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### Strong preimages Formal definition

### Strong preimage

Strong preimages

The strong preimage of a set T of states with respect to an operator ois the set of those states from which a state in T is always reached when executing o.



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Definition (Strong preimage of a set of states)  $spreimg_o(T) = \{s \in S | s' \in T, sos', img_o(s) \subseteq T\}$ 

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### Algorithms for fully observable problems

1. Heuristic search (forward)

Nondeterministic planning can be viewed as AND-OR search.

OR nodes: Choice between operators

AND nodes: Nondeterministically reached state

Heuristic AND-OR search algorithms: AO\*, ...

2. Dynamic programming (backward)

Idea Compute operator/distance/value for a state based on the operators/distances/values of its all successor states.

- 2.1 0 actions needed for goal states.
- 2.2 If states with i actions to goals are known, states with  $\leq i+1$ actions to goals can be easily identified.

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Automatic reuse of already found plan suffixes.

Algorithms Dynamic programming

### Dynamic programming

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### Planning by dynamic programming

If for all successors of state s with respect to operator o a plan exists, assign operator o to s.

Base case i = 0: In goal states there is nothing to do.

Inductive case  $i \ge 1$ : If there is  $o \in O$  such that for all  $s' \in img_o(s) s'$  is a goal state or  $\pi(s')$  was assigned on iteration i-1, then assign  $\pi(s) = o$ .

Connection to distances

If s is assigned a value on iteration  $i \ge 1$ , then the backward distance of s is i.

The dynamic programming algorithm essentially computes the backward distances of states.

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Algorithms Backward distances

### **Backward distances** Definition of distance sets

### Definition

Let G be a set of states and O a set of operators. Define the backward distance sets  $D_i^{bwd}$  for G, O that consist of those states for which there is a guarantee of reaching a state in G with at most i operator applications.

$$\begin{array}{lll} D_0^{\textit{bwd}} &= G \\ D_i^{\textit{bwd}} &= D_{i-1}^{\textit{bwd}} \cup \bigcup_{o \in O} \textit{spreimg}_o(D_{i-1}^{\textit{bwd}}) \textit{ for all } i \geq 1 \end{array}$$

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### **Backward distances** Example



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Algorithms Backward distances

### **Backward distances** Definition

Definition

Let G be as set of states and O a set of operators, and let  $D_0^{\textit{bwd}}, D_1^{\textit{bwd}}, \dots$  be the backward distance sets for G and O. Then the backward distance from a state s to G is

$$\delta_G^{\textit{bwd}}(s) = \begin{cases} 0 \text{ if } s \in G\\ i \text{ if } s \in D_i^{\textit{bwd}} \backslash D_{i-1}^{\textit{bwd}} \end{cases}$$

If  $s \notin D_i^{bwd}$  for all  $i \ge 0$  then  $\delta_C^{bwd}(s) = \infty$ .

### Algorithms Backward distances

1. Let  $S' \subseteq S$  be those states having a finite backward distance.

Assign to  $\pi(s)$  any operator  $o \in O$  such that  $img_o(s) \subseteq D_{i-1}^{bwd}$ .

Hence o decreases the backward distance by at least one.

The plan  $\pi$  solves the planning problem for  $\langle S, I, O, G, P \rangle$  iff  $I \subseteq S'$ .

### Construction of a plan based on distances

2. Let s be a state with distance  $i = \delta_G^{bwd}(s) \ge 1$ .

Extraction of a plan from distance sets

3.

### Making the algorithm a logic-based algorithm

- An algorithm that represents the states explicitly is feasible for transition systems with at most 10<sup>6</sup> or 10<sup>7</sup> states.
- For planning with bigger transition systems structural properties of the transition system have to be taken advantage of.
- Representing state sets as propositional formulae often allow taking advantage of the structural properties: a formula that represents a set of states or a transition relation that has certain regularities may be very small in comparison to the set or relation.



