

Algorithms for Model Checking (2IW55)

Lecture 2

Symbolic Model Checking for CTL

("Model Checking", Chapter 2, 6.1, 6.2. Also read Chapter 5.)

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Fixed Points

Fixed Point Algorithm for CTL

Symbolic Model Checking

Model checking complexity:

- ▶ In general, there are **infinitely many states and transitions**.
- ▶ Many of the states behave very similarly (e.g. the start value of some variables may not matter)
- ▶ We're interested in an algorithm that can benefit from this.

Consider a Kripke Structure $M = \langle S, R, L \rangle$

In what follows, we (temporarily) ignore the difference between syntax and semantics

- ▶ Identify **sets of states** and **predicates on states**
- ▶ So, two notations are often mixed:
 - subsets: $X \subseteq S$ or $X \in \mathcal{P}(S)$, versus
 - predicates: $X \in 2^S$ or $X : S \rightarrow \{0, 1\}$
 $s \in X \Leftrightarrow X(s) = 1$ and $s \notin X \Leftrightarrow X(s) = 0$
- ▶ In general: we identify CTL formulae with the set of states where they hold: f versus $\{s \mid s \models f\}$
- ▶ We freely mix \forall, \wedge and \cup, \cap : compare $\emptyset \cup E G f$ and $\text{false} \vee E G f$

Predicate Transformers and Monotonicity

Consider a Kripke Structure $M = \langle S, R, L \rangle$

- ▶ The set $(\mathcal{P}(S), \subseteq)$ is a **complete lattice**.
- ▶ A **predicate transformer** is a function on predicates. For example, the relations *Pre* and *Post* that lift the transition relation *R* to **sets** of states:

$$\begin{aligned} \text{Pre}_R(X) &= \{s \in S \mid \exists t \in X. s R t\} \\ \text{Post}_R(X) &= \{t \in S \mid \exists s \in X. s R t\} \end{aligned}$$

- ▶ Let $\tau : \mathcal{P}(S) \rightarrow \mathcal{P}(S)$ be an arbitrary predicate transformer.
- ▶ τ is **monotonic** iff $P \subseteq Q$ implies $\tau(P) \subseteq \tau(Q)$.
- ▶ We write $\tau^i(X)$ for applying τ *i* times to *X*:

$$\begin{cases} \tau^0(X) &= X \\ \tau^{i+1}(X) &= \tau(\tau^i(X)) \end{cases}$$

Let $\tau : \mathcal{P}(S) \rightarrow \mathcal{P}(S)$.

- ▶ A **fixed point** of τ is a set Z such that $\tau(Z) = Z$
- ▶ The **least fixed point** of τ , denoted $\mu X. \tau(X)$ is a set Z such that:
 - $Z = \tau(Z)$ (i.e. Z is a fixed point)
 - for all X , if $\tau(X) = X$, then $Z \subseteq X$
- ▶ The **greatest fixed point** of τ , denoted $\nu X. \tau(X)$ is a set Z such that:
 - $Z = \tau(Z)$ (i.e. Z is a fixed point)
 - for all X , if $\tau(X) = X$, then $X \subseteq Z$

A theorem by Tarski: a **monotonic** operator on $\mathcal{P}(S)$ always has least and greatest fixed points:

- ▶ $\mu Z. \tau(Z) = \bigcap \{X \mid \tau(X) \subseteq X\}$
- ▶ $\nu Z. \tau(Z) = \bigcup \{X \mid X \subseteq \tau(X)\}$

Assume now that:

- ▶ S (hence also $\mathcal{P}(S)$) is finite, and
- ▶ $\tau : \mathcal{P}(S) \rightarrow \mathcal{P}(S)$ is monotonic

Then:

1. $\forall i. \tau^i(\emptyset) \subseteq \tau^{i+1}(\emptyset)$ (induction on i and monotonicity)
2. There exists an i such that $\tau^i(\emptyset) = \tau^{i+1}(\emptyset)$ (sets become bigger and S is finite)
3. If $\tau^i(\emptyset) = \tau^{i+1}(\emptyset)$, then $\tau^i(\emptyset)$ is a fixed point of τ (by definition)
4. If X is a fixed point of τ , then $\forall i. \tau^i(\emptyset) \subseteq X$ (induction on i and monotonicity)

So an approximant τ^i can be found such that $\tau^i(\emptyset) = \tau^{i+1}(\emptyset)$, and this set is the least fixed point of τ .

