

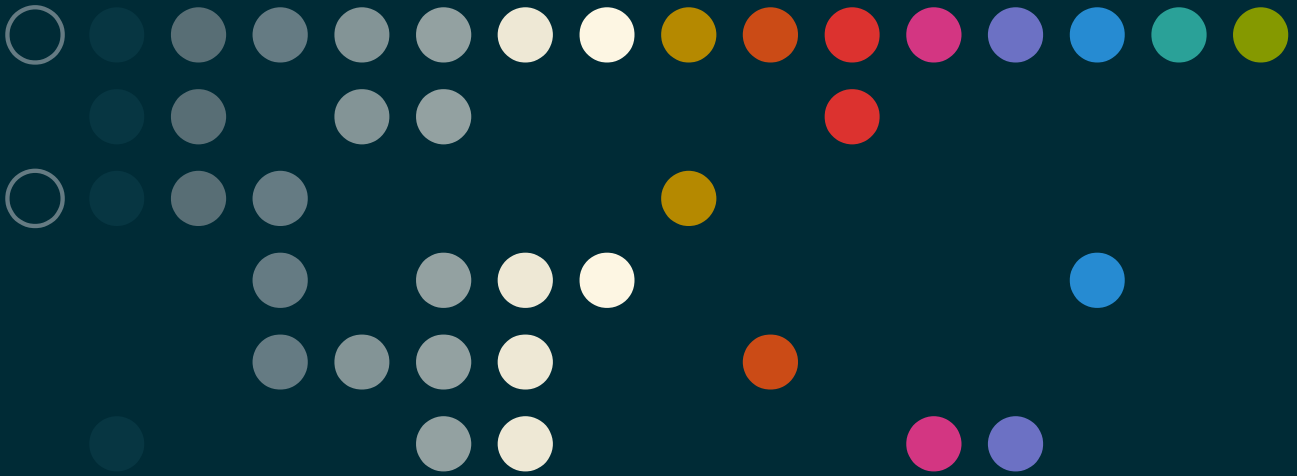
Présentation du jury

Bonjour à tous et merci d'assister à ma soutenance de thèse intitulée [titre].

Je suis Antoine Gréa, doctorant à l'université Lyon 1 et durant cette présentation je vais vous exposer mes diverses contributions.

1 Introduction

3



Tout d'abord on va présenter les base du problème

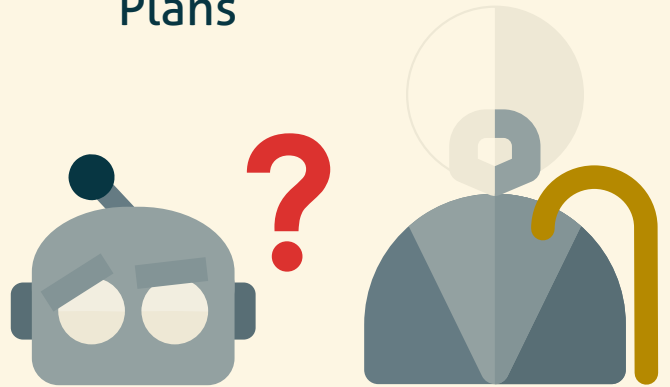
A what ?

4

- *Dependent people need help !*
 - Not **annoying** the person
 - Can't see *everything* they are doing
- How to help without asking ?
 - Guessing the intent somehow

- **Intent Recognition**

- Observed behavior → Goal
- Using action sequences: Plans



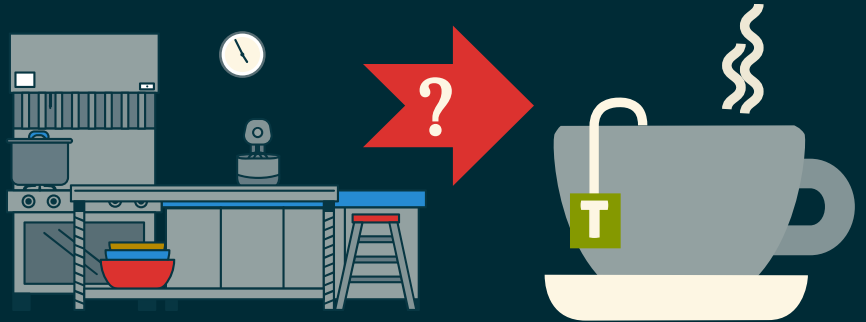
La reconnaissance d'intention c'est quoi ?

On ne doit pas déranger la personne et l'observation est toujours partielle.

Kitchen Example

5

- **Observation**
 - Bob goes in the kitchen
 - **Available goals**
 - Bob cleans the dishes
 - Bob makes tea
 - ...
 - **Infer correct one**
- **Issues**
 - Multiple goals
 - Interleaving
 - Partial Observation



Voici l'exemple utilisé tout du long de cette présentation.

