

# Certifying Unsolvability using CNF Formulas

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## STRIPS Planning Task

#### Definition

### A STRIPS planning task $\Pi$ is defined as $\Pi = \langle V^{\Pi}, A^{\Pi}, I^{\Pi}, G^{\Pi} \rangle$ where

- $> V^{\Pi}$  is a finite set of propositional variables
- $> A^{\Pi}$  is a finite set of actions
- $> I^{\Pi} \subseteq V^{\Pi}$  is the initial state
- $> G^{\mathsf{\Pi}} \subseteq V^{\mathsf{\Pi}}$  is the goal

A subset  $s \subseteq V^{\Pi}$  is called a **state** of  $\Pi$ . The set of all states of  $\Pi$  is denoted by  $S^{\Pi}$ .

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## Certifying Planning Systems



Introduction	Generation	Validation	Experiments	Comparison	Conclusion

## Certifying Planning Systems



#### Definition



Conclusion

## Inductive Certificates

### Definition



Conclusion

## Inductive Certificates

### Definition

### $> I^{\Pi} \in S$



### Definition

$$P I^{\Pi} \in S$$
$$P S \cap S_{G}^{\Pi} = \emptyset$$



#### "cannot be left"

#### Definition

 $I^{\Pi} \in S$   $S \cap S^{\Pi}_{G} = \emptyset$ S is inductive in  $\Pi$ 



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#### Definition

An **inductive certificate** for planning task  $\Pi$  is given by a set  $S \subseteq S^{\Pi}$  of states, such that

 $I^{\Pi} \in S$ 

$$S\cap S_G^{\Pi}=\emptyset$$

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#### Theorem

Planning task  $\Pi$  is unsolvable iff there exists an inductive certificate for  $\Pi$ 



# Conjunctive Normal Form (CNF)

A finite conjunction of clauses is a formula in conjunctive normal form (CNF).

$$arphi = \bigwedge \bigvee \mathit{lit}$$

> Widely studied and commonly used in Computer Science> Testing a CNF formula for satisfiability is NP-complete

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Why CNF?					

### > STRIPS problem descriptions are very close to propositional logic

- > e.g. state  $s = \{v, w\}$  over variables  $V^{\Pi} = \{q, v, w\}$  described by  $\varphi_s = v \land w \land \neg q$
- > SAT-solver allow certified verification

#### But: SAT-solving is NP-complete

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# Certifying Unsolvability using CNF Formulas



#### Formula $\varphi_{\mathcal{S}}$ should represent the set of reachable states $\mathcal{S}$

In blind search: all reachable states are expanded

- > Start with  $\varphi_S := \bot$
- > During search: append each expanded state s

$$\varphi_{S} = \varphi_{S} \lor (\bigwedge_{v \in s} v \land \bigwedge_{v \notin s} \neg v)$$

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$$\varphi_{S} = \bot$$

- I	ntroduction	Generation	Validation	Experiments	Comparison	Conclusion

$$\varphi_{\mathcal{S}} = (q \wedge v \wedge \neg w)$$



Introduction	Generation	Validation	Experiments	Comparison	Conclusion

$$\varphi_{S} = (q \land v \land \neg w) \\ \lor (w \land \neg q \land \neg v)$$



Introduction	Generation	Validation	Experiments	Comparison	Conclusion
Blind Searc	ch				

$$\varphi_{S} = (q \land v \land \neg w)$$
$$\lor (w \land \neg q \land \neg v)$$
$$\lor (v \land w \land \neg q)$$



Introduction	Generation	Validation	Experiments	Comparison	Conclusion
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$$\varphi_{S} = (q \land v \land \neg w)$$
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$$\lor (v \land w \land \neg q)$$
$$\lor (q \land \neg v \land \neg w)$$



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Blind Searc	:h				

$$arphi_S = (q \wedge v \wedge \neg w) \ arphi (w \wedge \neg q \wedge \neg v) \ arphi (v \wedge w \wedge \neg q) \ arphi (q \wedge \neg v \wedge \neg w)$$



 $\varphi_S$  describes the inductive certificate since  $\forall s \in S : s \models \varphi_S$ 





- > Infinite heuristic values may prune the search space
  - $\rightarrow$  We don't expand all reachable states  $\rightarrow$   $S_{exp}$  is not inductive

- > Assume we have an inductive set R<sub>sd</sub> for each dead-end sd
- Expanded states lead to expanded states and dead-ends





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How to regain inductivity?

- > Assume we have an inductive set  $R_{s_d}$  for each dead-end  $s_d$
- > Expanded states lead to expanded states and dead-ends
  - $\rightarrow S = S_{exp} \cup R_w$  is inductive



 $\varphi_V := \varphi_{init} \lor \varphi_{goal} \lor \varphi_{inductive}$  $\varphi_V$  is unsatisfiable iff subformulas are unsatisfiable

 $\varphi_{goal} := \varphi_G \land \varphi_S$  $= \bigwedge_{v \in G} v \land \bigvee_{s \in S} \varphi_s$ 



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Introduction	Generation	Validation	Experiments	Comparison	Conclusion
Validation F	ormula				

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 $arphi_{goal} := arphi_G \land arphi_S$ =  $\bigwedge_{v \in G} v \land \bigvee_{s \in S} arphi_s$ 

