# OBDD-based Planning with Real Variables in a Non-Deterministic Environment

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# Background



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## **Action Languages**

 In general, action languages represent states (using fluents) and transitions (using actions)

Simple example in C where A is an action and P,Q are fluents. caused P if P after P, caused -P if-P after -P, caused Q if Q after Q, caused -Q if -Q after -Q, caused P if TRUE after Q^A.

- STRIPS -- (Fikes & Nilsson, 1971)
- A,B,C -- (Gelfond & Lifschitz, 1998)
  - **PDDL** -- emerging standard for action description

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### **Current Process**

Assume a blocks world with 3 blocks and portion of an action language description



 Action
 \*Moving a block B onto B1 means B is on B1 at next time step

 Language
 \*Moving a block B onto B1 means B is on B1 at next time step

 nonexecutable move(B,B1) if on(B2,B) && on(B3,B1)

 \*Moving a block B onto B1 is impossible if either B or B1 have

 another block on them

 on(a,a)₁ = move(a,a)₀ ∧¬on(a,a)₀ ∧¬on(b,a)₀ ∧¬on(c,a)₀

 on(a,b)₁ = move(a,b)₀ ∧¬on(a,a)₀ ∧¬on(b,a)₀ ∧¬on(c,a)₀

 $on(a,b)_{1} \equiv move(a,b)_{0} \land \neg on(a,a)_{0} \land \neg on(b,a)_{0} \land \neg on(c,a)_{0} \land \neg on(a,b)_{0} \land \neg on(b,b)_{0} \land \neg on(c,a)_{0} \land \neg on(a,b)_{0} \land \neg on(b,b)_{0} \land \neg on(c,b)_{0} \circ n(a,c)_{1} \equiv move(a,c)_{0} \land \neg on(a,a)_{0} \land \neg on(b,a)_{0} \land \neg on(c,a)_{0} \land \neg on(a,c)_{0} \land \neg on(c,c)_{0} \land \neg$ 

#### **Pass to SAT Checker**

length

## Satisfiability (SAT) Checkers

A variety of satisfiability checkers are available for planning problems:

- VIS -- (Brayton et al., 1996)
- SMV/NuSMV -- (Manzo, 1998)
- WalkSAT -- (Selman et al., 1994)

Question: How to apply satisfiability research efficiently in the causal planning domain in order to mitigate state space explosion and improve planning speed?

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# **Query Language Support**

Given a possible set of initial states and actions --

Query languages formulate a set of queries concerning the system's future state

- P,Q,R (Gelfond & Lifschitz, 1998) Query languages for the A,B,C set of action languages
- CTL (Computational Tree Logic) Widely used standard in satisfiability research and logic synthesis
- Various implementation specific query languages developed by individual researchers

## **Problems with State-of-the-Art**

### State Space Explosion

- Grounded representation size dependent on plan length, number of actions, number of fluents and number of possible parameters
- Instantiation of all plan times results in heavy performance penalty for replanning

### Query Languages

 Query languages vary between action languages; leading to confusion

### Satisfiability Checking

 Usage of CNF for state encoding produces slow satisfiability checking for large problems

# **Proposed Improvements**



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## **Proposed Theoretical Improvements**

### State Space Reduction

- Innovative use of new encodings facilitated by new satisfiability checkers
- Query Language Expressiveness
  - Use of standards from other fields (e.g. CTL)
- Encoding for Satisfiability Checking
  - BDD (Binary Decision Diagram)
  - Efficient compact representation of states provided by certain satisfiability tools

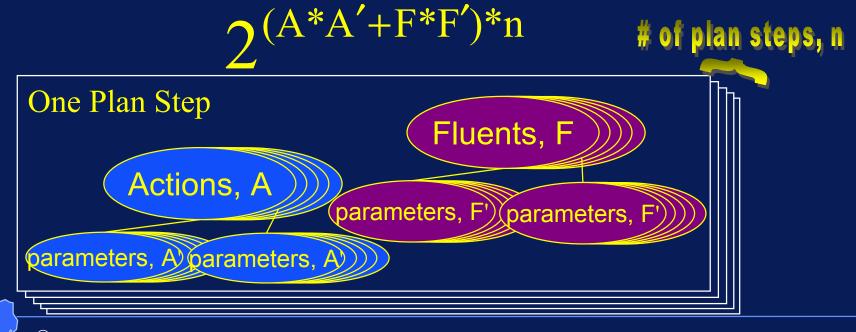


## **State Space Reduction (I)**

### Expected size:

\*

- A = # of actions at any given time
- A'= Average # of possible parameters on any action A
- F = # of fluent variables
- F'= Average # of parameters on any action F
- n = # of time steps in plan



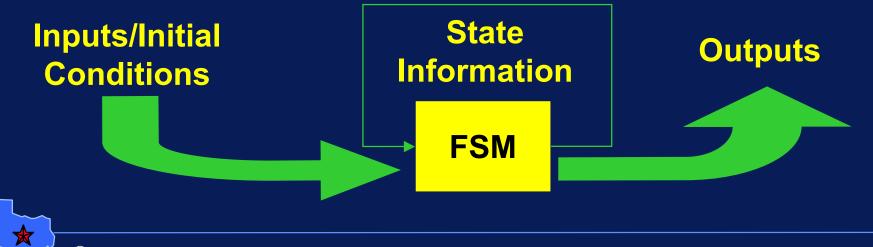
## **State Space Reduction (II)**

### Approach: State-based Encodings

- Reduce state space by using a Finite State Machine and calculating available next states.
- Dynamic environment = lots of replanning, current methods ground representation of unreached states

#### Impact:

- Reduces memory usage by only encoding current and next state
- Grounded state space size not related to plan length; results in a reduction by a factor of 2<sup>n</sup>



## **State Space Reduction(III)**

#### Most current tools:

- requires explicitly instantiation of each numerical parameter
- force relative boolean representations to describe absolute values.

