.

# Dense vs. discrete time domain

A fundamental choice in real-time modeling is the nature of the **time domain**.

DENSE/CONTINUOUS	DISCRETE
time domain: $\mathbb{R}_{\geq 0}$	time domain: $\mathbb{N}$
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This lecture focuses on the more challenging problem of using a **dense time domain**  $\mathbb{R}_{\geq 0}$ .

Later, we will also mention the complexity of the simpler real-time model checking using the discrete time domain  $\mathbb{N}$ .

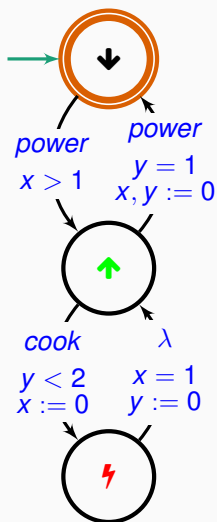
# Real-time model checking

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## Timed automata

# Timed automata: example

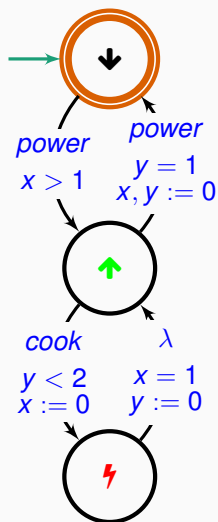
We model the timed behavior of a **microwave oven** using a TA, which is a finite state automaton equipped with **clocks**.




The oven may be powered on  $\uparrow$  or off  $\downarrow$ .

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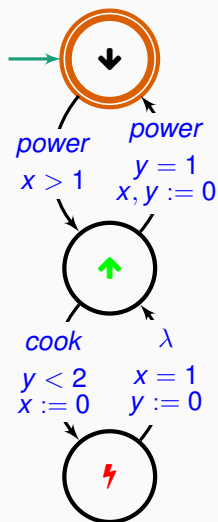


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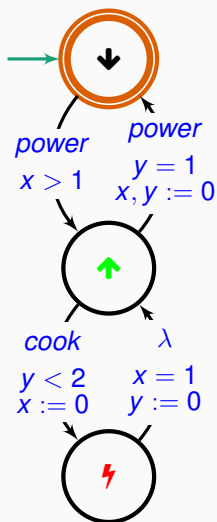
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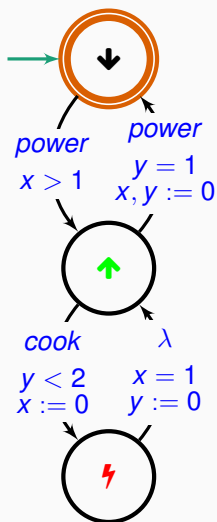
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The oven may be powered on  $\uparrow$  or off  $\downarrow$ .

When the oven is on, pressing the cook button *cook* starts cooking ⚡.

Each cooking round lasts exactly one **time unit**.

When cooking ends, you can turn off the oven exactly after another **time unit**.

After the oven is turned off, you have to one **time unit** before turning it on again.



# Timed automata: syntax

A **nondeterministic timed automaton (TA)**  $A$   
is a tuple  $\langle \Sigma, S, I, F, C, E \rangle$ :

- $\Sigma$ : finite nonempty input **alphabet**
- $S$ : finite nonempty set of **locations** (discrete states)
- $I, F \subseteq S$ : **initial** and **final (accepting)** states
- $C$ : finite set of **clocks**
- $E \subseteq S \times \Sigma \times \chi \times C \times S$ : finite set of **transitions (edges)**

An edge  $E \ni e = (s_1, \sigma, c, \rho, s_2)$  indicates a transition:

- **from** location  $s_1$  **to** location  $s_2$
- reading **input** symbol  $\sigma \in \Sigma$
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We commonly represent TAs with a **graph** similar to FSAs.

## Clock constraints: syntax

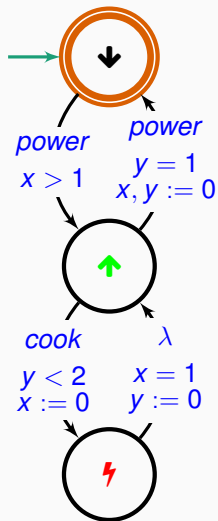
A **clock constraint**  $c$  is Boolean combination of predicates comparing a **clock** to an **integer** constant:

$$\chi \ni c ::= x \bowtie n \mid \neg c \mid c_1 \wedge c_2 \mid c_1 \vee c_2$$

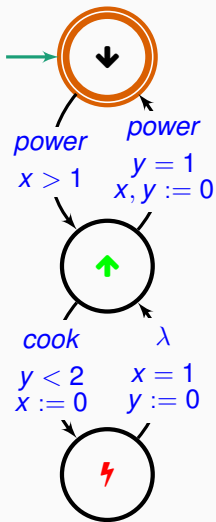
where  $\bowtie$  can be any of  $<, \leq, =, \neq, \geq, >$   
and  $n \in \mathbb{N}$  can be any nonnegative integer (natural number).

Note that clock constraints compare clocks to **integers** even though the clock themselves range over the **nonnegative reals**. Later we will comment on the impact of this restriction.