Plan

6

- 1 Introduction
- 2 Intent Recognition
- 3 Knowledge Representation
- 4 General Planning
- 5 Flexible Online Planning
- 6 Conclusion

2 Intent Recognition

7



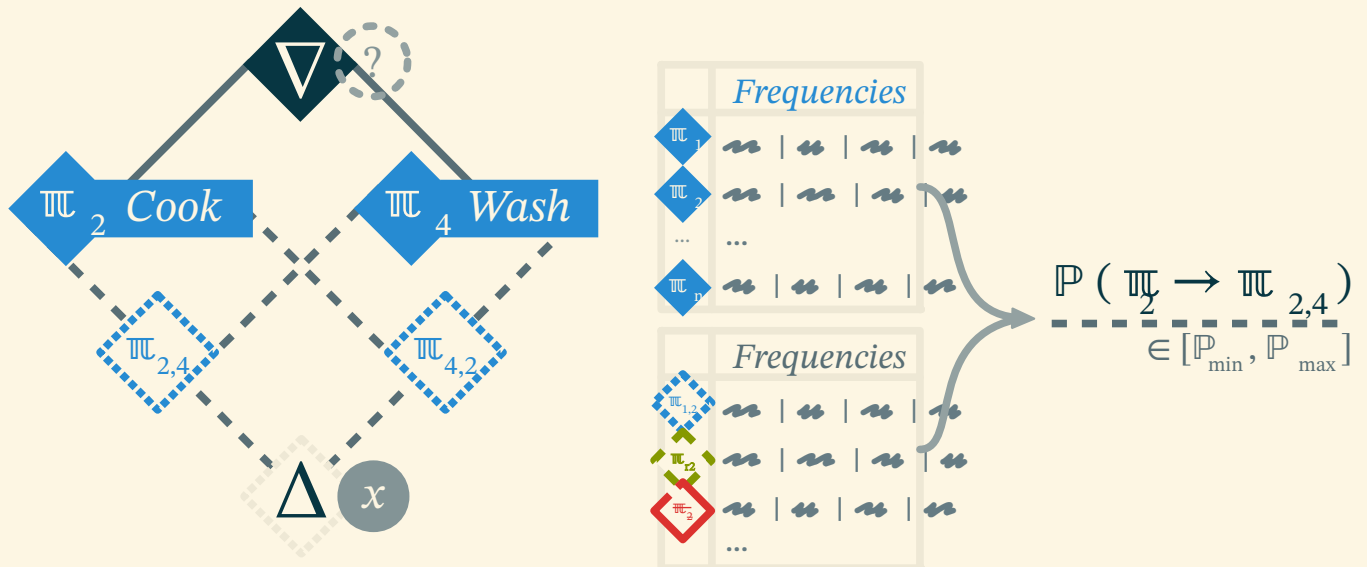
Comment faire de la reconnaissance d'intention

2.1 Logic Approach

[@BOUCHARD_2006]

8

- Lattice Based : ✓ Fast computations ✗ Exponential growth



Approche logique Bouchard
Treilli : objet mathématique ordonné

2.2 Stochastic Approach

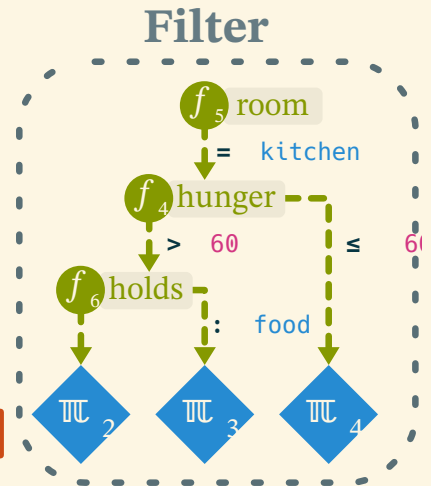
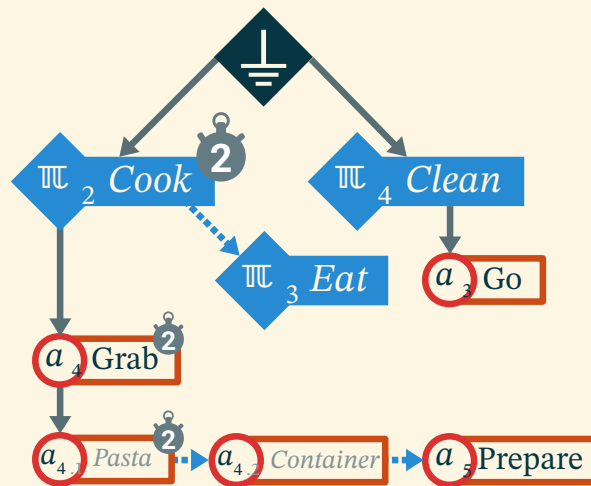
[@avrahami_2006]

9

- And/Or and decision tree :

