

## 4. Computation Tree Logic (3)

Huixing Fang

School of Information Engineering  
Yangzhou University

- 1 Reduced OBDD
- 2 Implementation of ROBDD-Based Algorithms
- 3 CTL\*

# 1 Reduced OBDD

## Definition 1 (Reduced OBDD)

Let  $\mathfrak{B}$  be a  $\wp$ -OBDD.  $\mathfrak{B}$  is called **reduced** if for every pair  $(v, w)$  of nodes in  $\mathfrak{B}$  :

$$v \neq w \implies f_v \neq f_w.$$

Let  $\wp$ -ROBDD denote a reduced  $\wp$ -OBDD.

- In reduced  $\wp$ -OBDDs any  $\wp$ -consistent cofactor is represented by exactly one node
- This is the key property to prove that reduced OBDDs yield a universal and canonical data structure for switching functions

# 1 Reduced OBDD

## Theorem 2 (Universality and Canonicity of ROBDDs)

Let  $Var$  be a finite set of Boolean variables and  $\wp$  a variable ordering for  $Var$ . Then:

- 1 For each switching function  $f$  for  $Var$  there exists a  $\wp$ -ROBDD  $\mathfrak{B}$  with  $f_{\mathfrak{B}} = f$
- 2 Given two  $\wp$ -ROBDDs  $\mathfrak{B}$  and  $\mathfrak{C}$  with  $f_{\mathfrak{B}} = f_{\mathfrak{C}}$ , then  $\mathfrak{B}$  and  $\mathfrak{C}$  are isomorphic, i.e., agree up to renaming of the nodes

## Corollary 3 (Minimality of Reduced OBDDs)

Let  $\mathfrak{B}$  be a  $\wp$ -OBDD for  $f$ . Then,  $\mathfrak{B}$  is reduced if and only if  $size(\mathfrak{B}) \leq size(\mathfrak{C})$  for each  $\wp$ -OBDD  $\mathfrak{C}$  for  $f$ .

- 1 Reduced OBDD
  - Reduction Rules
  - Variable Ordering Problem
- 2 Implementation of ROBDD-Based Algorithms
- 3 CTL\*

## 1.1 Reduction Rules

When the variable ordering  $\wp$  for  $Var$  is fixed, then **reduced**  $\wp$ -OBDDs provide **unique** representations of switching functions for  $Var$

- 1 **Elimination rule:** If  $v$  is an **inner node** of  $\mathfrak{B}$  with

$$succ_0(v) = succ_1(v) = w,$$

then remove  $v$  and redirect all incoming edges  $u \rightsquigarrow v$  to  $w$

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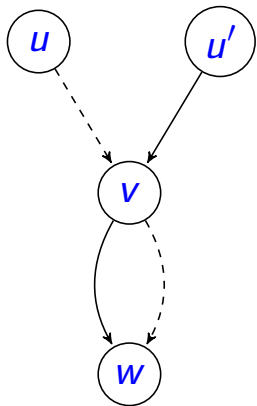
then remove  $v$  and redirect all incoming edges  $u \rightsquigarrow v$  to  $w$

- ② **Isomorphism rule:** If  $v, w$  are nodes in  $\mathfrak{B}$  with  $v \neq w$  and
- either  $v, w$  are drains (i.e., terminal nodes) with  $val(v) = val(w)$  or
  - $v, w$  are inner nodes with

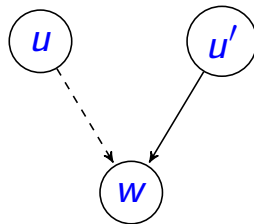
$$\langle var(v), succ_1(v), succ_0(v) \rangle = \langle var(w), succ_1(w), succ_0(w) \rangle$$

then remove node  $v$  and redirect all incoming edges  $u \rightsquigarrow v$  to node  $w$