## TA syntax: example

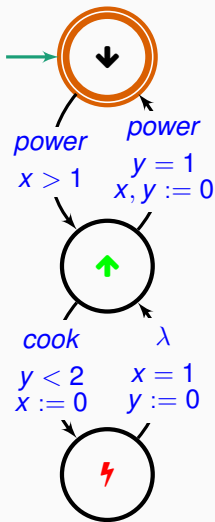


## TA syntax: example



alphabet  $\Sigma = \{power, cook, \lambda\}$

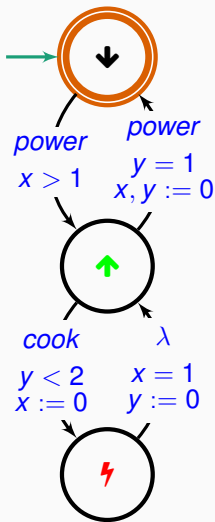
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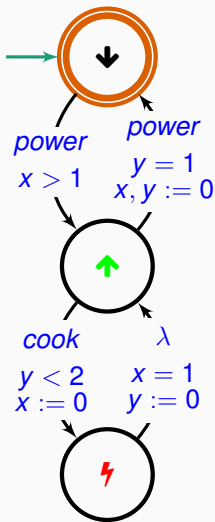


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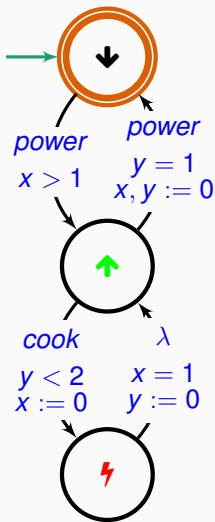
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initial and final location  $I = F = \{\blacktriangledown\}$

clocks  $C = \{x, y\}$

transitions  $E$ :

- $(\blacktriangledown, power, x > 1, \{\}, \blacktriangleup)$
- $(\blacktriangleup, power, y < 2, \{x\}, \text{⚡})$
- $(\text{⚡}, \lambda, x = 1, \{y\}, \blacktriangleup)$
- $(\blacktriangleup, power, y = 1, \{x, y\}, \blacktriangledown)$

# Timed words

Let  $A = \langle \Sigma, S, I, F, C, E \rangle$  be a TA.

An input **timed word** is an input sequence of any (finite) length with **timestamps** (time values ranging over the **nonnegative reals**):

$$\begin{aligned} w &= w[1] w[2] \dots w[n] \in (\Sigma \times \mathbb{R}_{\geq 0})^* \\ &= \langle \sigma[1], t[1] \rangle \langle \sigma[2], t[2] \rangle \dots \langle \sigma[n], t[n] \rangle \end{aligned}$$

such that the sequence  $t[1] t[2] \dots t[n]$  of timestamps is **increasing**.

A timed word element  $\langle \sigma[k], t[k] \rangle$  denotes **input**  $\sigma[k]$  read at **time**  $t[k]$ .

As in the **untimed** case, the **empty** word  $\epsilon$  has **zero length** ( $n = |\epsilon| = 0$ ).

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Examples:

$w_1: \langle power, 3.0 \rangle$

$w_2: \langle power, 1.3 \rangle \langle cook, 1.7 \rangle \langle \lambda, 2.7 \rangle \langle power, 3.7 \rangle$

# Timed automata: runs

A **run** of  $A$  over  $w$  is a sequence of **states** (location + clock values):

$$\begin{aligned} r &= r[0] r[1] \dots r[n] \in (S \times \mathbb{R}_{\geq 0}^{|C|})^* \\ &= \langle s[0], x_1[0], \dots, x_{|C|}[0] \rangle \dots \langle s[n], x_1[n], \dots, x_{|C|}[n] \rangle \quad \text{that:} \end{aligned}$$

- **starts** from an **initial** location with all clocks reset to **zero**:  
 $s[0] \in I$  and, for all  $1 \leq k \leq |C|$ ,  $x_k[0] = 0$
- **follows**  $A$ 's **transitions**: for all  $0 \leq i < n$ , the  $i$ th transition  $\langle s[i], x_1[i], \dots, x_{|C|}[i] \rangle \longrightarrow \langle s[i+1], x_1[i+1], \dots, x_{|C|}[i+1] \rangle$ 
  - follows an **edge**:  $(s[i], \sigma[i+1], c, \rho, s[i+1])$
  - the updated **clock** values  $x_1[i] + \Delta_i, \dots, x_{|C|}[i] + \Delta_i$  satisfy clock **constraint**  $c$ , where  $\Delta_i = t[i+1] - t[i]$  is the **time spent** in  $s[i]$
  - all clocks  $x_r \in \rho$  are **reset**:  $x_r[i+1] = 0$
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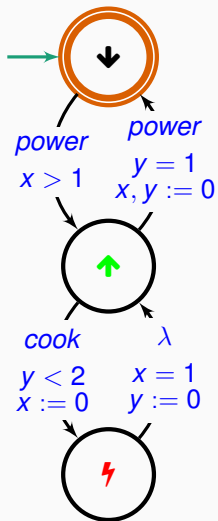
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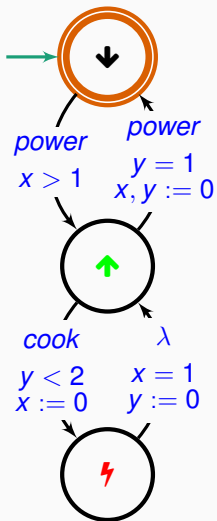
A run  $r$  is **accepting** if it **ends** in a **final** location:  $s[n] \in F$ .

In this case we say that  $A$  **accepts**  $w$ .