✓ Accurate & efficient

✗ Handmade plan library & tree

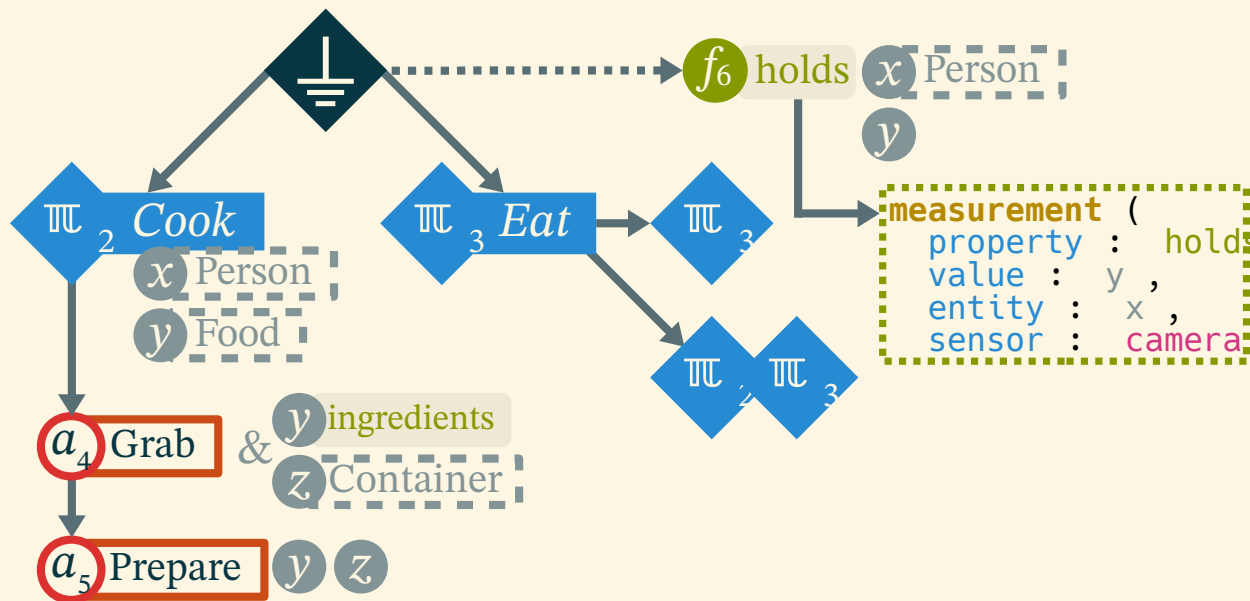


2.3 Grammatical Approach

[@VIDAL_2010]

10

- Valued Grammar : ✓ Versatile ✗ Slow refresh rate (~40s)

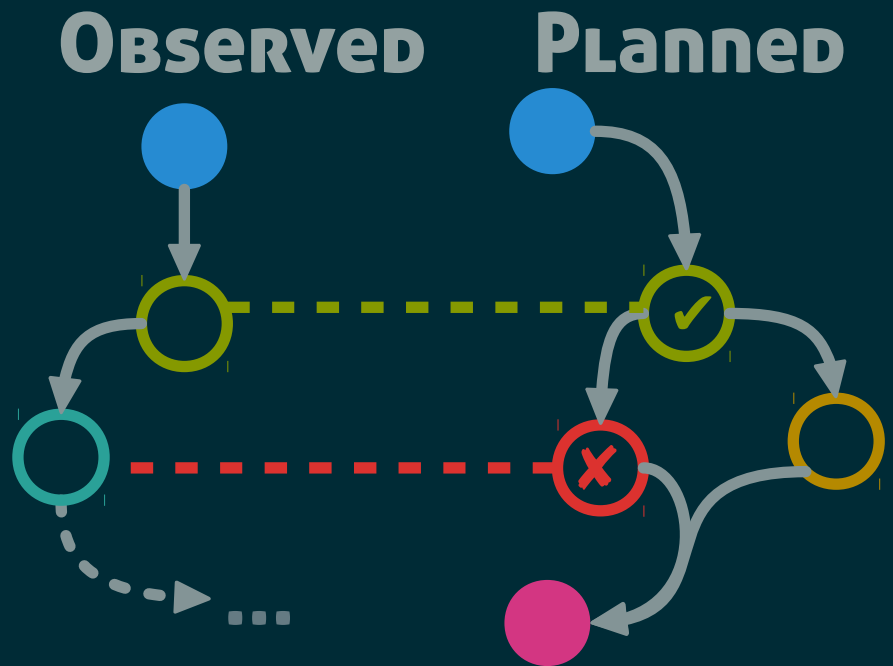


2.4 Inverted Planning

[@RAMIREZ_2008]

11

- **Intent Recognition**
 - Find the goal of a plan
- **Planning**
 - Find the plan to a goal
- *Theory of Mind*
 - The easier the plan, the more likely the goal