### Regression for nondeterministic operators Illustration

$$\operatorname{regr}_{(c,(e_1|e_2))}(\phi) = \operatorname{regr}_{(c,e_1)}(\phi) \wedge \operatorname{regr}_{(c,e_2)}(\phi)$$

Regression Definition



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AI Planning Regression Definition

### Regression for nondeterministic operators Example

### Example

Let  $o = \langle d, (b | \neg c) \rangle$ . Then

$$\begin{aligned} \operatorname{regr}_{o}^{nd}(b\leftrightarrow c) &= \operatorname{regr}_{(d,b)}(b\leftrightarrow c) \wedge \operatorname{regr}_{(d,\neg c)}(b\leftrightarrow c) \\ &= (d\wedge (\top\leftrightarrow c)) \wedge (d\wedge (b\leftrightarrow \bot)) \\ &\equiv d\wedge c \wedge \neg b. \end{aligned}$$

Regression for nondeterministic operators Correctness

### Theorem

Let  $\phi$  be a formula over A, o an operator over A, and S the set of all states over A. Then  $\{s \in S | s \models \operatorname{regr}_{o}^{nd}(\phi)\} = \operatorname{spreimg}_{o}(\{s \in S | s \models \phi\}).$ 

Regression Definition

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Proof.
Let o = \langle c, (e_1 | \cdots | e_n) \rangle.
  \{s \in S | s \models regr_o^{nd}(\phi)\}
  = \{s \in S | s \models \textit{regr}_{\langle c, e_1 \rangle}(\phi) \land \dots \land \textit{regr}_{\langle c, e_n \rangle}(\phi) \}
  = \{s \in S | s \models \textit{regr}_{\langle c, e_1 \rangle}(\phi), \dots, s \models \textit{regr}_{\langle c, e_n \rangle}(\phi)\}
  =\{s\in S|\textit{app}_{\langle c,e_1\rangle}(s)\models\phi,\ldots,\textit{app}_{\langle c,e_n\rangle}(s)\models\phi\}
  = \{s \in S | s' \models \phi \text{ for all } s' \in img_o(s), \text{ there is } s' \models \phi \text{ with } sos'\}
   = spreimg<sub>o</sub>({s \in S | s \models \phi})
3rd = is by properties of deterministic regression.
                                                                                                                                               4th = is by img_o(s) = \{app_{\langle c,e_1 \rangle}(s), \dots, app_{\langle c,e_n \rangle}(s)\}.
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### sion Definition

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### Backward distances with formulas

By using regression we can compute formulas that represent backward distance sets.

### Definition

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Let G be a formula and O a set of operators. The backward distance sets  $D_i^{bwd}$  for G, O are represented by the following formulae.

$$\begin{array}{l} D_0^{\textit{bwd}} = G \\ D_i^{\textit{bwd}} = D_{i-1}^{\textit{bwd}} \vee \bigvee_{o \in O} \textit{regr}_o^{\textit{nd}}(D_{i-1}^{\textit{bwd}}) \text{ for all } i \geq 1 \end{array}$$

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### Regression Definition

### Backward distances with formulas

### General images and preimages with formulas

Definition

Let *G* be a formula and *O* a set of operators, and let  $D_0^{bwd}$ ,  $D_1^{bwd}$ ,... be the formulae representing the backward distance sets for *G* and *O*. Then the backward distance from a state *s* to *G* is

$$\delta_G^{\textit{bwd}}(s) = \begin{cases} 0 \text{ if } s \models G\\ i \text{ if } s \models D_i^{\textit{bwd}} \land \neg D_{i-1}^{\textit{bwd}} \end{cases}$$

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If  $s \not\models D_i^{bwd}$  for all  $i \ge 0$  then  $\delta_G^{bwd}(s) = \infty$ .

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Operators in CPC

### General images and preimages with formulas

### Definition

Define the set of state variables possibly changed by e as

 $\begin{array}{l} \textit{changes}(a) = \{a\} \\ \textit{changes}(\neg a) = \{a\} \\ \textit{changes}(c \triangleright e) = \textit{changes}(e) \\ \textit{changes}(e_1 \land \dots \land e_n) = \textit{changes}(e_1) \cup \dots \cup \textit{changes}(e_n) \\ \textit{changes}(e_1| \dots | e_n) = \textit{changes}(e_1) \cup \dots \cup \textit{changes}(e_n) \\ \end{array}$ 

### Assumption

Let  $e_1 \land \dots \land e_n$  occur in the effect of an operator. If  $e_1, \dots, e_n$  are not all deterministic then a and  $\neg a$  may occur as an atomic effect in at most one of  $e_1, \dots, e_n$ .

This assumption rules out effects like  $(a|b) \land (\neg a|c)$  that may make a simultaneously true and false.