# 1.1 Reduction Rules

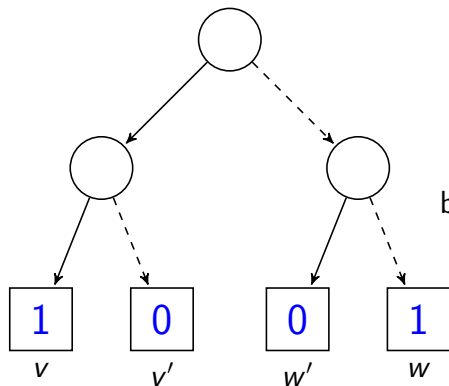


becomes

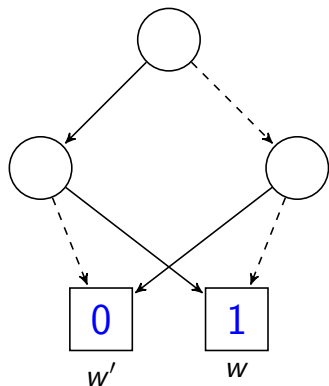




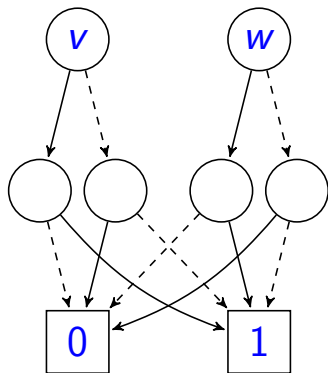
# 1.1 Reduction Rules



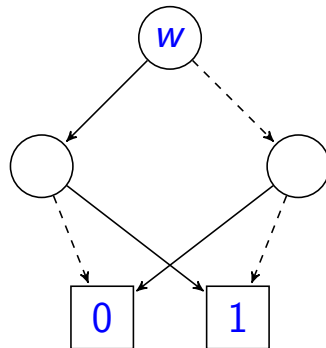
becomes



# 1.1 Reduction Rules



becomes



## 1.1 Reduction Rules

Both reduction rules are sound in the sense that they do not affect the semantics.

- 1 For the elimination rule applied to a  $z$ -node  $v$  with  $w = succ_0(v) = succ_1(v)$ , we have

$$f_v = (\neg z \wedge f_{succ_0(v)}) \wedge (z \wedge f_{succ_1(v)}) = (\neg z \wedge f_w) \wedge (z \wedge f_w) = f_w$$

- 2 If the isomorphism rule is applied to  $z$ -nodes  $v$  and  $w$  then

$$\begin{aligned} f_v &= (\neg z \wedge f_{succ_0(v)}) \wedge (z \wedge f_{succ_1(v)}) \\ &= (\neg z \wedge f_{succ_0(w)}) \wedge (z \wedge f_{succ_1(w)}) = f_w \end{aligned}$$

### Theorem 4 (Completeness of Reduction Rules)

$\wp$ -OBDD  $\mathfrak{B}$  is reduced if and only if no reduction rule is applicable to  $\mathfrak{B}$ .

- 1 Reduced OBDD
  - Reduction Rules
  - Variable Ordering Problem
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## 1.2 Variable Ordering Problem

Let  $m \geq 1$  and

$$f_m = (z_1 \wedge y_1) \vee (z_2 \wedge y_2) \vee \dots \vee (z_m \wedge y_m).$$

- 1 For the variable ordering

$$\wp = (z_m, y_m, z_{m-1}, y_{m-1}, \dots, z_1, y_1),$$

the  $\wp$ -ROBDD for  $f_m$  has  $2m + 2$  nodes,

- 2 while  $\Omega(2^m)$  nodes are required for the ordering

$$\wp' = (z_1, z_2, \dots, z_m, y_1, \dots, y_m).$$

## 1.2 Variable Ordering Problem

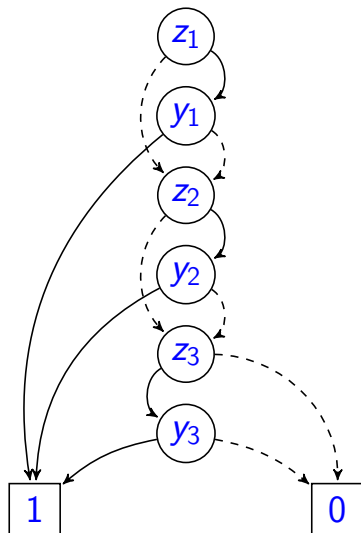


Figure: ROBDD for  $f_3 = (z_1 \wedge y_1) \vee (z_2 \wedge y_2) \vee (z_3 \wedge y_3)$