## TA runs: example

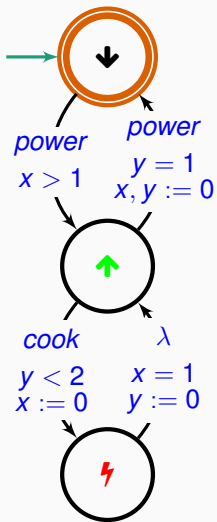


# TA runs: example



Run  $r_1 = \langle \blacktriangledown, x = 0, y = 0 \rangle \langle \blacktriangledown, x = 3.0, y = 3.0 \rangle$   
over  $w_1 = \langle \text{power}, 3.0 \rangle$

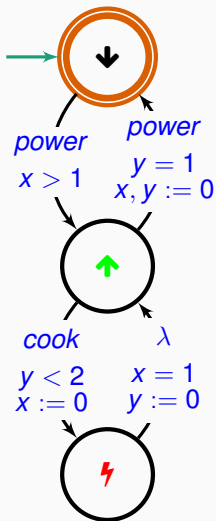
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over  $w_1 = \langle \text{power}, 3.0 \rangle$   
is **not** accepting.



# TA runs: example



Run  $r_1 = \langle \blacktriangledown, x = 0, y = 0 \rangle \langle \blacktriangledup, x = 3.0, y = 3.0 \rangle$   
 over  $w_1 = \langle power, 3.0 \rangle$   
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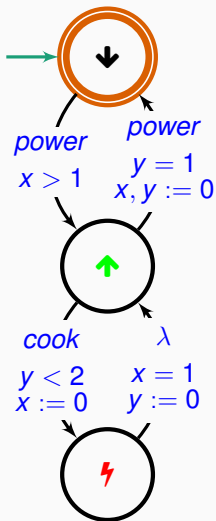
Run

$r_2 = \langle \blacktriangledown, x = 0, y = 0 \rangle \langle \blacktriangledup, x = 1.3, y = 1.3 \rangle$   
 $\langle \text{⚡}, x = 0, y = 1.7 \rangle \langle \blacktriangledup, x = 1, y = 0 \rangle \langle \blacktriangledown, x = 0, y = 0 \rangle$

over

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# Timed automata: semantics

The **language** of a timed automaton  $A = \langle \Sigma, S, I, F, C, E \rangle$  is the **set of timed words** that  $A$  **accepts**:

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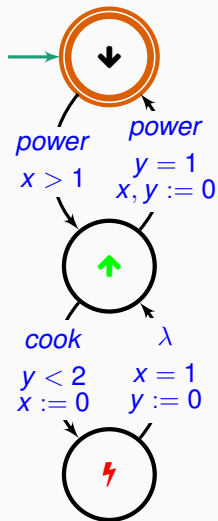
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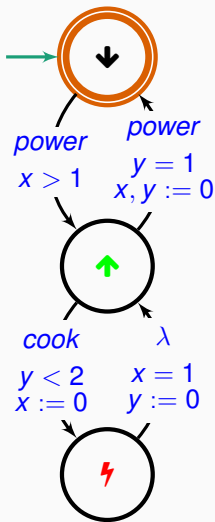
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Note that even though TAs have a **finite** number of **locations**, their **extended states** (including clock values) may be **infinite**. Therefore, checking emptiness is not as straightforward as for FSAs.

## TA semantics: example

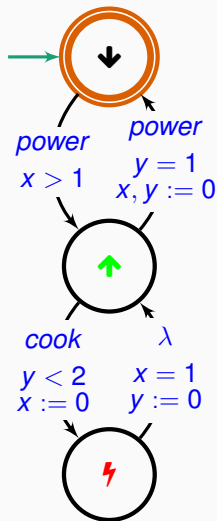


## TA semantics: example



The language of the microwave timed automaton is **not empty**.

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Words in the automaton's language include:

- $\epsilon$
- $w_2$
- ...



# Real-time model checking

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Formulas of propositional **metric temporal logic (MTL)** are defined as:

$$\begin{aligned} F ::= & p \mid \neg F \mid F_1 \wedge F_2 \mid F_1 \vee F_2 \mid F_1 \implies F_2 && \text{(propositional connectives)} \\ & \mid \mathsf{X} F \mid \square_J F \mid \diamond_J F \mid F_1 \mathsf{U}_J F_2 && \text{(metric temporal operators)} \end{aligned}$$

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**Intervals** can be:

**open:**  $(3, 7)$  denotes all reals  $t$  such that  $3 < t < 7$

**closed:**  $[2, 5]$  denotes all reals  $t$  such that  $2 \leq t \leq 5$

**half-open:**  $[0, 3)$  denotes all reals  $t$  such that  $0 \leq t < 3$

**unbounded:**  $[1, \infty)$  denotes all reals  $t$  such that  $1 \leq t$

**pointwise:**  $[3, 3]$  denotes only the real  $t = 3$

We often use **abbreviations** for special intervals:

- $\geq 3$  for  $[3, \infty)$ ,  $< 4$  for  $[0, 4)$ ,  $= 7$  for  $[7, 7]$
- **omitting**  $J$  stands for the whole time domain  $[0, \infty)$

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# Metric temporal logic: satisfaction relation

A **timed** word  $w \in (\Pi \times \mathbb{R}_{\geq 0})^*$

$$w = w[1] w[2] \dots w[n] = \langle \sigma[1], t[1] \rangle \langle \sigma[2], t[2] \rangle \dots \langle \sigma[n], t[n] \rangle$$