Similarly, the smallest i such that $\tau^i(S) = \tau^{i+1}(S)$ yields the greatest fixed point.

Algorithms for computing the least fixed point and the greatest fixed point based on the observations on the previous slide.

```
function lfp( $\tau:\mathcal{P}(S)\rightarrow\mathcal{P}(S)$ ) :  $\mathcal{P}(S)$ 
```

```
   $Q := \emptyset$ ;
```

```
   $Q' := \tau(Q)$ ;
```

```
  while  $Q \neq Q'$  do
```

```
     $Q := Q'$ ;
```

```
     $Q' := \tau(Q')$ ;
```

```
  end while
```

```
  return  $Q$ ;
```

```
end function
```

```
function gfp( $\tau:\mathcal{P}(S)\rightarrow\mathcal{P}(S)$ ) :  $\mathcal{P}(S)$ 
```

```
   $Q := S$ ;
```

```
   $Q' := \tau(Q)$ ;
```

```
  while  $Q \neq Q'$  do
```

```
     $Q := Q'$ ;
```

```
     $Q' := \tau(Q')$ ;
```

```
  end while
```

```
  return  $Q$ ;
```

```
end function
```


Fixed Points

Fixed Point Algorithm for CTL

Symbolic Model Checking

Recall that CTL has the following ten temporal operators:

- ▶ $A X$ and $E X$: for all/some next state
- ▶ $A F$ and $E F$: inevitably and potentially
- ▶ $A G$ and $E G$: invariantly and potentially always
- ▶ $A [U]$ and $E [U]$: for all/some paths, until
- ▶ $A [R]$ and $E [R]$: for all/some paths, releases

Besides atomic propositions (AP), the constant true and the Boolean connectives (\neg, \vee), the following temporal operators are sufficient: $E X$, $E G$, $E [U]$.

Hence: only algorithms for computing formulae of the above form are needed.

CTL operators can be seen as fixed point operators. Fix a Kripke Structure $M = \langle S, R, L \rangle$. Identify a CTL formula f with predicate $\{s \mid s \models f\}$.

- ▶ $A X f = \neg E X \neg f$ and $E X f = Pre_R(f)$
- ▶ $A F f = \mu Z. f \cup A X Z$ and $E F f = \mu Z. f \cup E X Z$
- ▶ $A G f = \nu Z. f \cap A X Z$ and $E G f = \nu Z. f \cap E X Z$
- ▶ $E [f U g] = \mu Z. g \cup (f \cap E X Z)$

Intuition:

- ▶ least and greatest fixed points deal differently with **loops**:
 - Greatest fixed point: recursion includes loops, so possibly infinitely many “steps”
 - Least fixed point: finite recursion through loops, so only finitely many “steps”
- ▶ Eventualities least fixed points
(a **witness** of the eventuality is needed in finitely many steps)
- ▶ Globally greatest fixed points
(an infinite path without error is OK)

Proof obligations for E G :

1. The transformer $Z \mapsto f \wedge E X Z$ is monotonic, so its fixed point can be computed by iteration, see lfp and gfp
(If $Z_1 \subseteq Z_2$ then $f \wedge E X Z_1 \subseteq f \wedge E X Z_2$).
2. E G f is a fixed point of $Z \mapsto f \wedge E X Z$
(E G $f = f \wedge E X E G f$)
3. E G f is the largest such fixed point
(for all Z : if $Z = f \wedge E X Z$, then $Z \subseteq E G f$)
 - ▶ For 1,2,3: prove $X \subseteq Y$ by $\forall s. s \in X \Rightarrow s \in Y$.
 - ▶ For 2: prove \subseteq and \supseteq .
 - ▶ For 2,3: use the semantics of CTL-formulae

Proof obligations for E [U] are similar (see for yourself)

Proofs for (2):