## 1.2 Variable Ordering Problem

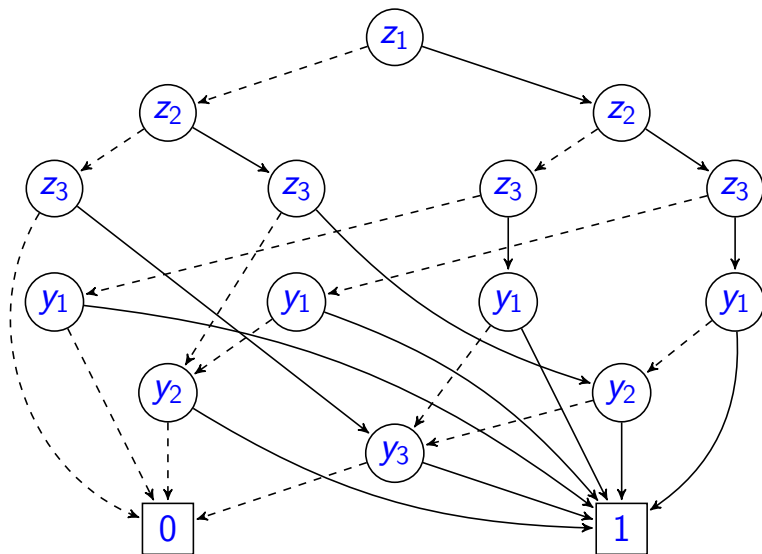


Figure: ROBDD for  $f_3 = (z_1 \wedge y_1) \vee (z_2 \wedge y_2) \vee (z_3 \wedge y_3)$

## 1.2 Variable Ordering Problem

- ① The size of ROBDDs is dependent on the variable ordering
- ② To check whether a variable ordering is optimal is NP-hard
- ③ Many switching functions with large ROBDDs

### Sifting algorithm

Dynamic variable ordering using variable swapping:

- ① Select a variable  $x_i$  in OBDD
- ② Successively swap  $x_i$  to determine  $size(\mathfrak{B})$  at any position for  $x_i$
- ③ Shift  $x_i$  to position for which  $size(\mathfrak{B})$  is minimal
- ④ Go back to the first step until no improvement is made



## 1.2 Variable Ordering Problem

### Symmetric Functions

$f \in Eval(z_1, \dots, z_m)$  is **symmetric** if and only if

$$f([z_1 = b_1, \dots, z_m = b_m]) = f([z_1 = b_{i_1}, \dots, z_m = b_{i_m}])$$

for **each** permutation  $(i_1, \dots, i_m)$  of  $(1, \dots, m)$ . Examples of symmetric functions:

- 1  $z_1 \vee z_2 \vee \dots \vee z_m$
- 2  $z_1 \wedge z_2 \wedge \dots \wedge z_m$
- 3 parity function,  $z_1 \oplus z_2 \oplus \dots \oplus z_m$ , returns  $\#(v \mapsto 1) = \text{odd}$
- 4 majority function, returns  $\#(v \mapsto 1) > \#(v \mapsto 0)$

## 1.2 Variable Ordering Problem

### Lemma 5 (ROBDD-Size for Symmetric Functions)

If  $f$  is a symmetric function with  $m$  **essential** variables, then for **each** variable ordering  $\wp$  the  $\wp$ -ROBDD has size  $O(m^2)$ .

- 1 Given a symmetric function  $f$  for  $m$  variables and  $\wp = (z_1, \dots, z_m)$ ,
- 2 then the  $\wp$ -consistent cofactors  $f|_{z_1 = b_1, \dots, z_i = b_i}$  and  $f|_{z_1 = c_1, \dots, z_i = c_i}$  agree for all bit tuples  $(b_1, \dots, b_i)$  and  $(c_1, \dots, c_i)$  that contain the same number of 1's.
- 3 Thus, there are at most  $i + 1$  different  $\wp$ -consistent cofactors of  $f$  that arise by assigning values to the first  $i$  variables
- 4 Hence, the total number of  $\wp$ -consistent cofactors is bounded above by  $\sum_{i=0}^m (i + 1) = O(m^2)$  □

## 1.2 Variable Ordering Problem

### ROBDDs vs. CNF/DNF

- 1 Given a CNF representation for  $f$ , the task to generate a CNF for  $\neg f$  is expensive, CNF for  $\neg f$  may have at least exponentially many clauses
- 2 However, negation is trivial as we may simply swap the values of the drains
- 3 For any variable ordering  $\wp$ , the  $\wp$ -ROBDDs for  $f$  and  $\neg f$  have the same size
- 4 Regard the satisfiability problem, is NP-complete for CNF, but trivial for ROBDDs, since  $f \neq 0$  if and only if the  $\wp$ -ROBDD for  $f$  does not contain a 0-drain
- 5 “linear time” for a ROBDD-based algorithm means linear in the size of the input ROBDDs

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## 2 Implementation of ROBDD-Based Algorithms

The efficiency of ROBDD-based algorithms crucially relies on implementation techniques

- 1 Use a **single** reduced decision **graph** with **one global** variable ordering  $\wp$  to represent several switching functions
- 2 All computations on these decision graphs are interleaved with the reduction rules to guarantee **redundance-freedom** at any time
- 3 The comparison of two represented **functions** simply requires checking equality of the **nodes** for them