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### General images and preimages with formulas

### Example

We translate the effect

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$$e = (a|(d \rhd a)) \land (c|d)$$

into a propositional formula. The set of state variables is  $A = \{a, b, c, d\}.$ 

$$\begin{aligned} \tau^{\textit{nd}}_{\{a,b,c,d\}}(e) &= \tau^{\textit{nd}}_{\{a,b\}}(a|(d \rhd a)) \wedge \tau^{\textit{nd}}_{\{c,d\}}(c|d) \\ &= (\tau^{\textit{nd}}_{\{a,b\}}(a) \vee \tau^{\textit{nd}}_{\{a,b\}}(d \rhd a)) \wedge (\tau^{\textit{nd}}_{\{c,d\}}(c) \vee \tau^{\textit{nd}}_{\{c,d\}}(d)) \\ &= ((a' \wedge (b \leftrightarrow b')) \vee (((a \lor d) \leftrightarrow a') \wedge (b \leftrightarrow b'))) \wedge \\ &\quad ((c' \wedge (d \leftrightarrow d')) \vee ((c \leftrightarrow c') \wedge d')) \end{aligned}$$

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### Images in CPC ∃/∀-abstraction

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### Existential and universal abstraction

The most important operations performed on transition relations represented as propositional formulae are based on existential abstraction and universal abstraction.

### Definition

Existential abstraction of a formula  $\phi$  with respect to  $a \in A$ :

$$\exists a.\phi = \phi[\top/a] \lor \phi[\bot/a].$$

Universal abstraction is defined analogously by using conjunction instead of disjunction.

Definition

Universal abstraction of a formula  $\phi$  with respect to  $a \in A$ :

$$\forall a.\phi = \phi[\top/a] \land \phi[\perp/a].$$

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 The definition of regression covers only a subclass of nondeterministic operators.

- How to define strong preimages for all operators, and images and preimages?
- Now we apply a general idea:

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- 1. View operators/actions as binary relations.
- 2. Represent these binary relations as formulae.
- 3. Define relational operations for relations represented as formulae.

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Operators in CPC General images and preimages with formulas

In nondeterministic choices  $e_1 | \cdots | e_n$  the formula for each  $e_i$  has to express the changes for exactly the same set B of state variables.

Definition  

$$\begin{aligned} \tau_B^{nd}(e) &= \tau_B(e) \text{ when } e \text{ is deterministic} \\ \tau_B^{nd}(e_1|\cdots|e_n) &= \tau_B^{nd}(e_1) \vee \cdots \vee \tau_B^{nd}(e_n) \\ \tau_B^{nd}(e_1 \wedge \cdots \wedge e_n) &= \tau_{B(l_2 \cup \cdots \cup B_n)}^{nd}(e_1) \wedge \tau_{B_2}^{nd}(e_2) \wedge \cdots \wedge \tau_{B_n}^{nd}(e_n) \\ \text{ where } B_i &= \text{changes}(e_i) \text{ for } i \in \{2, \dots, n\} \end{aligned}$$

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### General images and preimages with formulas

Definition

Let *A* be a set of state variables. Let  $o = \langle c, e \rangle$  be an operator over *A* in normal form. Define  $\tau_A^{nd}(o) = c \wedge \tau_A^{nd}(e)$ .

Lemma Let o be an operator. Then

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$$\{v|v \text{ is a valuation of } A \cup A', v \models \tau_A^{nd}(o)\} \\ = \{s \cup s'[A'/A]|s, s' \in S, s' \in img_o(s)\}.$$

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# Images in CPC 3/V-abstraction

∃-abstraction

### Examples Example

 $\exists b.((a \to b) \land (b \to c)) \\ = ((a \to \top) \land (\top \to c)) \lor ((a \to \bot) \land (\bot \to c)) \\ \equiv c \lor \neg a \\ \equiv a \to c$ 

$$\begin{aligned} \exists ab.(a \lor b) &= \exists b.(\top \lor b) \lor (\bot \lor b) \\ &= ((\top \lor \top) \lor (\bot \lor \top)) \lor ((\top \lor \bot) \lor (\bot \lor \bot)) \\ &= (\top \lor \top) \lor (\top \lor \bot) = \top \end{aligned}$$

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Example

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∃-abstraction is also known as forgetting:  $\exists mon\exists tue((mon \lor tue) \land (mon → work) \land (tue → work)))$  $\equiv \exists tue((work \land (tue → work)) \lor (tue \land (tue → work)))) \equiv work$ 