$E G f = f \wedge E X E G f$ follows from:

- ▶ “ \subseteq ”. Let $s \in E G f$. Then $\exists \pi \in \text{path}(s) : \forall i : \pi^i \models f$. Let π be such a path starting in s . We have

$$\pi^0 \models f \tag{a}$$

$$\forall j : \pi^{j+1} \models f \tag{b}$$

Since f is a state formula (we are in the CTL fragment), from (a) we conclude $s \models f$. By definition, from (b) we conclude $\pi^1 \models G f$. Hence $\pi^1(0) \models E G f$. Since $s R \pi^1(0)$, we have $s \models E X E G f$. Thus $s \in f \wedge E X E G f$.

- ▶ “ \supseteq ”. Let $s \in f \wedge E X E G f$. Then $s \in f$ and $s R t$ for some $t \in E G f$. But then also $s \in E G f$.

Proofs sketches for (3):

Let $\tau(Z) = f \wedge E X Z$. We show that $E G f = \bigcap_i \tau^i(S)$ by proving both inclusions.

- ▶ “ \subseteq ”. By means of an induction on i . Clearly $E G f \subseteq S = \tau^0(S)$. Assume $E G f \subseteq \tau^i(S)$. Then by monotonicity, we have $\tau(E G f) \subseteq \tau^{i+1}(S)$. Since $E G f$ is a fixpoint of τ , we have $\tau(E G f) = E G f$. So $E G f \subseteq \tau^{i+1}(S)$.
- ▶ “ \supseteq ”. Let $s \in \bigcap_i \tau^i(S)$. Observe that since S is finite and τ is monotonic, $\tau^k(S) = \tau^{k+1}(S)$ for some k , and $\tau^l(S) \supseteq \tau^k(S)$ for all $l \leq k$. Then, for every $s \in \tau^k(S)$, we have (by definition of τ) $s \in f$ and $s \in E X \tau^k(S)$. So, for all $t \in \tau^k(S)$, we have $t R u$ for some $u \in \tau^k(S)$. Therefore, there must be some infinite path starting in s , satisfying f . Thus, $s \in E G f$.

CTL model checking with Fixed Points

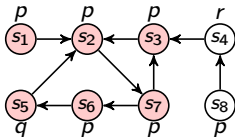
Function $\text{check}(f)$ takes a formula f and returns the set of states where f holds: $\{s \mid s \models f\}$ (given a fixed Kripke Structure $M = \langle S, R, L \rangle$).

$\text{check}(p)$	$\{s \mid p \in L(s)\}$
$\text{check}(\neg f)$	$S \setminus \text{check}(f)$
$\text{check}(f \vee g)$	$\text{check}(f) \cup \text{check}(g)$
$\text{check}(E X f)$	$\text{Pre}_R(\text{check}(f))$
$\text{check}(E [f U g])$	$\text{lfp}(Z \mapsto \text{check}(g) \cup (\text{check}(f) \cap \text{Pre}_R(Z)))$
$\text{check}(E G f)$	$\text{gfp}(Z \mapsto \text{check}(f) \cap \text{Pre}_R(Z))$

Recall: $\text{Pre}_R(Z) = \{s \in S \mid \exists t \in Z. s R t\}$

Example

- ▶ To check: $E [p U q]$
- ▶ Compute: $\mu Z. q \vee (p \wedge E X Z)$ (with lfp)



$$Z_0 = \text{false} = \emptyset$$

$$Z_1 = q \vee (p \wedge E X Z_0) = \{s_5\}$$

$$Z_2 = q \vee (p \wedge E X Z_1) = \{s_5, s_6\}$$

$$Z_3 = q \vee (p \wedge E X Z_2) = \{s_5, s_6, s_7\}$$

$$Z_4 = q \vee (p \wedge E X Z_3) = \{s_2, s_5, s_6, s_7\}$$

$$Z_5 = q \vee (p \wedge E X Z_4) = \{s_1, s_2, s_3, s_5, s_6, s_7\}$$

$$Z_6 = q \vee (p \wedge E X Z_5) = \{s_1, s_2, s_3, s_5, s_6, s_7\}$$

$Z_5 = Z_6$, so this is the least fixed point.