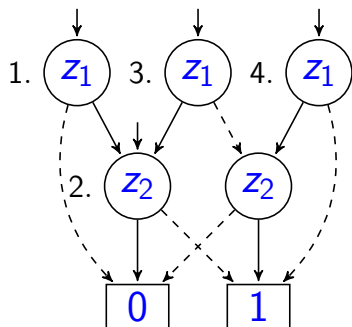
- 1 Reduced OBDD
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  - Shared OBDD
  - Synthesizing shared ROBDDs
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## 2.1 Shared OBDD

Combine several OBDDs with same variable ordering such that common  $\wp$ -consistent co-factors are shared

### Definition 6 (Shared OBDD)

- $Var$ : a finite set of Boolean variables
- $\wp$  a variable ordering for  $Var$
- **Shared**  $\wp$ -OBDD  $\bar{\mathfrak{B}} = (V, V_I, V_T, succ_0, succ_1, var, val, \bar{v}_0)$
- $\bar{v}_0 = (v_0^1, \dots, v_0^k)$  is a tuple of roots



1.  $z_1 \wedge \neg z_2$ , 2.  $\neg z_2$ , 3.  $z_1 \oplus z_2$  and 4.  $\neg z_1 \vee z_2$   
 $\wp$ -SOBDD is short for Shared  $\wp$ -OBDD

## 2.1 Shared OBDD

Symbolic CTL model-checking  $\Phi$ :

- 1 the shared OBDD  $\bar{\mathfrak{B}}$  with root nodes for  $\Delta$  and
- 2 the  $f_a$ 's, extended by new root nodes representing the characteristic functions of the satisfaction sets  $Sat(\Psi)$ ,  $\Psi \in Sub(\Phi)$
- 3  $\wp = (x_1, x'_1, \dots, x_n, x'_n)$



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## 2.2 Synthesizing shared ROBDDs

To support dynamic changes of the set of switching functions to be represented, two tables are required

- ① **unique table:** contains the relevant information about the nodes and serves to keep the diagram reduced
- ② **computed table:** for efficiency reasons

## 2.2 Synthesizing shared ROBDDs

### The unique table

For each **inner node**  $v$ , the entries of the unique table are triples of the form

$$info(v) = \langle var(v), succ_1(v), succ_0(v) \rangle$$

- 1 contain the relevant information which is necessary for the applicability of the **isomorphism rule**
- 2 accessing via  $op \triangleq find\_or\_add(z, v_1, v_0)$  with  $v_1 \neq v_0$ , and
- 3 return  $v$  **if there exists** a node  $v = \langle z, v_1, v_0 \rangle$  in the ROBDD, or
- 4 create a **new**  $z$ -node  $v$  with  $succ_0(v) = v_0$  and  $succ_1(v) = v_1$
- 5 this  $op$  can be implemented using **hash functions**

## 2.2 Synthesizing shared ROBDDs

### ITE operator

ITE(“if-then-else”) operator for switching functions  $g$ ,  $f_1$ , and  $f_2$ :

$$ITE(g, f_1, f_2) = (g \wedge f_1) \vee (\neg g \wedge f_2) .$$

- 1 ITE fits very well with the unique table by their info-triples

$$f_v = ITE(z, f_{succ_1(v)}, f_{succ_0(v)}) .$$

- 2  $ITE(0, f_1, f_2) = f_2$ ,  $ITE(1, f_1, f_2) = f_1$
- 3  $\neg f = ITE(f, 0, 1)$
- 4  $f_1 \vee f_2 = ITE(f_1, 1, f_2)$
- 5  $f_1 \wedge f_2 = ITE(f_1, f_2, 0)$
- 6  $f_1 \oplus f_2 = ITE(f_1, \neg f_2, f_2) = ITE(f_1, ITE(f_2, 0, 1), f_2)$
- 7  $f_1 \rightarrow f_2 = ITE(f_1, f_2, 1)$

## 2.2 Synthesizing shared ROBDDs

### Lemma 7 (Cofactors of ITE( $\cdot$ ))

If  $g, f_1, f_2$  are switching functions for  $Var$ ,  $z \in Var$  and  $b \in \{0, 1\}$ , then

$$ITE(g, f_1, f_2)|_{z=b} = ITE(g|_{z=b}, f_1|_{z=b}, f_2|_{z=b}) .$$

*Proof:* For any  $(a, \bar{c}) = [z = a, \bar{y} = \bar{c}] \in Eval(Var)$

$$\begin{aligned} & ITE(g, f_1, f_2)|_{z=b}(a, \bar{c}) \\ &= ITE(g, f_1, f_2)(b, \bar{c}) \\ &= (g(b, \bar{c}) \wedge f_1(b, \bar{c})) \vee (\neg g(b, \bar{c}) \wedge f_2(b, \bar{c})) \\ &= (g|_{z=b}(a, \bar{c}) \wedge f_1|_{z=b}(a, \bar{c})) \vee (\neg g|_{z=b}(a, \bar{c}) \wedge f_2|_{z=b}(a, \bar{c})) \\ &= ITE(g|_{z=b}, f_1|_{z=b}, f_2|_{z=b})(a, \bar{c}) . \end{aligned}$$