Planning: On a le but et on cherche le plan

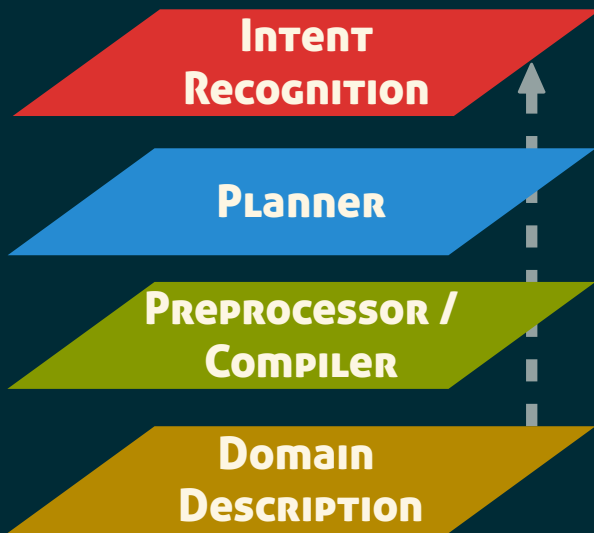
ToM: On se met à la place de l'autre

On observe un plan en cours d'exécution donc partiel

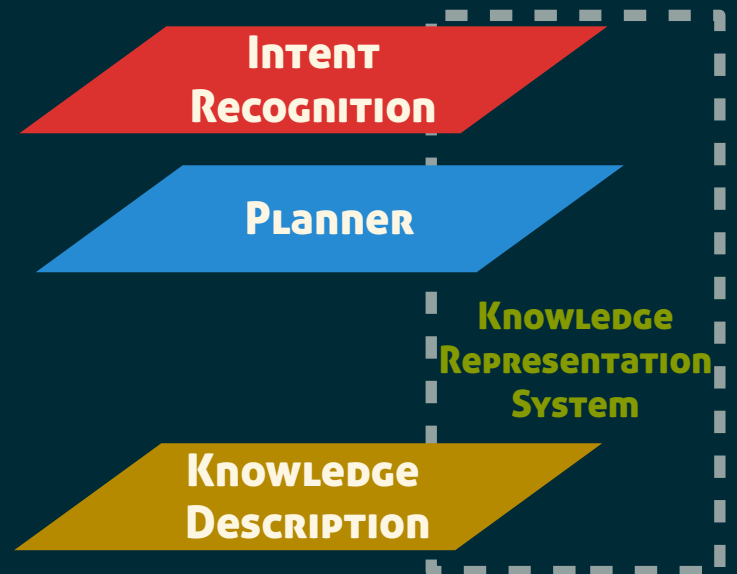
2.5 Framework Stacks

12

- Existing



- Contribution



1 Introduction | 2 Intent Recognition | 3 Knowledge Representation | 4 General Planning | 5 Flexible Online Planning | 6 Conclusion

Le système de connaissance est utilisé à tout les niveau et c'est le même donc on rends la chose plus flexible

3 Knowledge Representation

13

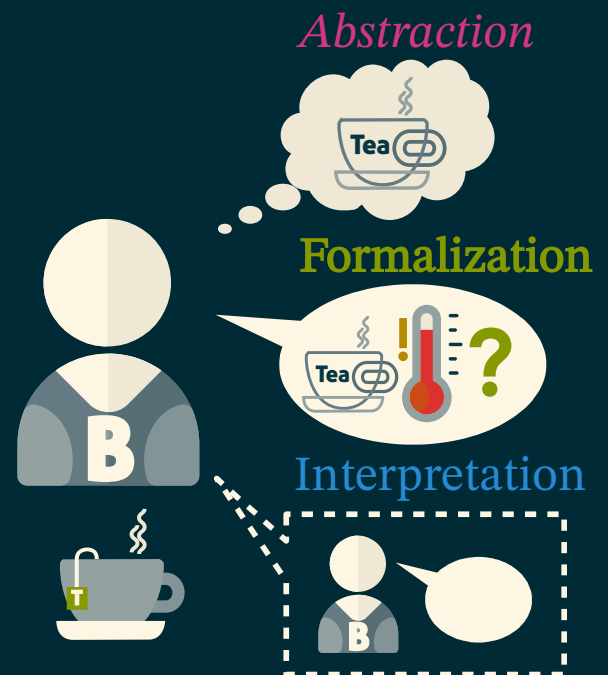
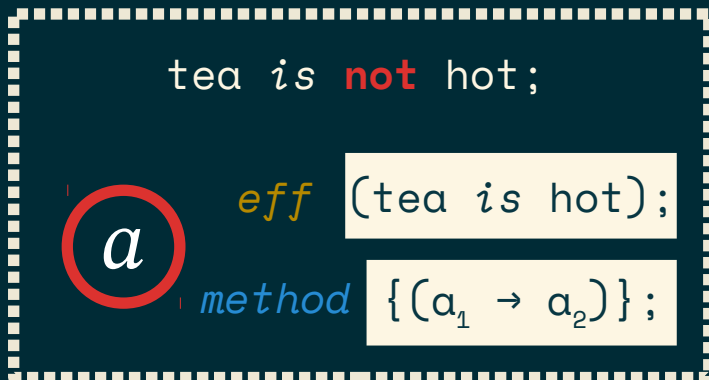


Knowledge in Planning

14

• Reification [camBRIDGE]

“The act of changing something abstract into something real



1 Introduction | 2 Intent Recognition | 3 Knowledge Representation | 4 General Planning | 5 Flexible Online Planning | 6 Conclusion

Abstraire c'est réduire une entité à ses caractéristiques élémentaire pour plus facilement y référer (perte d'info)

Formalization: Exprimer les connaissances en utilisant un langage sous-jacent (langage hôte)

Inter : Connaître les relation entre entité et ainsi leur définition (circularité)

Existing Tools

15

- **Ontologies**

- Based on Description Logic

```
<?xml version="1.0"?>

<RDF>
  <Description about="Bob">
    <likes>Tea</likes>
    <location>Kitchen</location>
  </Description>
</RDF>
```

- **Languages**

- RDF
- OWL-(Lite, DL, Full)
- ...