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Example (GCD)

Consider the following program:

```
repeat
  if  $x > y$   $\rightarrow$   $x := x - y$ ;
  []  $x < y$   $\rightarrow$   $y := y - x$ ;
fi
until false
```

This program uses:

- ▶ **variables**: $\{x, y\}$, with an (implicit) **domain** of variables: \mathbb{N}
- ▶ **States** of this program are **functions** of type: $\{x, y\} \rightarrow \mathbb{N}$
- ▶ An example state could be: $\{x \mapsto 5, y \mapsto 15\}$
- ▶ An **execution** is a sequence of transitions: e.g.

$$\{x \mapsto 5, y \mapsto 15\} \rightarrow \{x \mapsto 5, y \mapsto 10\} \rightarrow \{x \mapsto 5, y \mapsto 5\} \rightarrow \{x \mapsto 5, y \mapsto 5\} \rightarrow \dots$$

Example (SWAP)

Consider the following program fragment:

```
z := x;    % l1
x := y;    % l2
y := z;    % l3
```

- ▶ Besides variables $x, y, z : \mathbb{N}$, this program has a **program counter**, whose values are **labels** (line numbers)
- ▶ Let $pc : \{l_1, l_2, l_3\}$. Now, a state is a function that gives a value to $\{x, y, z, pc\}$
- ▶ A possible **execution** is the following sequence:

$$\begin{aligned} & \{x \mapsto 5, y \mapsto 15, z \mapsto 500, pc \mapsto l_1\} \\ \rightarrow & \{x \mapsto 5, y \mapsto 15, z \mapsto 5, pc \mapsto l_2\} \\ \rightarrow & \{x \mapsto 15, y \mapsto 15, z \mapsto 5, pc \mapsto l_3\} \\ \rightarrow & \{x \mapsto 15, y \mapsto 5, z \mapsto 5, pc \mapsto l_4\} \end{aligned}$$

Idea: the set of states can be represented very **concisely** by a number of formulae

▶ for GCD:

- initial set of states: $x < 100 \wedge y < 100$
- next state predicate:

$$(x > y \wedge x' = x - y \wedge y' = y) \vee (x < y \wedge y' = y - x \wedge x' = x)$$

▶ for SWAP:

- initial states: $x = 5 \wedge y = 15$
- next state predicate:

$$(pc = l_1 \wedge pc' = l_2 \wedge z' = x \wedge \dots) \vee \dots$$

The system specification is represented by propositional logic formula

- ▶ Let V be a set of **variables** v_0, v_1, \dots, v_n
- ▶ Let D be the **domain** of these variables
- ▶ The **states** of the Kripke Structure will be functions $\nu : V \rightarrow D$
- ▶ A formula $S_0(V)$ represents the initial states
- ▶ Let V' be a copy of the variables in V : v'_0, v'_1, \dots, v'_n
- ▶ A formula $\mathcal{R}(V, V')$ represents the transition relation.
 - V denotes the value of the variables **before** the transition
 - V' denotes the value of the variables **after** the transition.

Example

- ▶ $V = \{TL_1, TL_2\}$,
- ▶ $D = \{r(ed), y(ellow), g(reen)\}$
- ▶ $S_0(TL_1, TL_2) := TL_1 = r \wedge TL_2 = r$
- ▶ $\mathcal{R}(TL_1, TL_2, TL'_1, TL'_2) := R_1 \vee R_2 \vee R_3 \vee R_4 \vee R_5 \vee R_6$, where:
 - $R_1 := TL_1 = r \wedge TL'_1 = g \wedge TL'_2 = TL_2$
 - $R_2 := TL_1 = g \wedge TL'_1 = y \wedge TL'_2 = TL_2$
 - $R_3 := TL_1 = y \wedge TL'_1 = r \wedge TL'_2 = TL_2$
 - $R_4 := TL_2 = r \wedge TL'_2 = g \wedge TL'_1 = TL_1$
 - $R_5 := TL_2 = g \wedge TL'_2 = y \wedge TL'_1 = TL_1$
 - $R_6 := TL_2 = y \wedge TL'_2 = r \wedge TL'_1 = TL_1$