## 2.2 Synthesizing shared ROBDDs

A node in a  $\wp$ -SOBDD for representing  $ITE(g, f_1, f_2)$  is a node  $w$  such that  $info(w) = \langle z, w_1, w_0 \rangle$  where

- 1  $z$  is the minimal essential variable of  $ITE(g, f_1, f_2)$  according to  $<_{\wp}$
- 2  $w_1, w_0$  are SOBDD nodes with:

$$f_{w_1} = ITE(g|_{z=1}, f_1|_{z=1}, f_2|_{z=1})$$

$$f_{w_0} = ITE(g|_{z=0}, f_1|_{z=0}, f_2|_{z=0})$$

## 2.2 Synthesizing shared ROBDDs

This suggests a recursive algorithm

- 1 determine  $z = \min\{var(u), var(v_1), var(v_2)\}$
- 2 recursively computes the nodes for ITE applied to the cofactors of  $g = f_u, f_1 = f_{v_1}, f_2 = f_{v_2}$  for  $z$

## 2.2 Synthesizing shared ROBDDs

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**Algorithm 1:**  $ITE(u, v_1, v_2)$ (first version)

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```
1 if  $u$  is terminal then
2   if  $var(u) = 1$  then
3      $w := v_1$ ;
4   else
5      $w := v_2$ ;
6   end
7 else
8    $z := \min\{var(u), var(v_1), var(v_2)\}$ ;
9    $w_1 := ITE(u|_{z=1}, v_1|_{z=1}, v_2|_{z=1})$ ;
10   $w_0 := ITE(u|_{z=0}, v_1|_{z=0}, v_2|_{z=0})$ ;
11  if  $w_0 = w_1$  then
12     $w := w_1$ ;
13  else
14     $w := find\_or\_add(z, w_1, w_0)$ ;
15  end
16 end
17 return  $w$ ;
```



## 2.2 Synthesizing shared ROBDDs

### Lemma 8 (ROBDD size of $ITE(g, f_1, f_2)$ )

*The size of the  $\wp$ -ROBDD for  $ITE(g, f_1, f_2)$  is bounded by  $N_g \cdot N_{f_1} \cdot N_{f_2}$  where  $N_f$  denotes the size of the  $\wp$ -ROBDD for  $f$ .*

*Proof:* Let  $\wp = (z_1, \dots, z_m)$  and  $\mathfrak{B}_f$  denote the  $\wp$ -ROBDD for  $f$  where the nodes are the  $\wp$ -consistent cofactors of  $f$ . The node set of  $\mathfrak{B}_f$

$$V_f = \{f|_{z_1=b_1, \dots, z_i=b_i} \mid 0 \leq i \leq m, b_1, \dots, b_i \in \{0, 1\}\}$$

Note that several of the cofactors  $f|_{z_1=b_1, \dots, z_i=b_i}$  might agree.

## 2.2 Synthesizing shared ROBDDs

*Continue Proof:*

- By Lemma 7, the node set  $V_{ITE(g,f_1,f_2)}$  agrees with the set of switching functions

$$ITE(g|_{z_1=b_1, \dots, z_i=b_i}, f_1|_{z_1=b_1, \dots, z_i=b_i}, f_2|_{z_1=b_1, \dots, z_i=b_i})$$

where  $0 \leq i \leq m$ ,  $b_1, \dots, b_i \in \{0, 1\}$ .

- We can construct the surjective function

$$\iota : V_g \times V_{f_1} \times V_{f_2} \rightarrow V_{ITE(g,f_1,f_2)}.$$

Hence,

$$\begin{aligned} N_{ITE(g,f_1,f_2)} &= |V_{ITE(g,f_1,f_2)}| \\ &\leq |V_g \times V_{f_1} \times V_{f_2}| = N_g \cdot N_{f_1} \cdot N_{f_2} \quad \square \end{aligned}$$

## 2.2 Synthesizing shared ROBDDs

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### Algorithm 2: $ITE(u, v_1, v_2)$ (Use Computed Table)

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```
1 if there is an entry for  $(u, v_1, v_2, w)$  in the computed table then
2   | return  $w$ ;
3 else
4   | if  $u$  is terminal then
5     | if  $var(u) = 1$  then  $w := v_1$  else  $w := v_2$  ;
6   | else
7     |  $z := \min\{var(u), var(v_1), var(v_2)\}$ ;
8     |  $w_1 := ITE(u|_{z=1}, v_1|_{z=1}, v_2|_{z=1})$ ;
9     |  $w_0 := ITE(u|_{z=0}, v_1|_{z=0}, v_2|_{z=0})$ ;
10    | if  $w_0 = w_1$  then  $w := w_1$  else  $w := find\_or\_add(z, w_1, w_0)$  ;
11    | insert  $(u, v_1, v_2, w)$  in the computed table;
12  | end
13  | return  $w$ ;
14 end
```