- **Issues**

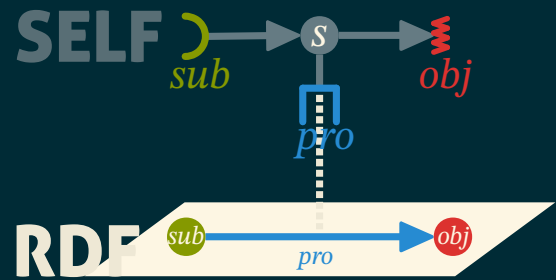
- Reification inefficient
- Higher order knowledge
- Flexibility of the structure

SELF

[@GRéa]

16

- Self defined
 - Structure = meaning
 - **Ex:** $*x = x$;
- More expressive
- Native reification
 - Express fluents and states in higher order spaces
 - Methods for hierarchical planning



Examples:

```
s = (bob @ kitchen);  
α pre s;  
α methods  
{go(kitchen) → take(cup)};
```


4 General Planning

17



Classical Planning

18

- Domain
 - **Fluents**
 - Formula over objects
 - **States**
 - Properties of the world
 - Formula over fluents
 - **Actions**
 - Precondition
 - Effects
- **Problem**
 - Initial state
 - Goal state
- **Plan (solution)**
 - Action sequence
 - Order
 - Total
 - Partial

Example

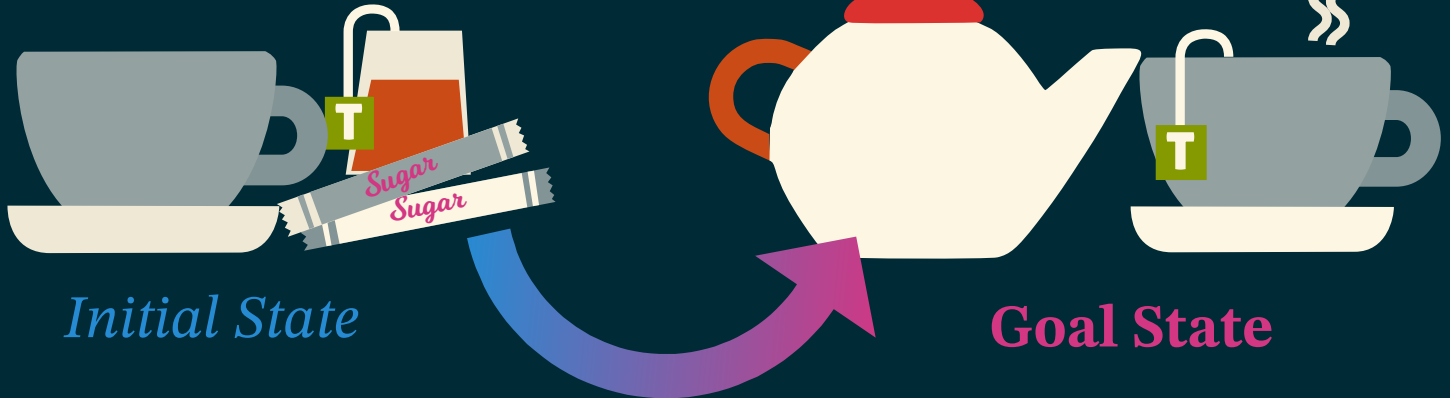
19

- **Fluents**

- thing *taken*
- *hot water, tea ready*

- **Actions**

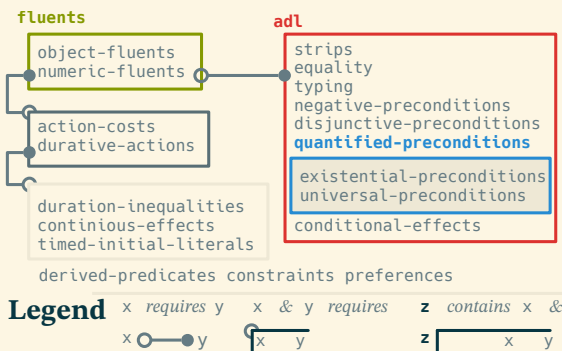
- *take, brew, boil, ...*



Existing Frameworks

20

- Standard language: **PDDL**
 - Numerous extensions to the language
 - Not used in probabilistic or hierarchical planning
 - Most of the time translated into an intermediate language for planners