# ∀ and ∃-abstraction in terms of truth-tables

# $\forall a$ and $\exists a$ correspond to combining pairs of lines with the same valuation for variables other than a.

### Example

$\exists c.(a \lor (b \land c)) \equiv a \lor b  \forall c.(a \lor (b \land c)) \equiv a$								
a         b         c           0         0         0           0         1         0           0         1         1           1         0         0           1         1         0           1         1         0           1         1         1           (Albert-Ludw)	$\frac{a \lor (b \land c)}{0}$ 0 1 1 1 1 igs-Universität Freib	a b ∃ 0 0 0 1 1 0 1 1	<u>c.(a ∨ (b ∧ c))</u> 0 1 1 1 1 1	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\frac{a \lor (b \land 0)}{0}$	. <u>c))</u> May 30, 2005	33 / 56	

### Properties of abstraction operations

# Definition Existential and universal abstraction of $\phi$ with respect to a set of atomic propositions $B = \{b_1, \ldots, b_n\}$ are $\exists B. \phi = \exists b_1. (\exists b_2. (\ldots \exists b_n. \phi \ldots))$ $\exists B. \phi = \forall b_1. (\forall b_2. (\ldots \forall b_n. \phi \ldots))$ $\forall B. \phi = \forall b_1. (\forall b_2. (\ldots \forall b_n. \phi \ldots))$ (Abert-Ludwigs-Universität Freiburg) A Paning Mey 0.202 34/50

Properties of  $\forall$  and  $\exists$  abstraction

1. Let  $\phi$  be a formula over A. Then  $\exists A.\phi$  and  $\forall A.\phi$  are formulae that consist of the constants  $\top$  and  $\bot$  and the logical connectives only.

Images in CPC ∃/∀-abstraction

- 2. The truth-values of these formulae are independent of the valuation of A, that is, their values are the same for all valuations.
- 3.  $\exists A.\phi \equiv \top$  if and only if  $\phi$  is satisfiable.

Properties of abstracted formulas

4.  $\forall A.\phi \equiv \top$  if and only if  $\phi$  is valid.

Lemma If $\phi$ is a formula over $A \cup A'$ and $v$ a valuation of $A$ then
1. $v \models \exists A'.\phi$ iff $v \cup v' \models \phi$ for some valuation $v'$ of $A'$ .
<b>2</b> . $v \models \forall A'.\phi$ iff $v \cup v' \models \phi$ for all valuations $v'$ of $A'$ .

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•  $\phi_2$  a formula on  $A \cup A'$  representing a matrix  $M_{2^n \times 2^n}$  (equivalently,

 $\exists A.(\phi_1 \land \phi_2)$ 

 $(\exists A.(\phi_1 \land \phi_2))[A/A']$ 

•  $\phi_1$  be a formula on A representing a row vector  $V_{1 \times 2^n}$ 

(equivalently, a set of valuations of A), and

The product matrix VM of size  $1 \times 2^n$  is represented by

To obtain a formula over A we have to rename the variables.

a binary relation on valuations of A).

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### Size of abstracted formulae

- Abstracting one variable takes polynomial time in the size of the formula.
- Abstracting one variable may double the formula size.
- Abstracting n variables may increase size by factor  $2^n$ .
- For making abstraction practical the formulae must be simplified, for example with equivalences like ⊤ ∧ φ ≡ φ, ⊥ ∧ φ ≡ ⊥, ⊤ ∨ φ ≡ ⊤, ⊥ ∨ φ ≡ φ, ¬⊥ ≡ ⊤, and ¬⊤ ≡ ⊥.

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### Images by ∃-abstraction

### Example

Let  $A=\{a,b\}$  be the state variables.

$$(1\ 0\ 1\ 0\)\times \begin{pmatrix} 0\ 1\ 0\ 1\\ 1\ 0\ 1\ 0\\ 0\ 1\ 0\ 1\\ 1\ 0\ 1\ 0 \end{pmatrix} = (\ 0\ 1\ 0\ 1)$$

represents the image of  $\{00,10\}$  with respect to a relation.

$$\begin{array}{l} \exists a. \exists b. (\neg b \land (b \leftrightarrow \neg b')) \\ \equiv \exists b. (\neg b \land (b \leftrightarrow \neg b')) \\ \equiv (\neg \top \land (\top \leftrightarrow \neg b')) \lor (\neg \bot \land (\bot \leftrightarrow \neg b')) \\ \equiv b' \end{array}$$

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The formula b represents  $\{01, 11\}$ .