- 1 Reduced OBDD
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- 3 CTL\***

- 1 **CTL\*** is an extension of CTL as it allows path quantifiers  $\exists$  and  $\forall$  to be **arbitrarily nested with linear temporal operators** such as  $\bigcirc$  and **U**
  - $\forall \bigcirc \bigcirc a$  is a legal CTL\* formula
  - but  $\forall \bigcirc \bigcirc a$  is not in CTL
- 2 In contrast, in **CTL** each linear temporal operator must be **immediately preceded** by a path quantifier

## 3.1 Syntax of CTL\*

CTL\* **state formulae**  $\Phi$  over  $AP$ , briefly called CTL\* formulae,

$$\Phi ::= \text{true} \mid a \mid \Phi_1 \wedge \Phi_2 \mid \neg\Phi \mid \exists\varphi$$

where  $a \in AP$  and  $\varphi$  is a **path formula**:

$$\varphi ::= \Phi \mid \varphi_1 \wedge \varphi_2 \mid \neg\varphi \mid \bigcirc\varphi \mid \varphi_1 \mathbf{U}\varphi_2$$

where  $\Phi$  is a **state formula**, and  $\varphi$ ,  $\varphi_1$ , and  $\varphi_2$  are **path formulae**

$$\diamond\varphi = \text{true}\mathbf{U}\varphi, \quad \square\varphi = \neg\diamond\neg\varphi, \quad \forall\varphi = \neg\exists\neg\varphi$$

## 3.2 Satisfaction Relation for CTL\*

The satisfaction relation  $\models$  is defined for state formulae by

$$s \models a \quad \text{iff} \quad a \in L(s),$$

$$s \models \neg\Phi \quad \text{iff} \quad \text{not } s \models \Phi,$$

$$s \models \Phi \wedge \Psi \quad \text{iff} \quad (s \models \Phi) \text{ and } (s \models \Psi)$$

$$s \models \exists\varphi \quad \text{iff} \quad \pi \models \varphi \text{ for some } \pi \in \text{Paths}(s).$$

## 3.2 Satisfaction Relation for CTL\*

For path  $\pi = s_0s_1s_2\dots$ , the satisfaction relation  $\models$  for path formulae is defined by:

$\pi \models \Phi$	iff	$s_0 \models \Phi,$
$\pi \models \varphi_1 \wedge \varphi_2$	iff	$(\pi \models \varphi_1)$ and $(\pi \models \varphi_2),$
$\pi \models \neg\varphi$	iff	$\pi \not\models \varphi,$
$\pi \models \bigcirc\varphi$	iff	$\pi[1..] \models \varphi,$
$\pi \models \varphi_1 \mathbf{U} \varphi_2$	iff	$\exists j \geq 0. (\pi[j..] \models \varphi_2 \wedge$ $(\forall 0 \leq k < j. \pi[k..] \models \varphi_1))$

- $\pi[i..]$  denotes the suffix of  $\pi$  from index  $i \geq 0$  on



### 3.3 CTL\* Semantics for Transition Systems

For CTL\*-state formula  $\Phi$ , the satisfaction set  $Sat(\Phi)$  is defined by

$$Sat(\Phi) = \{s \in S \mid s \models \Phi\} .$$

The transition system  $TS$  satisfies CTL\* formula  $\Phi$  if and only if  $\Phi$  holds in **all initial states** of  $TS$ :

$$TS \models \Phi \quad \text{if and only if} \quad \forall s_0 \in I. s_0 \models \Phi .$$

## 3.4 Embedding of LTL in CTL\*

For each LTL formula  $\varphi$  over  $AP$  and for each  $s \in S$ :

$$\underbrace{s \models \varphi}_{\text{LTL semantics}} \quad \text{iff} \quad \underbrace{s \models \forall \varphi}_{\text{CTL}^* \text{ semantics}}$$

$$\underbrace{TS \models \varphi}_{\text{LTL semantics}} \quad \text{iff} \quad \underbrace{TS \models \forall \varphi}_{\text{CTL}^* \text{ semantics}}$$

It is justified to understand LTL as a sublogic of CTL\*

## 3.5 CTL\* vs LTL and CTL

Theorem 9 ( CTL\* is More Expressive Than LTL and CTL)