- Temporal
 - PDDL+
 - ANML
- Probabilistic
 - PPDDL
 - **RDDL**
- Multi-Agent
 - MAPL
 - MA-PDDL
- Hierarchical
 - UMCP
 - SHOP2
 - **HDDL**
 - HPDDL
- Ontological
 - *WebPDDL*
 - *OPT*
- Hybrids
 - SIADEx

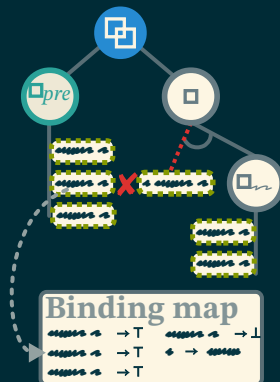
1 Introduction | 2 Intent Recognition | 3 Knowledge Representation | 4 General Planning | 5 Flexible Online Planning | 6 Conclusion

HDDL pour HTN Track à l'IPC et converti vers SHOP2 parfois.

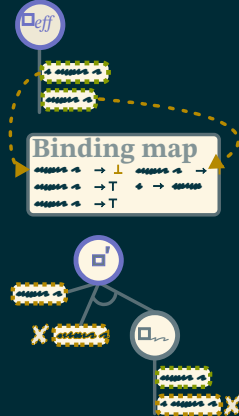
Planning Formalism Revisited [@GRéa] 21

- **States**
 - And/Or trees of **Fluents**
- **Actions**
 - **Precondition, Effects**
 - **Constraints**
 - **Cost, Duration, Probability**
 - **Methods** ($\text{eff} \rightarrow \text{pre}$)

Verifying



Applying

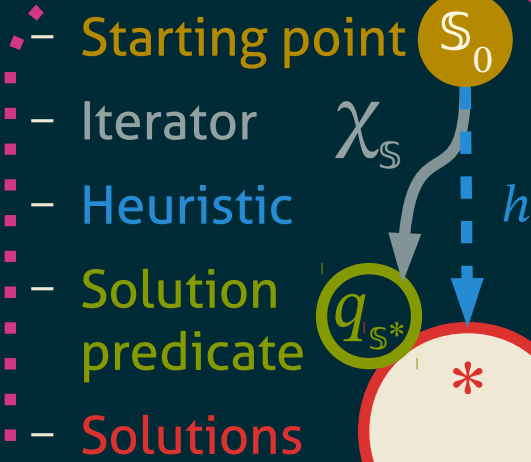


On construit un nouvel arbre avec les deux états et on vérifie que la valeur ne soit pas “faux”

Application : On construit la conjonction et on enlève tout ce qui est dans l'état courant qui n'est pas compatible en utilisant la carte des correspondances.

General Planning Framework [@gréa] 22

- **Search Space**

- Starting point s_0
 - Iterator χ_s
 - Heuristic h
 - Solution predicate q_{s^*}
 - Solutions $*$
- 

- Instances available for

- **State-transition** ↴

- Plan space

- Case based

- Probabilistic

- Hierarchical

$$\begin{aligned}
 s_0 &= a^0_{(\text{initial})} \\
 \chi &= a \in A_{(\text{actions})} \\
 q_* &= (=a^*)_{(\text{goal})} \\
 &= \square_{(\text{states})}
 \end{aligned}$$

COLOR Framework

[@GRéa]

23


```
(define (domain tea)
  (:requirements :equality :object-fluents)
  (:types container, liquid, item)
  (:constants no-item - item, water - liquid, cup - container)
  (:predicates (hot ?x - liquid))
  (:functions (taken) - item)

  (:action take
    :parameters (?x - item)
    :precondition (and (= (taken ?x) no-item))
    :effect (and (assign (taken) ?x)))
  (:action heat
    :parameters (?x - liquid)
    :precondition (and (not (hot ?x))
                       (= (taken ?x) ?x))
    :effect (and (hot ?x)))

  "planning.w" = ? ;
  take(item) pre (taken(~), ?(item));
  take(item) eff (taken(item));

  heat(thing) pre (~(hot(thing)), taken(thing));
  heat(thing) eff (hot(thing));

  make(drink) method (
    init(make(drink)) → take(spoon),
    take(spoon) → put(spoon),
    init(make(drink)) → infuse(drink,water,cup),
    infuse(drink,water,cup) → take(cup),
    take(cup) → put(cup),
    put(spoon) → goal(make(drink)),
    infuse(drink,water,cup) → goal(make(drink)),
    put(cup) → goal(make(drink))
  );
```



1 Introduction | 2 Intent Recognition | 3 Knowledge Representation | 4 General Planning | 5 Flexible Online Planning | 6 Conclusion