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which is a formula on A'.

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Let

Images by ∃-abstraction

•  $A = \{a_1, \dots, a_n\},$ •  $A' = \{a'_1, \dots, a'_n\},$ 

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### Matrix multiplication by ∃-abstraction

Let

- $\blacktriangleright A = \{a_1, \ldots, a_n\},\$
- $A' = \{a'_1, \dots, a'_n\},\$
- $A'' = \{a''_1, \dots, a''_n\},\$
- $\phi_1$  be a formula on  $A \cup A'$  representing matrix  $M_1$  and

▶  $\phi_2$  a formula on  $A' \cup A''$  representing matrix  $M_2$ .

The matrices  $M_1$  and  $M_2$  have size  $2^n \times 2^n$ .

The product matrix  $M_1M_2$  is represented by

$$\exists A'.(\phi_1 \land \phi_2)$$

which is a formula on  $A \cup A''$ .

### Images in CPC ∃/∀-abstraction

Let  $\phi_1 = a \leftrightarrow \neg a'$  and  $\phi_2 = a' \leftrightarrow a''$  represent two actions, reversing the truth-value of a and doing nothing. The sequential composition of

 $\exists a'.\phi_1 \land \phi_2 = ((a \leftrightarrow \neg \top) \land (\top \leftrightarrow a'')) \lor ((a \leftrightarrow \neg \bot) \land (\bot \leftrightarrow a'')) \\ \equiv ((a \leftrightarrow \bot) \land (\top \leftrightarrow a'')) \lor ((a \leftrightarrow \top) \land (\bot \leftrightarrow a''))$ 

 $\equiv (\neg a \wedge a'') \lor (a \wedge \neg a'')$ 

 $\equiv a \leftrightarrow \neg a''$ 

### Matrix multiplication by ∃-abstraction Example

### Matrix multiplication

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Multiply (\neg a \leftrightarrow a') \land (\neg b \leftrightarrow b') and (a' \leftrightarrow b'') \land (b' \leftrightarrow a''):
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$$-\begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \times \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$

This is

$$\exists a'. \exists b'. (\neg a \leftrightarrow a') \land (\neg b \leftrightarrow b') \land (a' \leftrightarrow b'') \land (b' \leftrightarrow a'') \\ \equiv (\neg a \leftrightarrow b'') \land (\neg b \leftrightarrow a'').$$

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Example

these actions is

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## Images and preimages by formula manipulation

Define  $s[A'/A] = \{ \langle a', s(a) \rangle | a \in A \}.$ 

Lemma Let  $\phi$  be a formula on A and v a valuation of A. Then  $v \models \phi$  iff  $v[A'/A] \models \phi[A'/A].$ 

Definition Let o be an operator and  $\phi$  a formula. Define

> $img_o(\phi) = (\exists A.(\phi \land \tau_A^{nd}(o)))[A/A']$  $preimg_{o}(\phi) = \exists A'.(\tau_{A}^{nd}(o) \land \phi[A'/A])$   $preimg_{o}(\phi) = \exists A'.(\tau_{A}^{nd}(o) \land \phi[A'/A])$   $spreimg_{o}(\phi) = \forall A'.(\tau_{A}^{nd}(o) \rightarrow \phi[A'/A]) \land \exists A'.\tau_{A}^{nd}(o).$

> > Images in CPC Preimages

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### Preimages by formula manipulation