**satisfies** MTL formula  $F$  at **position**  $1 \leq k \leq n$ , written  $w, k \models F$ , iff:

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$w, k \models p$	iff	$p = \sigma[k]$
$w, k \models \neg F$	iff	$w, k \not\models F$
$w, k \models F_1 \wedge F_2$	iff	$w, k \models F_1$ and $w, k \models F_2$
$w, k \models F_1 \vee F_2$	iff	$w, k \models F_1$ or $w, k \models F_2$
$w, k \models F_1 \implies F_2$	iff	$w, k \models \neg F_1 \vee F_2$

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$$w, k \models \chi F \quad \text{iff} \quad k < n \text{ and } w, k+1 \models F$$

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$$w, k \models F_1 \cup_J F_2 \quad \text{iff} \quad \text{for some } k \leq h \leq n:$$

$$t[h] - t[k] \in J \text{ and } w, h \models F_2,$$

$$\text{and, for all } k \leq j < h: w, j \models F_1$$

# MTL satisfaction: example

$$w = \begin{matrix} \langle power, 1.3 \rangle & \langle cook, 1.7 \rangle & \langle \lambda, 2.7 \rangle & \langle power, 3.7 \rangle \\ w[1] & w[2] & w[3] & w[4] \end{matrix}$$

$$w, 1 \models \Diamond_{<2} power$$

$$w, 1 \not\models \Box_{<2} power$$

$$w, 2 \models cook \wedge \Diamond_{=2} power$$

$$w, 1 \models power \cup_{<1} cook$$

$$w, 3 \models \Diamond_{<2} power$$

$$w, 1 \models \Box_{(4,5)} power$$

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A **timed** word  $w$  **satisfies** an MTL formula  $F$  if it satisfies it **initially**:

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The **language** of a metric temporal logic formula  $F$  is the **set** of **timed words** that **satisfy**  $F$ :

$$\mathcal{L}(F) = \{w \in (\Pi \times \mathbb{R}_{\geq 0})^* \mid w \models F\}$$



# Real-time model checking

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## Real-time model-checking algorithm

# Real-time model checking as emptiness checking

Real-time model checking: given

- $A$ : a **timed automaton** with alphabet  $\Sigma$
- $P$ : a **metric temporal logic** property over propositions in  $\Sigma$

$\mathcal{L}(A) \cap \mathcal{L}(\neg P)$  is **empty**

every timed word **accepted**  
by  $A$  **satisfies**  $P$

✓  $A \models P$

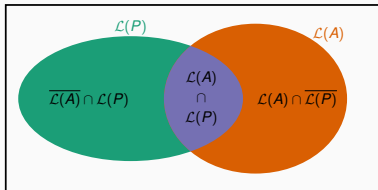
$\mathcal{L}(A) \cap \mathcal{L}(\neg P)$  is **not empty**

some timed word **accepted** by  $A$   
does **not** satisfy  $P$

✗  $A \not\models P$

every timed word in  $\mathcal{L}(A) \cap \mathcal{L}(\neg P)$   
is a **counterexample**

$(\Sigma \times \mathbb{R}_{\geq 0})^*$



# Real-time model-checking algorithm

To apply automata-based model checking to real-time verification, we need to extend its **three algorithms** to work with **timed automata** and **metric temporal logic**:

**MONITOR:** given a **metric** temporal logic formula  $P$  build a **timed** automaton  $\mathcal{A}(P)$  such that  $\mathcal{L}(\mathcal{A}(P)) = \mathcal{L}(P)$

**INTERSECTION:** given timed automata  $A$  and  $B$ , build a **timed** automaton  $A \times B$  such that  $\mathcal{L}(A \times B) = \mathcal{L}(A) \cap \mathcal{L}(B)$

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Things are more difficult with real-time:

- monitor: there exist MTL formulas that **cannot be encoded** as **timed automata**
- intersection works similarly as in the untimed case
- emptiness is still decidable but requires a significantly more **complex algorithm**

# Monitors: from MTL to timed automata

There exist MTL formulas  $P$  such that it impossible to build a monitor of  $P$ : there exist no TA  $A$  that accepts precisely the timed words that satisfy  $P$ .

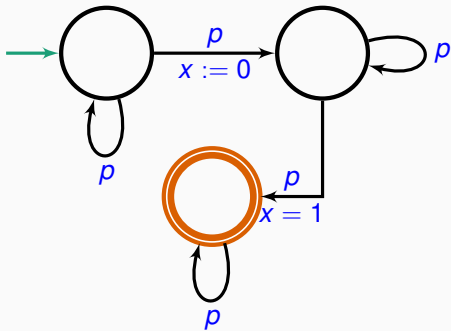
This is not a nonstarter for real-time model checking, since we can still build a monitor for interesting classes of MTL properties — but not all of them!

# Expressiveness gap between MTL and timed automata

$O = \Box (p \implies \Box_{=1} \neg p)$       no two  $p$ 's are separated by one time unit

$\overline{O} = \Diamond (p \wedge \Diamond_{=1} p) = \neg O$       there exist two  $p$ 's separated by one time unit

A timed automaton equivalent to  $\overline{O}$ :

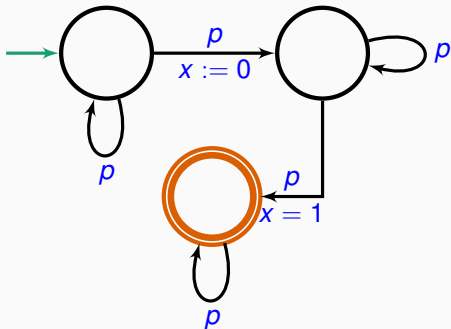


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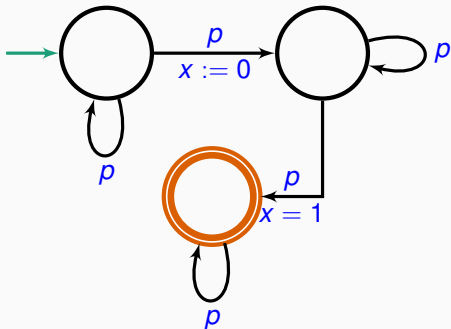
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This also shows that timed automata are not closed under complement.