Pas de Plannification hierarchique en PDDL

5 Flexible Online Planning

24

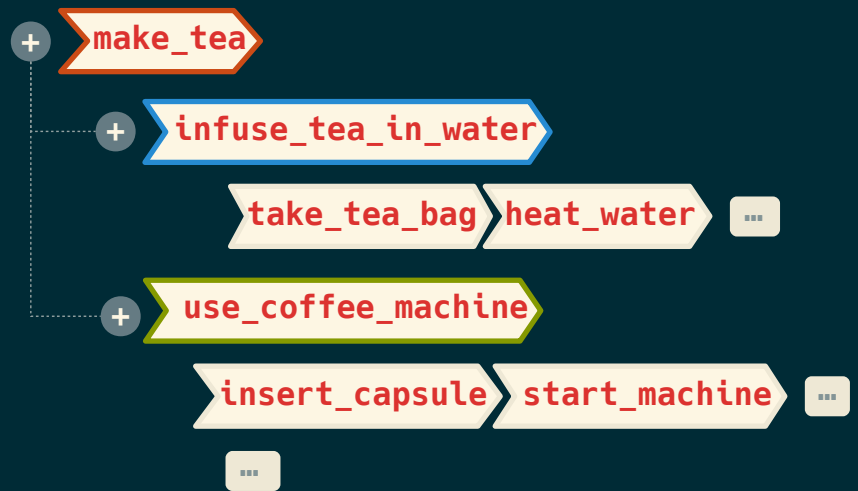


-
- The diagram illustrates a causal model for the blocks world domain. It consists of three main parts: a top causal graph, a bottom causal graph, and a legend.
- Top Causal Graph:** This graph shows a sequence of states a^0, a_1, a_2, a_3, a_4 leading to a goal state a^* . The nodes are represented by circles. The edges represent causal links, labeled f_1, f_2, f_3, f_4, f_5 . Some links are highlighted in red with a red arrow and a red cross, indicating a threat. Specifically, the link from a^0 to a_1 is red, and the link from a_2 to a_3 is red.
- Bottom Causal Graph:** This graph shows a similar sequence of states a^0, a_1, a_2, a_3, a_4 leading to a goal state a^* . The nodes are represented by circles. The edges represent causal links, labeled f_1, f_2, f_3, f_4, f_5 . Some links are highlighted in red with a red arrow and a red cross, indicating a subgoal. Specifically, the link from a^0 to a_1 is red, and the link from a_2 to a_3 is red.
- Legend:** The legend defines the symbols used in the graphs:
- a^0 : INITIAL
 - a : Step
 - a^* : GOAL
 - f : Causal link
 - f : Subgoal link

Hierarchical Task Networks

26

- Based on tasks decomposition
 - Replace task with method
- Lots of different approaches



Planning Phases

27

- Phases dependent on
 - Available information
 - Timing constraints
 - Planning paradigm



Domain 
compilation

Initialisation



Planning

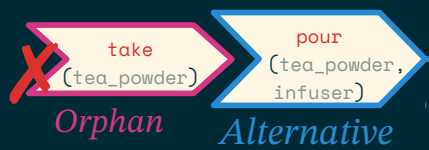


Solution
optimisation

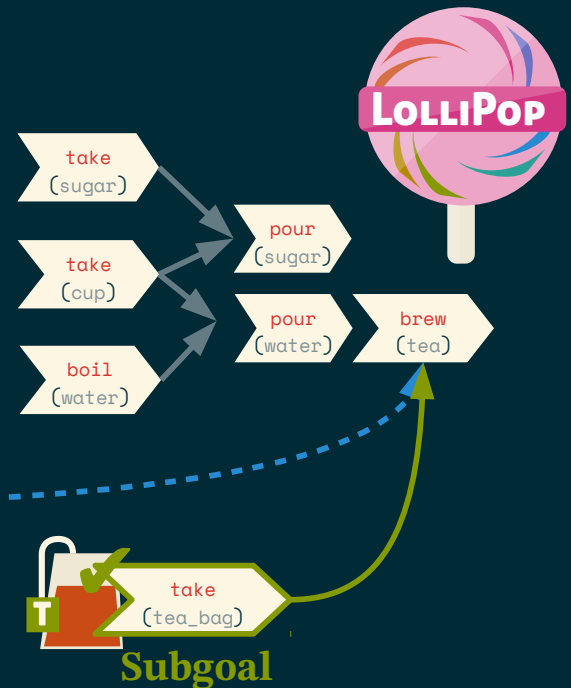
Plan Repair Prototype

[@GRéa] 28

- Partial Order Planner (POP)
- Operator dependency graph
- Negative refinements
- Alternatives & Orphans