### Theorem

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Let  $T = \{s \in S | s \models \phi\}$ . Then  $\{s \in S | s \models \mathsf{preimg}_o(\phi)\} = \{s \in S | s \models$  $\exists A'.(\tau_A^{nd}(o) \land \phi[A'/A]) \} = preimg_o(T).$ Proof.  $s \models \exists A'.(\tau_A^{nd}(o) \land \phi[A'/A])$ iff there is  $s'_0 : A' \to \{0,1\}$  s.t.  $(s \cup s'_0) \models \tau_A^{nd}(o) \land \phi[A'/A]$ iff there is  $s'_0 : A' \to \{0,1\}$  s.t.  $s'_0 \models \phi[A'/A]$  and  $(s \cup s'_0) \models \tau_A^{nd}(o)$ iff there is  $s'_1 : A \to \{0,1\}$  s.t.  $s'_1 \models \phi$  and  $(s \cup s'_0) \models \tau_A^{nd}(o)$ If there is  $s' \in T$  s.t.  $s \cup s'[A'/A] \models \tau_A^{nd}(o)$ iff there is  $s' \in T$  s.t.  $s' \in img_o(s)$ iff there is  $s' \in T$  s.t.  $s \in preimg_o(s')$ iff  $s \in preimg_o(T)$ . Above we define  $s' = s'_0[A/A']$  (and hence  $s'_0 = s'[A'/A]$ .) 

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### Strong preimages vs. regression

### Corollary

Let  $o = \langle c, (e_1 | \cdots | e_n) \rangle$  be an operator such that all  $e_i$  are deterministic. The formula spreimg<sub>o</sub>( $\phi$ ) is logically equivalent to regr<sup>nd</sup><sub>o</sub>( $\phi$ ).

Proof.

 $\{s \in S | s \models \textit{regr}_o(\phi)\} = \textit{spreimg}_o(\{s \in S | s \models \phi\}) = \{s \in S | s \models \phi\}$ spreimg<sub>o</sub>( $\phi$ )}.

Images by formula manipulation

### Theorem

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Let  $T = \{s \in S | s \models \phi\}$ . Then  $\{s \in S | s \models img_o(\phi)\} = \{s \in S | s \models img_o(\phi)\}$  $(\exists A.(\phi \wedge \tau_A^{nd}(o)))[A/A']\} = img_o(T).$ 

### Proof.

 $s' \models (\exists A.(\phi \land \tau_A^{nd}(o)))[A/A']$ iff  $s'[A'/A] \models \exists A.(\phi \land \tau_A^{nd}(o))$ iff there is valuation s of A s.t.  $(s \cup s'[A'/A]) \models \phi \land \tau_A^{nd}(o)$ iff there is valuation s of A s.t.  $s \models \phi$  and  $(s \cup s'[A'/A]) \models \tau_A^{nd}(o)$ iff there is  $s \in T$  s.t.  $(s \cup s'[A'/A]) \models \tau_A^{nd}(o)$ iff there is  $s \in T$  s.t.  $s' \in img_o(s)$ iff  $s' \in img_o(T)$ .

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### Strong preimages by formula manipulation

Theorem Let  $T = \{s \in S | s \models \phi\}$ . Then  $\{s \in S | s \models \mathsf{spreimg}_o(\phi)\} = \{s \in S | s \models s\}$  $\forall A'.(\tau_A^{nd}(o) \to \phi[A'/A]) \land \exists A'.\tau_A^{nd}(o)\} = spreimg_o(T).$ Proof. See the lecture notes.

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Images in CPC Summary

### Summary of matrix/logic/relational operations

matrices	formulas	state sets
vector $V_{1 \times n}$	formula on A	set
matrix $M_{n \times n}$	formula on $A \cup A'$	relation
$V_{1 \times n} + V'_{1 \times n}$	$\phi_1 \lor \phi_2$	union
	$\phi_1 \wedge \phi_2$	intersection
$V_{1 \times n} \times M_{n \times n}$	$(\exists A.(\phi \land \tau_A^{nd}(o)))[A/A']$	$img_o(T)$
$M_{n \times n} \times V_{n \times 1}$	$\exists A'.(\tau_A^{nd}(o) \land \phi[A'/A])$	$preimg_o(T)$
	$\forall A'.(\tau_A^{nd}(o) \rightarrow \phi[A'/A]) \land \exists A'.\tau_A^{nd}(o)$	$spreimg_o(T)$

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### Images and preimages of sets of operators

The union of images of  $\phi$  with respect to all operators  $o \in O$  is

$$\bigvee_{o \in O} img_o(\phi)$$

This can be computed more directly by using the disjunction  $\bigvee_{o \in O} \tau_A(o)$  of the transition formulae:

$$\exists A.(\phi \land (\bigvee_{o \in O} \tau_A(o)))[A/A'].$$

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Same works for preimages.

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### Shannon expansion

### Definition

3-place connective if-then-else is defined by

$$\mathsf{ite}(a,\phi_1,\phi_2) = (a \land \phi_1) \lor (\neg a \land \phi_2)$$

where a is a proposition.