For the CTL\* formula over  $AP = \{a, b\}$ ,

$$\Phi = (\forall \diamond \square a) \vee (\forall \square \exists \diamond b) ,$$

there does not exist any equivalent LTL or CTL formula

*Proof:*

- 1  $\forall \square \exists \diamond b$  is a CTL formula but not in LTL
- 2  $\diamond \square a$  is an LTL formula but not in CTL



## 3.5 CTL\* vs LTL and CTL

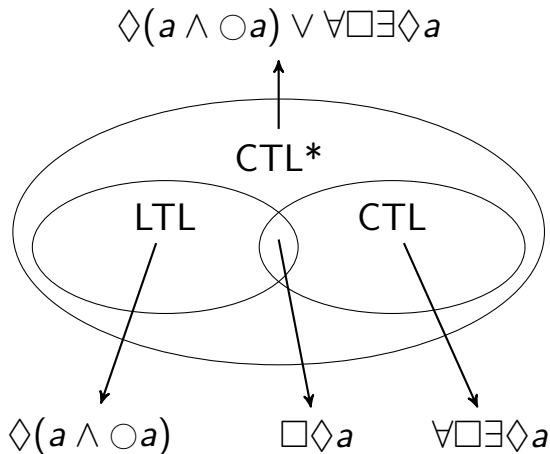


Figure: Relationship between LTL, CTL and CTL\*

## 3.5 CTL\* vs LTL and CTL

duality laws for path quantifiers

$$\neg \forall \varphi \equiv \exists \neg \varphi$$

$$\neg \exists \varphi \equiv \forall \neg \varphi$$

distributive laws

$$\forall (\varphi_1 \wedge \varphi_2) \equiv \forall \varphi_1 \wedge \forall \varphi_2$$

$$\exists (\varphi_1 \vee \varphi_2) \equiv \exists \varphi_1 \vee \exists \varphi_2$$

quantifier absorption laws

$$\forall \square \diamond \varphi \equiv \forall \square \forall \diamond \varphi$$

$$\exists \diamond \square \varphi \equiv \exists \diamond \exists \square \varphi$$

## 3.6 CTL<sup>+</sup>

CTL<sup>+</sup> extends CTL by allowing Boolean operators in path formulae:

- 1 CTL<sup>+</sup> state formulae  $\Phi$

$$\Phi ::= \text{true} \mid a \mid \Phi_1 \wedge \Phi_2 \mid \neg\Phi \mid \exists\varphi \mid \forall\varphi$$

- 2 CTL<sup>+</sup> path formulae  $\varphi$

$$\varphi ::= \varphi_1 \wedge \varphi_2 \mid \neg\varphi \mid \bigcirc\Phi \mid \Phi_1 \mathbf{U}\Phi_2$$

- 3 CTL<sup>+</sup> is as expressive as CTL

## 3.6 CTL<sup>+</sup>

For any CTL<sup>+</sup> state formula  $\Phi^+$  there exists an equivalent CTL formula  $\Phi$ . For example:

$$\underbrace{\exists(a\mathbf{W}b)}_{\text{CTL formula}} \equiv \underbrace{\exists((a\mathbf{U}b) \vee \square a)}_{\text{CTL}^+ \text{ formula}}$$
$$\underbrace{\exists(\diamond a \wedge \diamond b)}_{\text{CTL}^+ \text{ formula}} \equiv \underbrace{\exists \diamond (a \wedge \exists \diamond b) \wedge \exists \diamond (b \wedge \exists a)}_{\text{CTL formula}}$$

The transformation relies on equivalence laws such as

$$\exists(\neg\bigcirc\Phi) \equiv \exists\bigcirc\neg\Phi$$

$$\exists(\neg(\Phi_1 \mathbf{U} \Phi_2)) \equiv \exists((\Phi_1 \wedge \neg\Phi_2) \mathbf{U} (\neg\Phi_1 \wedge \neg\Phi_2)) \vee \exists\Box\neg\Phi_2$$

$$\exists(\bigcirc\Phi_1 \wedge \bigcirc\Phi_2) \equiv \exists\bigcirc(\Phi_1 \wedge \Phi_2)$$

⋮

- 1 CTL can be expanded by means of a Boolean operator for path formulae without changing the expressiveness
- 2 CTL<sup>+</sup> formulae can be much shorter than the shortest equivalent CTL formulae



## 3.7 CTL\* Model Checking

The model-checking algorithm for CTL\* is based on a bottom-up traversal of the syntax tree of  $\Phi$

### Definition 10 (Maximal Proper State Subformula)

State formula  $\Psi$  is a maximal proper state subformula of  $\Phi$  whenever

- 1  $\Psi$  is a subformula of  $\Phi$  that differs from  $\Phi$
- 2 and  $\Psi$  is not contained in any other proper state subformula of  $\Phi$