- Utility Heuristics



Abstract Planning

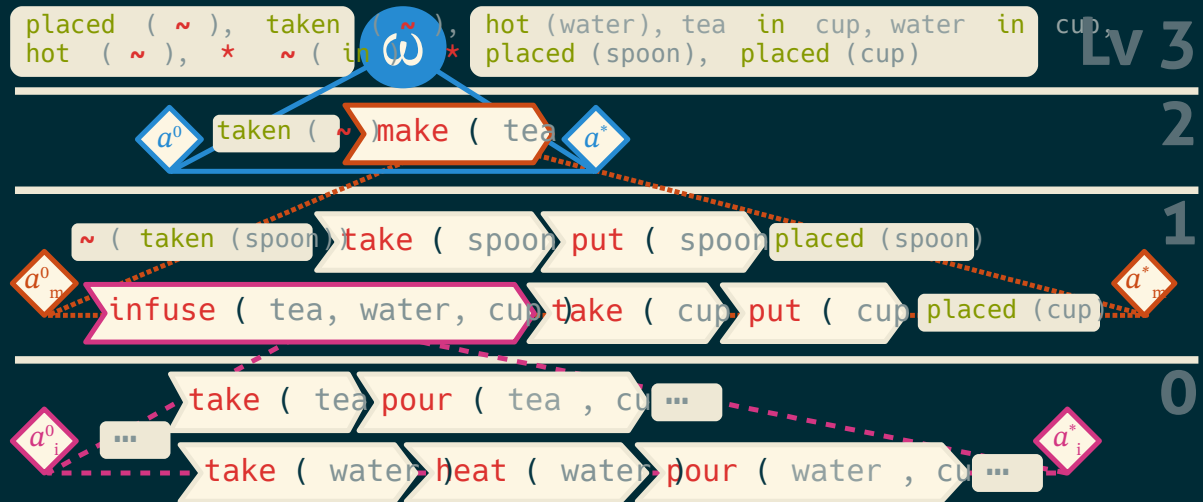
[@GRéa]

29

- HTN + POP planning
- Partial Resolution
 - An abstract solution at every level of abstraction
- Search by level
 - Expansion after completion :
- Decomposition flow
 - Resolver : Decompose one composite action in the plan

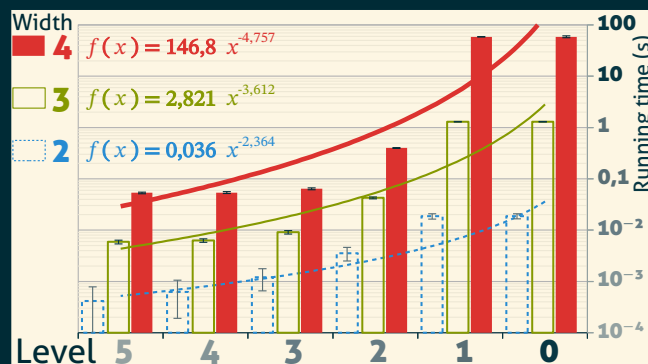
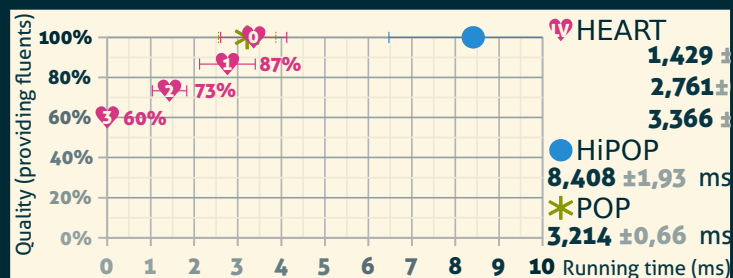


- Low priority for expansion
- Each level is a plan (abstract solution)
 - Change of level
 - Propagation of atomic actions
 - Expansion of Composite Equities



Results

- 60% of the fluents before planning
- Exponentially faster at high abstraction levels
- Faster than HiPOP on some problems
- Common problems solved in milliseconds!

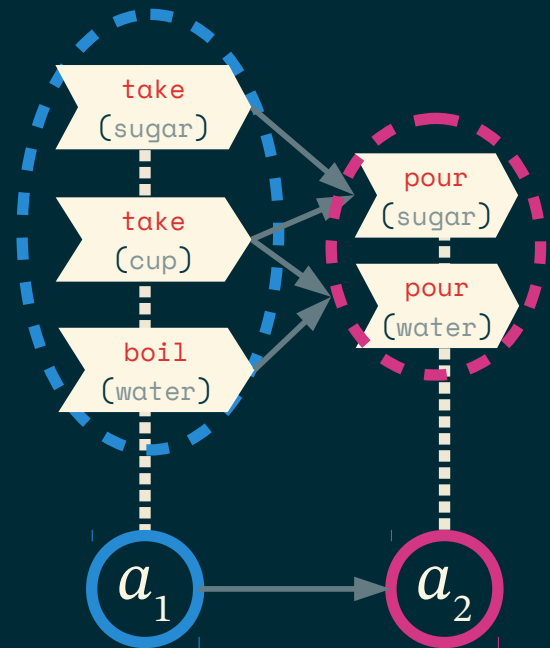


Toward Intent Recognition

[@GRéa]

32

- Linearized parallel actions using graph quotient
- Abstraction makes it easier (smaller plans)
- Backward chaining is inefficient



6 Conclusion

33



Contributions & Results

34

- SELF: A knowledge description language defined by structure
- COLOR: A general framework for planning with its formalization
- LOLLIPOP: A plan repair planner for online planning
- HEART: A flexible approach to real-time planning for abstract planning

- **SELF Improvement**
 - Improve the instantiation workflow
 - Parameterize flexibility performances
- **Planning Colorized**
 - Conversion tool from PDDL
 - Make a clean implementation for community use
- **Fixing Planning Domains**
 - Allow HEART to discover new HTN methods (macro-action learning)