Definition Shannon expansion of a formula  $\phi$  with respect to  $a \in A$  is

$$\phi \equiv (a \land \phi[\top/a]) \lor (\neg a \land \phi[\perp/a]) = \mathsf{ite}(a, \phi[\top/a], \phi[\perp/a])$$

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### Binary decision diagrams Canonicity

### Transformation to ordered BDDs

- 1. Fix an ordering  $a_1, \ldots, a_n$  on all propositional variables.
- 2. Apply Shannon expansion to all variables in this order.
- 3. Represent the resulting formulae as directed acyclic graphs (DAG) so that shared subformulae occur only once.

Theorem Let  $\phi_1$  and  $\phi_2$  be two ordered BDDs obtained by using the same variable ordering. Then  $\phi_1 \equiv \phi_2$  if and only if  $\phi_1$  and  $\phi_2$  are isomorphic (the same DAG.)

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### Satisfiability algorithms vs. BDDs

Comparison: formula size, runtime technique $ s_{12}  = \alpha \mathcal{R}_{1}(P, P')  $ runtime for plan length $n$							
satisfiability	not a problem	exponential in $n$					
DDDe	maiar problem						
BDDS	major problem	less dependent on n					
Comparison: resource consumption technique   critical resource							
satisfiability	runtime						
BDDs	memory						
Comparison: application domain technique types of problems							
satisfiability	iols of state variables, short plans						
BDDs	few state variables, long plans						

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Images in CPC vs. SAT

### Image computation vs. planning by satisfiability

### We tested plan existence by testing satisfiability of

$$\iota^0 \wedge \mathcal{R}_1(A^0, A^1) \wedge \cdots \wedge \mathcal{R}_1(A^{t-1}, A^t) \wedge G^t$$

where 
$$\mathcal{R}_1(A, A') = \bigvee_{o \in O} \tau_A(o)$$
.  
 $\exists$ -abstracting  $A^0 \cup \cdots \cup A^{t-1}$  yields

$$\exists A^{t-1}.(\cdots \exists A^{0}.(\iota^{0} \land \mathcal{R}_{1}(A^{0}, A^{1})) \land \cdots \land \mathcal{R}_{1}(A^{t-1}, A^{t}) \land G^{t}).$$

This is equivalent to conjoining the t-fold image of t

$$\bigvee_{o \in O} img_o(\cdots \bigvee_{o \in O} img_o(\iota) \cdots)$$

with G to test goal reachability in t steps. We can do the same with preimages starting from G.

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### Binary decision diagrams Example

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By repeated application of Shannon expansion any propositional formula can be transformed to an equivalent formula containing no other connectives than ite and propositional variables only in the first position of ite.

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Example  $(a \lor b) \land (b \lor c)$  $\equiv ite(a, (\top \lor b) \land (b \lor c), (\bot \lor b) \land (b \lor c))$  $\equiv$  *ite*(*a*, *b*  $\lor$  *c*, *b*)  $\equiv ite(a, ite(b, \top \lor c, \bot \lor c), ite(b, \top, \bot))$  $\equiv$  *ite*(*a*, *ite*(*b*,  $\top$ , *c*), *ite*(*b*,  $\top$ ,  $\perp$ ))  $\equiv ite(a, ite(b, \top, ite(c, \top, \bot)), ite(b, \top, \bot))$ 

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### Binary decision diagrams: example



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### Properties of CPC normal forms

### Trade-offs between different CPC normal forms

Normal forms that allow faster reasoning are more expensive to construct from an arbitrary propositional formula and may be much bigger.

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Properties of different normal forms						
	$\vee$	$\wedge$	-	$\phi \in TAUT$ ?	$\phi \in SAT?$	$\phi \equiv \phi'$ ?
circuits	poly	poly	poly	co-NP-hard	NP-hard	co-NP-hard
formulae	poly	poly	poly	co-NP-hard	NP-hard	co-NP-hard
DNF	poly	exp	exp	co-NP-hard	in P	co-NP-hard
CNF	exp	poly	exp	in P	NP-hard	co-NP-hard
BDD	exp	exp	poly	in P	in P	in P

For BDDs one  $\vee/\wedge$  is polynomial time/size (size is doubled) but repeated  $\vee/\wedge$ lead to exponential size.