## Monitors: bounded response

$$P_1 = \Box (p \implies \Diamond_{<2} q) \quad p \text{ is followed by } q \text{ within 2}$$

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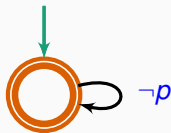
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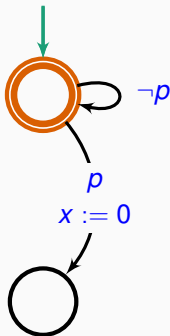
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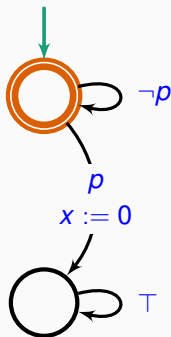
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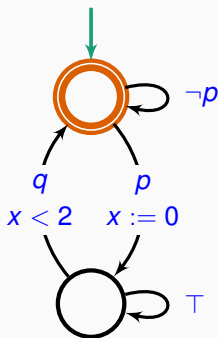
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- a  $q$  must eventually occur while  $x < 2$  ("covering" all  $p$ 's that occurred since the last reset of  $x$ )

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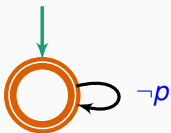
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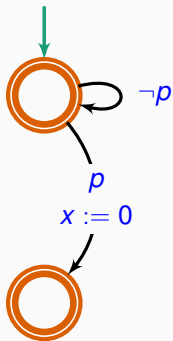
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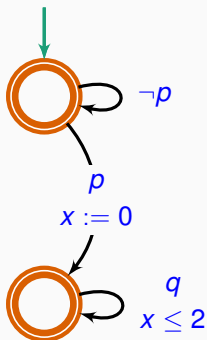
$$P_2 = \Box \left( p \implies \Box_{(0,2]} q \right) \quad \text{only } q \text{ occur after } p \text{ within 2}$$



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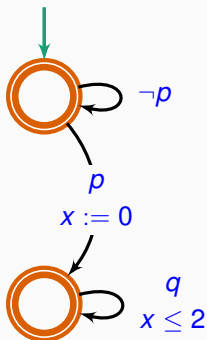
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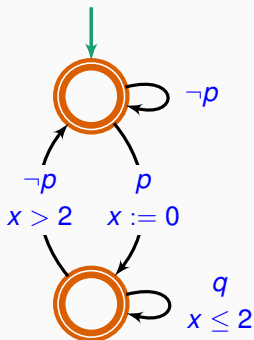
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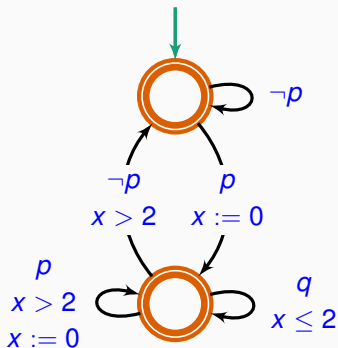
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- after 2 time units, we restart
- unless we read  $p$  again, in which case we reset  $x$



## Intersection: running automata in parallel

A timed automaton  $C$  that accepts the intersection of two TAs  $A$  and  $B$ 's languages runs  $A$  and  $B$  in parallel:

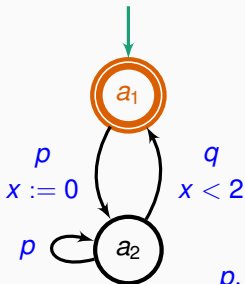
- starts from any combination of initial states of  $A$  and  $B$
- transitions only when both  $A$  and  $B$  have a transition for the current input, with the same constraints and clock resets
- accepts when both  $A$  and  $B$  accept

# Intersection: running automata in parallel

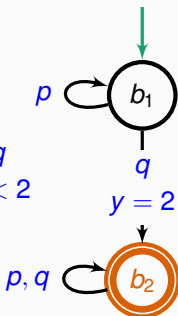
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automaton  $A$ :



automaton  $B$ :

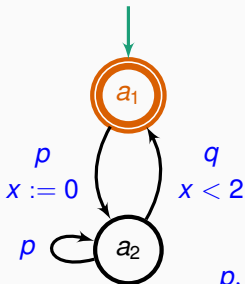


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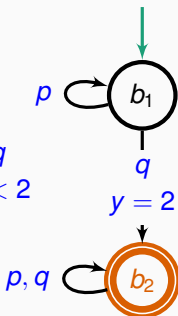
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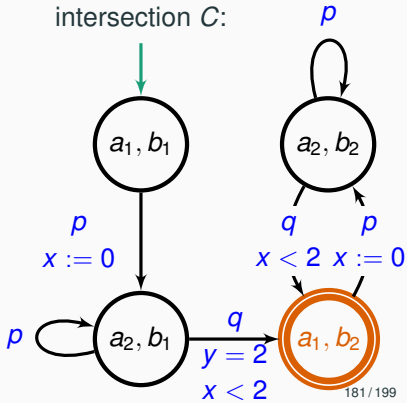
automaton A:



automaton B:



intersection C:



# Product timed automaton construction

Given TAs  $A = \langle \Sigma, S_A, I_A, F_A, C_A, E_A \rangle$  and  $B = \langle \Sigma, S_B, I_B, F_B, C_B, E_B \rangle$ , the **product automaton**  $A \times B = \langle \Sigma, S, I, F, C, E \rangle$  is defined as:

$$S = S_A \times S_B$$

$$I = \{(a, b) \mid a \in I_A \text{ and } b \in I_B\}$$

$$F = \{(a, b) \mid a \in F_A \text{ and } b \in F_B\}$$

$$C = C_A \cup C_B \quad \text{assuming } C_A \cap C_B = \emptyset$$

$$E \ni ((a, b), \sigma, c_A \wedge c_B, \rho_A \cup \rho_B, (a_2, b_2)) \text{ iff}$$

$$(a, \sigma, c_A, \rho_A, a_2) \in E_A \quad \text{and} \quad (b, \sigma, c_B, \rho_B, b_2) \in E_B$$

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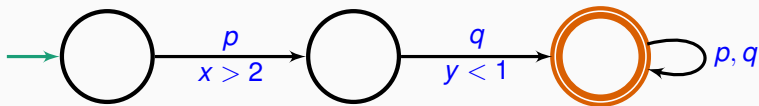
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The **language** of the product automaton is the **intersection** of the intersected automata's languages:

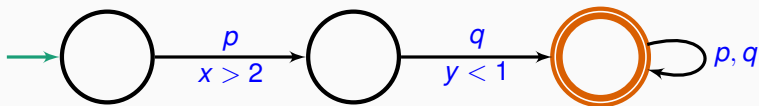
$$\mathcal{L}(A \times B) = \mathcal{L}(A) \cap \mathcal{L}(B)$$

## Timed automata emptiness: overview

Emptiness of TAs is not just reachability: **clocks** introduce additional **constraints** that may be unsatisfiable. For example, the following TA is **empty** even if an accepting state is reachable on the graph.



# Timed automata emptiness: overview



We present Alur and Dill's **region automaton** algorithm to determine the **emptiness** of TAs.

- a TA's run may traverse infinitely many **extended states** because clocks can take any value
- however, we may **approximate** the precise values of clocks by their **region**: an equivalence relation that only looks at the **relative order** of clocks compared to **integer constants** they are actually compared to in the automaton
- since the number of **regions** is always **finite**, we can build a **finite-state automaton** – the **region automaton** – whose language is empty **iff** the TA's language is empty

# Equivalence between clock values

Consider a TA  $A$  with  $m$  clocks  $x_1, \dots, x_m$  such that  $M_m$  is the **largest constant** used in  $A$ 's clock constraints involving  $x_m$ .

A **clock evaluation** is any tuple  $(t_1, \dots, t_m) \in \mathbb{R}_{\geq 0}^m$  of **clock values**, which is part of  $A$ 's extended state.

Two clock evaluations  $t = (t_1, \dots, t_m)$  and  $u = (u_1, \dots, u_m)$  are **equivalent** (written  $t \sim u$ ) iff **all** the following conditions hold:

1. corresponding **clocks** have the same **integer** value:  
for all  $1 \leq k \leq m$ :  $\text{int}(t_k) = \text{int}(u_k)$  or  $t_k, u_k > M_k$
2. corresponding clock **pairs** have the same **order of fractional** values:  
for all  $1 \leq j \leq k \leq m$  such that  $t_j \leq M_j$  and  $t_k \leq M_k$ :  
 $\text{frac}(t_j) \leq \text{frac}(t_k)$  iff  $\text{frac}(u_j) \leq \text{frac}(u_k)$
3. corresponding **clocks** agree on having **integer or fractional** value:  
for all  $1 \leq k \leq m$  such that  $t_k \leq M_j$ :  $\text{frac}(t_k) = 0$  iff  $\text{frac}(u_k) = 0$

Here  $\text{int}(x)$  is the **integer part** of  $x$ ;  $\text{frac}(x)$  is its **fractional part**.

For example:  $\text{int}(3.1412) = 3$ ,  $\text{frac}(3.1412) = 0.1412$ .



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$t_1, t_2$	$u_1, u_2$	$M_1$	$M_2$	$t \sim u?$
0.3, 2.2	0.4, 2.7	3	3	✓
0.3, 2.0	0.5, 4.9	1	5	✗
0.3, 2.0	0.5, 4.9	1	1	✓
1.0, 1.6	1.0, 1.8	2	2	✓

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Intuition behind the **equivalence**: A only compares clocks to **integer constants**, and all clocks proceed at the same rate; hence, we only need to keep track of

- condition 1: whether a clock is less than, equal to (condition 3), or greater than an **integer constant**
- condition 2: which clock will **reach** an integer value **next**
- all conditions: when clock  $x_k$  is greater than  $M_k$  its exact value does not matter until a **reset**

# Regions

Clock regions are the equivalence classes induced by the equivalence relation  $\sim$  between clock evaluations.

$\|t\|$  denotes the region clock evaluation  $t$  belongs to.

# Regions

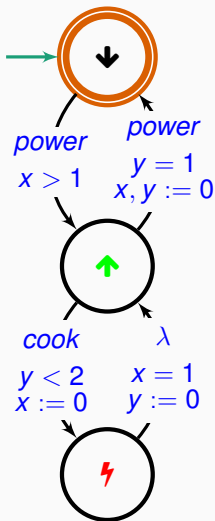
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In practice, we keep track of which integers each clock is between, and the relative order between clocks (including if they are equal).