Introduction	Generation	Validation	Experiments	Comparison	Conclusion
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Validation F	ormula				

$$\begin{split} \varphi_V &:= \varphi_{init} \lor \varphi_{goal} \lor \varphi_{inductive} & \varphi_{goal} ::= \varphi_G \land \varphi_S \\ \varphi_V & \text{is unsatisfiable iff subformulas are} & = \bigwedge_{v \in G} v \land \bigvee_{s \in S} \varphi_s \end{split}$$

**However**:  $\varphi_V$  is not in CNF  $\rightarrow$  We cannot use SAT-solver on  $\varphi_V$ 



In our case:  $\varphi_G \land \bigvee_{s \in S} \varphi_s$  unsatisfiable iff  $\varphi_G \land \varphi_s$  unsatisfiable  $\forall s \in S$ 

Trivial SAT-calls, but many:

## $> \varphi_{init}$



## > $\varphi_{inductive}$

Introduction	Generation	Validation	Experiments	Comparison	Conclusion
Split up the	Formula				

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 $\downarrow_{DNF}$ 

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 $arphi_{m{g} m{o} al}$ 

arphi inductive

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arphi arphigoal

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 $> arphi_{ ext{init}} o 1$ 

 $> arphi_{goal} \rightarrow \#$ expanded states

 $> \varphi_{inductive} \rightarrow \#$ expanded states  $\times \#$ actions

Introdu	iction	Generation	Validation	Experiments	Comparison	Conclusion
		_				

**Idea:** Transform  $\varphi_V$  into ...

> . . . an equivalent CNF formula

> ... an equisatisfiable CNF formula

 $\varphi_{\mathcal{V}} := \varphi_{\textit{init}} \vee \varphi_{\textit{goal}} \vee \varphi_{\textit{inductive}}$ 

Intro	duction	Generation	Validation	Experiments	Comparison	Conclusion

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Use the transformed formula as input to SAT-solver

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Tseitin Enc	oding				

simple Tseitin Encoding: e.g.  $x \leftrightarrow (v \lor w)$ 

 $\Rightarrow (\neg x \lor v \lor w) \land (x \lor \neg v) \land (x \lor \neg w)$ 

ightarrow substitutes each variable pair

 $\rightarrow$  equisatisfiable CNF

generalized Tseitin Encoding: e.g.  $x \leftrightarrow (\bigvee_i v_i)$   $\Rightarrow (\neg x \lor \bigvee_i v_i) \land (\bigwedge_i (x \lor \neg v_i))$  $\rightarrow$  can substitute larger subformula at once

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# Transformation Comparison



Introduction	Generation	Validation	Experiments	Comparison	Conclusion
Experiment	S				

## Split up

$$ightarrow \mathsf{FD}^{Sp} \xrightarrow{\mathsf{Task} \ \& \ \varphi_S} \mathsf{Ver}^{Sp}$$

incremental SAT-solver

 $\begin{array}{c} \textbf{Transform} \\ \geq \mathsf{FD}^{BC} \xrightarrow{\mathsf{Circuit}} \mathsf{Trans}^{BC} \xrightarrow{\mathsf{CNF}} \mathsf{Ver}^{BC} \\ bc2cnf \end{array}$ 

 $ightarrow \mathsf{FD}^D \xrightarrow{\mathsf{CNF}} \mathsf{Ver}^D$ direct transformation

Comparison of *blind* and  $h^{max}$
Introduction	Generation	Validation	Experiments	Comparison	Conclusion

#### CNF Coverage: blind vs. $h^{max}$



Generation Verification Transformation Generation Verification Transformation

	Introduction	Generation	Validation	Experiments	Comparison	Conclusion
т	ima Campari					
_ I	ime Compari	ISON				





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#### Coverage: blind vs. $h^{max}$





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## Generation CNF vs. BDD



Introduction Generat	ion
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# Verification CNF vs. BDD



Introduction	Generation	Validation	Experiments	Comparison	Conclusion
Conclusion					

- > Inductive Certificates capture unsolvability
- > Splitting the SAT-calls avoids inefficiency of SAT
- > Tseitin Encoding allows equisatisfiable transformation to CNF
- > CNF representation of certificates is practically viable
- > Exponential scaling of SAT

### Questions?

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Introduction	Generation	Validation	Experiments	Comparison	Conclusion
Failures					

	blind	d	h <sup>max</sup>		
	memory	time	memory	time	
FD <sup>Sp</sup>	0	0	5	32	
FD <sup>BC</sup>	0	37	25	34	
Trans <sup>BC</sup>	96	1	77	3	
FD <sup>D</sup>	1	53	21	40	

Table: Reason for failures during generation in tasks where FD generated a certificate

Introduction	Generation	Validation	Experiments	Comparison	Conclusion
Failures					

	blind	d	h <sup>max</sup>		
	memory	time	memory	time	
VER <sup>Sp</sup>	144	4	121	7	
VER <sup>BC</sup>	4	0	10	0	
VER <sup>D</sup>	50	0	45	0	

Table: Reason for failures during verification in tasks where a certificate was generated

Introduction	Generation	Validation	Experiments	Comparison	Conclusion
Failures					

	blind		h <sup>max</sup>	
	memory	time	memory	time
FD <sup>D</sup>	1	53	21	40
$FD^{BDD}$	54	0	21	35
VER <sup>D</sup>	50	0	45	0
VER <sup>BDD</sup>	0	5	0	17

Table: Reasons for failure in tasks that FD solved