### Approach: Parameterized Encoding

- does not require explicit instantiation
- allows direct representation of numerical values

#### Impact:

– State space reduction of 2<sup>A<sup>1</sup></sup>

at(x,y)	Encoding	Ground State	Comments
(2,2)	Explicit	at(2,0), at(2,1), at(2,2) at(1,0), at(1,1), at(1,2) at(0,0), at(0,1), at(0,2)	A total of 9 variables are needed.
(0,0)	Boolean	above(bottom), near(left), etc.	Absolute positioning is lost and all position is relative
	Parameter	at(int x, int y)	Preserves positioning and requires one variable; increases computation reqs.

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# **State Space Reduction (IV)**

### Intelligent branching - (Giungchiglia, et al. 1998)

 Many current SAT planners do not differentiate between fluents and actions when searching the state space.

### Approach:

- Note: Changes in fluents are the result of actions.
- Any fluents whose values can be deterministically chosen by action assignments can be pruned.

### Impact:

 Reduction of 2<sup>(F\*F')</sup> where F is a deterministically derived fluent value and F' is the average # of possible parameters.



### **Query Language Expressiveness**

### Approach:

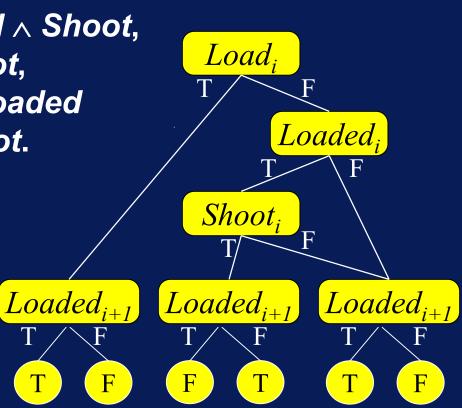
- Support for standard CTL syntax provides access to standard query representation without sacrificing expressiveness.
- CTL Syntax:
  - AF(x) x will be always eventually true (always finally)
  - AG(x) x is always true (always globally)
  - EF(x) it is possible for x to be true (eventually finally)
  - EG(x) it is possible for x to eventually always be true (eventually globally)

### Impact:

 Provides a common language-independent representation accepted by many existing tools

## **BDD - Binary Decision Diagram (I)**

interial Loaded, ¬Loaded, Alive, ¬Alive, caused Loaded after Load, caused ¬Alive after Loaded ∧ Shoot, caused ¬Loaded after Shoot, nonexecutable Shoot if ¬Loaded nonexecutable Load ∧ Shoot.





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# **BDD - Binary Decision Diagram (II)**

### Approach:

- BDDs supported by a variety of SAT checkers
- Provide an efficient and compact encoding of state

### Impact:

- Reduction in memory usage for representing grounded states
- Faster query language checking from SAT checkers
- Faster plan solutions from usage of SAT checkers

# **Current Implementation**



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## **Research Leveraging Existing Tools**

■ BLIF-MV → A logic file format that can be accepted by VIS.

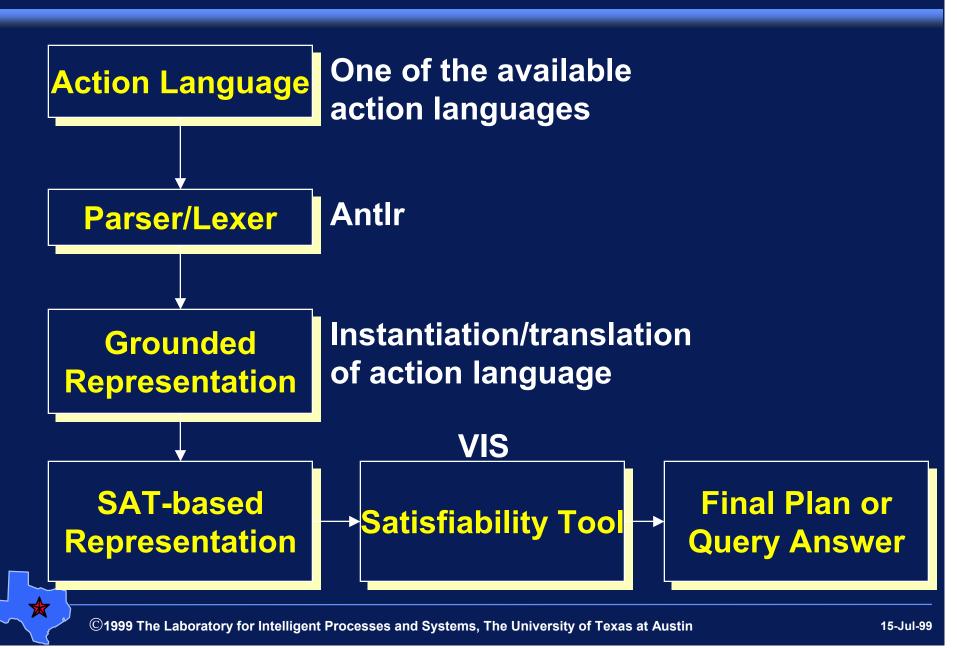
Antlr → A lex/yacc type parsing tool



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## Architecture



## **Current State of Research**

Causal Parser implementation is complete

- grounding and generation of SAT-based representation is being explored.
- Numerical value usage within a SAT checker is being explored.
- Speed/size testing against other planners remains to be done.



## Conclusions

SAT tools have been shown to perform efficiently when used for planning tasks.

Improvements are possible to:

- Enhance the language expressiveness
- Improve query utilization through standards usage

Usage of these techniques may reduce memory requirements and increase speed to plan solution