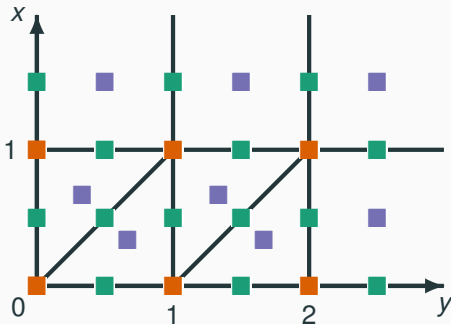
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0.3, 2.0	0.5, 4.9	1	5	✗	$0 < x_1 < 1$ $x_2 = 2$	$0 < x_1 < 1$ $4 < x_2 < 5$
0.3, 2.0	0.5, 4.9	1	1	✓	$0 < x_1 < 1$ $x_2 > 1$	$0 < x_1 < 1$ $x_2 > 1$
1.0, 1.6	1.0, 1.8	2	2	✓	$x_1 = 1$ $1 < x_2 < 2$	$x_1 = 1$ $1 < x_2 < 2$

# Clock regions: example



The microwave TA determines 28 possible regions:

- 8 open regions 
- 14 open line segments 
- 6 corner points 



# Time successors

The **time successors**  $\tilde{r}$  of a region  $r$  are all regions (including  $r$ ) that can be reached from  $r$  by letting **time pass**.

$M_1$	$M_2$	$r$	$\tilde{r}$
1	1	$0 < x_1 < x_2 < 1$	$0 < x_1 < x_2 < 1$ $0 < x_1 < 1 = x_2$ $x_1 = 1 < x_2$ $x_1, x_2 > 1$
2	2	$0 < x_1 < 1, x_2 = 2$	$0 < x_1 < 1, x_2 = 2$ $x_1 = 1, x_2 > 2$ $1 < x_1 < 2 < x_2$ $x_1 = 2 < x_2, x_2 > 2$
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# Region automaton

The **region automaton**  $\|A\|$  of timed automaton  $A = \langle \Sigma, S, I, F, C, E \rangle$  is the finite-state automaton  $\langle \Sigma, S_R, I_R, F_R, \rho_R \rangle$  defined as:

- $S_R = S \times R$ , where  $R$  is the set of all possible **regions** determined by  $A$
- $I_R = \{(s, \|(0, 0, \dots, 0)\|) \mid s \in I\}$ , corresponding to **initial** locations with all clocks **reset** to zero
- $F_R = \{(s, r \mid s \in F\}$  corresponding to **final** locations
- $\rho_R(\sigma, (s, r)) = \left\{ (s_2, r_2) \left| \begin{array}{l} (s, \sigma, c, \rho, s_2) \in E, \text{ there exists} \\ r' \in \widetilde{r} \text{ such that } r' \text{ satisfies } c, \\ \text{and } r_2 = r'[x \mapsto 0 \mid x \in \rho] \text{ is} \\ r' \text{ with all clocks } \rho \text{ reset to zero} \end{array} \right. \right\}$

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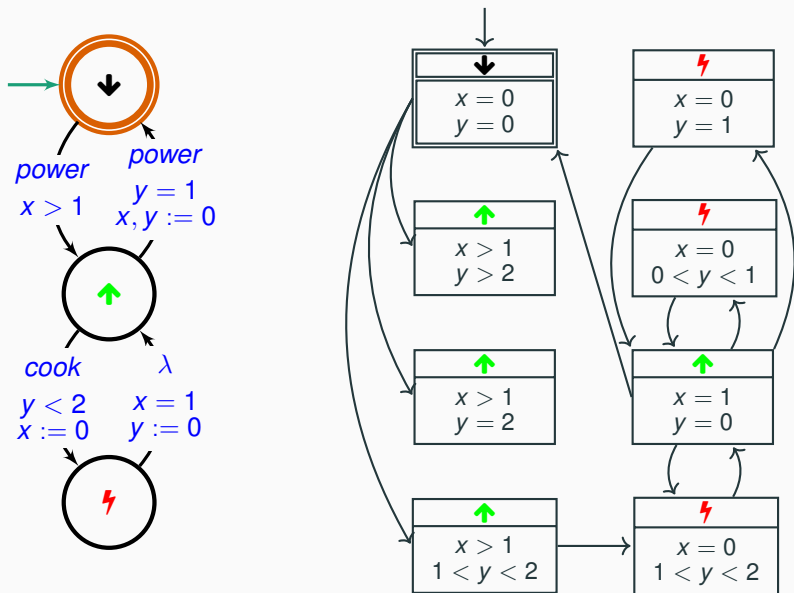
**Emptiness:**  $\mathcal{L}(\|A\|) = \emptyset$  if and only if  $\mathcal{L}(A) = \emptyset$

Since  $\|A\|$  is a finite-state automaton, emptiness of a TA  $A$  is decidable by checking reachability on its region automaton  $\|A\|$ .



# Region automaton: example

The region automaton of the running example's TA is not empty:



# **Real-time model checking**

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## **Complexity, variants, and tools**

# Complexity of the real-time model-checking algorithm

Real-time model-checking **algorithm**: given a timed automaton  $A$  and a metric temporal logic property  $P$ :

1. **monitor**: build  $\mathcal{A}(\neg P)$
2. **intersection**: build  $A \times \mathcal{A}(\neg P)$
3. **emptiness**: test whether  $\mathcal{L}(A \times \mathcal{A}(\neg P)) = \emptyset$

Let's analyze the complexity of **each step** in the algorithm.