Notes:

- ▶ this corresponds to a Kripke Structure modelling an **unsafe** traffic light system at a junction
- ▶ a specification for n traffic lights gives $O(3^n)$ states \Rightarrow **State space explosion**

We wish to avoid representing the state space and its subsets explicitly. To efficiently implement symbolic model checking, we need:

- ▶ A concise representation of sets of states
- ▶ Quick operations for:
 - Boolean operators \wedge, \vee, \neg
 - Existential quantification (for the relational composition)
 - Equivalence test

Solution: *Ordered Binary Decision Diagrams (OBDD)*

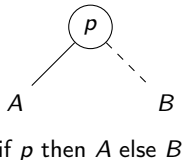
- ▶ Symbolic model checking is restricted to finite Kripke Structures
- ▶ All finite data can be encoded in “bits”
- ▶ Boolean functions can be represented **concisely** as (Ordered) Binary Decision Diagrams
- ▶ Binary Decision Diagrams are **directed acyclic graphs**, with the following ingredients:

1

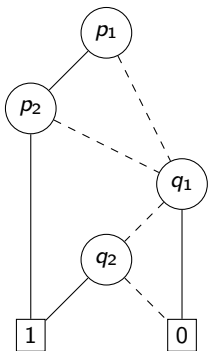
True

0

False



BDD representation of $(p_1 \wedge p_2) \vee (\neg q_1 \wedge q_2)$:



- ▶ In **ordered** BDDs, tests along a path occur in a **fixed** order (e.g. $p_1 < p_2 < q_1 < q_2$).
- ▶ Theorem[Bryant'86]: OBDDs are a **unique** representation for Boolean Functions.
- ▶ Claim: many practical formulae have a **concise OBDD representation** due to maximal sharing
- ▶ Disclaimer 1: some small formulae have only exponentially large BDDs. (multiplier)
- ▶ Disclaimer 2: the size of an OBDD can **crucially** depend on the ordering of the variables

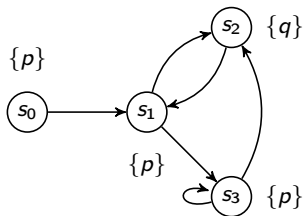
More on OBDDs:

- ▶ OBDDs are implemented as maximally shared **pointer structures** in memory.
- ▶ The order of variables is fixed (some implementations feature **dynamic reordering**)
- ▶ **Equivalence test** can be performed in **constant time**, in particular, also checking for **satisfiability** and **tautology**.
- ▶ Boolean operations can be performed efficiently. Let B_1 and B_2 be OBDDs with m and n nodes, respectively, then:
 - OBDDs for $B_1 \wedge B_2$ and $B_1 \vee B_2$ can be computed in $\mathcal{O}(m \cdot n)$ time.
 - OBDDs for $\neg B_1$ can be computed in $\mathcal{O}(m)$ time.
 - the OBDD of $\exists x. B_1$ can be computed in $\mathcal{O}(m^2)$ time.
- ▶ Note: still a formula of size $\mathcal{O}(n)$ may have a BDD of size $\mathcal{O}(2^n)$.

Attend *Automated Reasoning* (2IW15) for more information on OBDDs (Semester A.2).

- ▶ The implementation of a **symbolic model checking** relies on a representation of all sets in check , lfp and gfp by OBDDs.
- ▶ Hence, in summary, symbolic model checking:
 - **Recursively** processes subformulae
 - Represent the set of states satisfying a subformula by **OBDDs**
 - Treats temporal operators by **fixed point computations**
 - Relies on **efficient implementation** of equivalence test, and \wedge , \vee , \neg and \exists connectives on OBDDs.

Consider the following Kripke Structure:



Consider the following formulae, where p and q are atomic propositions:

- (A) $\mathbf{A}(\mathbf{F}(q))$
- (B) $\mathbf{A}[q \mathbf{R} p]$

Determine the set of states where (A) and (B) hold using the symbolic model checking algorithm for CTL. You may use explicit set notation to represent states.