## 3.7 CTL\* Model Checking

Recursive computation of satisfaction sets:

$$\text{Sat}(\text{true}) = S$$

$$\text{Sat}(a) = \{s \in S \mid a \in L(s)\}$$

$$\text{Sat}(\Phi_1 \wedge \Phi_2) = \text{Sat}(\Phi_1) \cap \text{Sat}(\Phi_2)$$

$$\text{Sat}(\neg\Phi) = S \setminus \text{Sat}(\Phi)$$

$$\text{Sat}(\forall\varphi) = \text{Sat}_{LTL}(\varphi)$$

$$\text{Sat}(\exists\varphi) = S \setminus \text{Sat}_{LTL}(\neg\varphi)$$

## 3.7 CTL\* Model Checking

$\Phi_1$  and  $\Phi_2$  are maximal proper state subformulae of  $\Phi$ :

$$\Phi = \underbrace{\exists \diamond \square a}_{\Phi_1} \wedge \exists \square (\circ b \wedge \underbrace{\diamond \neg \exists (a \mathbf{U} b)}_{\Phi_2})$$

- 1 calculate recursively the satisfaction sets  $Sat(\Phi_i)$
- 2 replace  $\Phi_i$  with the atomic proposition  $a_i$ ,  $i = 1, 2$

$$\Phi \rightsquigarrow a_1 \wedge \underbrace{\exists \square (\circ b \wedge \diamond a_2)}_{\text{LTL formula } \varphi} = a_1 \wedge \exists \varphi$$

- 3 compute  $Sat(\exists \varphi)$  via NBA  $\mathcal{A}$  for  $\varphi$  and nested DFS in  $TS \oplus \mathcal{A}$
- 4 return  $Sat(\Phi) = Sat(a_1 \wedge \exists \varphi) = Sat(a_1) \cap Sat(\exists \varphi)$

## 3.7 CTL\* Model Checking

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**Algorithm 3:** CTL\* model checking algorithm (basic idea)

---

**Input:**  $TS$  with initial states  $I$ , and CTL\* formula  $\Phi$

**Output:**  $I \subseteq Sat(\Phi)$

---

```
1 foreach  $i \leq |\Phi|$  do
2   foreach  $\Psi \in Sub(\Phi)$  with  $|\Psi| = i$  do
3     switch  $\Psi$  do
4       case true do  $Sat(\Psi) := S$ ;
5       case  $a$  do  $Sat(\Psi) := \{s \in S \mid a \in L(s)\}$ ;
6       case  $a_1 \wedge a_2$  do  $Sat(\Psi) := Sat(a_1) \cap Sat(a_2)$ ;
7       case  $\neg a$  do  $Sat(\Psi) := S \setminus Sat(a)$ ;
8       case  $\exists \varphi$  do  $Sat(\Psi) := S \setminus Sat_{LTL}(\neg \varphi)$ ;
9     end
10     $AP := AP \cup \{a_\Psi\}$ ;
11    replace  $\Psi$  with  $a_\Psi$ ;
12    foreach  $s \in Sat(\Psi)$  do  $L(s) := L(s) \cup \{a_\Psi\}$ ;
13  end
14 end
15 return  $I \subseteq Sat(\Phi)$ ;
```

---

## 3.7 CTL\* Model Checking

Evidently, the time complexity of the CTL\* model-checking algorithm is dominated by the LTL model-checking phases

### Theorem 11 (Time Complexity of CTL\* Model Checking)

*For transition system  $TS$  with  $N$  states and  $K$  transitions, and CTL\* formula  $\Phi$ , the CTL\* model-checking problem  $TS \models \Phi$  can be determined in time  $O((N + K) \cdot 2^{|\Phi|})$ .*

- CTL\* model checking can be solved by any LTL model-checking algorithm
- The CTL\* model-checking problem is PSPACE-complete

# Summary

- 1 CTL is a logic for formalizing properties over computation trees, i.e., the branching structure of the states
- 2 The expressivenesses of LTL and CTL are incomparable
- 3 The CTL model-checking problem can be solved by a recursive descent procedure over the parse tree of the state formula to be checked
- 4 The time complexity of the CTL model-checking algorithm is linear in the size of the transition system and the length of the formula
- 5 The CTL model-checking procedure can be realized symbolically by means of ordered binary decision diagrams
- 6 CTL\* is more expressive than either CTL and LTL
- 7 The CTL\* model-checking problem can be solved by an appropriate combination of the recursive descent procedure (as for CTL) and the LTL model-checking algorithm, PSPACE-complete