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**EMPTINESS** of a timed automaton  $A$  is linear in the size of  $\|A\|$ , which has  $2^{O(|A|)}$  states (in particular, the number of regions is exponential in the number of clocks)

# Complexity of the timed automata emptiness problem

The emptiness **problem** for timed automata is **PSPACE**-complete.

Therefore, the region-automaton construction is worst-case optimal.

There are algorithms that perform better **in practice** – based on more compact **symbolic** representations of regions (such as **zones**).

# Complexity of the real-time model-checking problem

What is the worst-case complexity of the real-time model-checking **problem** as a whole?

The model-checking **problem** for timed automata and MTL formulas is:

- **nonprimitive recursive** for dense time domains and finite words
- **undecidable** for dense time domains and infinite words
- **EXPSPACE**-complete for discrete time



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Dense time and finite words is the interpretation we have presented in this class. Nonprimitive recursive means that it is decidable but "barely so".

This result also implies that there is a way of performing real-time model checking over finite words that does **not require** to build timed automata **monitors** of MTL (since this is not always possible).

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In contrast, if we stick with **discrete** time, timed automata and MTL are just **exponentially more succinct** versions of finite-state automata and LTL – and hence fully decidable.

# Real-time and hybrid model checking tools

Model checkers and analyzers for **real-time** and **hybrid** models include:

**UPPAAL** is a verification system for **timed automata** – the first practical tool for such analysis – which has been extended with support for hybrid models and analysis that go beyond model checking (for example, synthesis)

**PAT** is an **extensible** model checking framework, including functionalities for **real-time models** and analysis as well as more traditional automata-based model checking

**Prism** is a model checker for **stochastic** models, which also support a notion of timed behavior

**FDR4** is a modern reimplement of the classic FDR2 model checker for **CSPs** (Communicating Sequential Processes) and timed CSPs

# Summary

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## Case studies and industrial usage

Model checking has been the first formal verification technique to gain significant **industrial adoption**.

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Model checking has been the first formal verification technique to gain significant **industrial adoption**.

The **hardware** industry – especially big companies such as Intel – has been using, for over 20 years, model checking tools to verify parts of **circuit logic**. Since electronic circuits are intrinsically **finite state**, but with lots of parallelism, model checking is an ideal technique since it is sound and complete.

## Case studies and industrial usage

Model checking has been the first formal verification technique to gain significant **industrial adoption**.

Adoption of model checking for **software** has been more selective and context dependent:

- **NASA** has been using the SPIN model checker to detect concurrency bugs, as part of a more general methodology to develop reliable control software
- **Microsoft** used the **Slam** model checker (the first implementation of CEGAR model checking) to verify memory safety of Windows **device drivers**

## Case studies and industrial usage

Model checking has been the first formal verification technique to gain significant **industrial adoption**.

**Real-time** model checking is still not widespread – mainly due to the high complexity of the problem – but some of its techniques have become part of **modern control theory**, and hence connected to practice that way.



## Case studies and industrial usage

Model checking has been the first formal verification technique to gain significant **industrial adoption**.

On a different level, model checking's **success** has helped make all of **formal verification technology** more appealing and practically useful.

# Model checking: techniques

**Model checking** denotes analysis techniques for **finite-state** models and properties expressed in some form of **temporal** logic.

The classic automata-based **techniques** are based on **reachability** on finite (but possibly very large) graphs.

**soundness/completeness:** **sound** and **complete** with respect to the finite-state model, which typically is an under-approximation of an infinite-state system

**complexity:** complex (exponential and more) but decidable

**automation:** fully automated (“**push button**”)

**expressiveness:** arbitrary properties expressible in **temporal logic** – such as event ordering, reachability, termination

# Model checking: tools and practice

Model checking **tools** have been highly optimized so that they scale to systems with billions of states. Model checkers can return an explicit **counterexample** when verification fails; this significantly helps **usability**.

**Case studies** of model checking include an extensive usage in **hardware** verification and **concurrency** debugging. Software model checking has been applied to system code such as device drivers.

Main outstanding **challenges**:

- further improving **scalability** (fighting state-space explosion)
- in **predicate** abstraction: supporting advanced program **features** and **inter**-procedural analysis
- in **real-time** model checking: finding expressive yet **tractable** models

## Credits and references

The presentation of automata-theoretic model checking revisits Vardi's classic presentation (Automata-theoretic model checking revisited, VMCAI, 2007; Automata-theoretic techniques for temporal reasoning) using finite words.

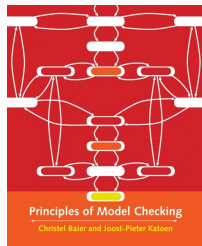
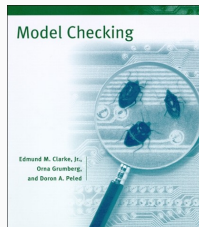
Parts of the presentation of model checking and real-time model checking are based on chapters in Furia et al.: Modeling time in computing, Springer, 2012.

The presentation of Boolean programs and predicate abstraction is derived from Ball and Rajamani's for the software model checker SLAM.

The region automaton construction follows the original in Alur and Dill: A theory of timed automata, TCS, 1994.

## Further reading

For a comprehensive presentation of model checking see one of these two textbooks:



For more details on real-time models and verification techniques:  
Modeling time in computing.



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