

# Iterative Depth-First Search for FOND Planning

Ramon Fraga Pereira<sup>1</sup> André Grahl Pereira<sup>2</sup> Frederico Messa<sup>2</sup> Giuseppe De Giacomo<sup>1</sup>



<sup>2</sup>Federal University of Rio Grande do Sul, Porto Alegre, Brazil <sup>1</sup>Sapienza University of Rome, Rome, Italy

#### **Motivation and Contributions**

Fully Observable Non-Deterministic (FOND) planning models uncertainty through actions with non-deterministic effects with uniform probabilities over the actions' effects.

Existing FOND planners are effective and employ a wide range of techniques. However, most of the existing FOND planners are not robust for dealing with both non-determinism and task size.

The main contributions of this work are as follows:

- Iterative Depth-First Search (IDFS) algorithms for FOND Planning:
  - Our algorithms are explicitly designed for solving FOND Planning;
  - Use heuristics to make the iterative process effective in FOND;
  - IDFS is effective to deal with different non-deterministic aspects of FOND Planning;

#### **Background and Notation**

#### **Definition (FOND Planning Task)**

A FOND planning task is tuple  $\Pi = \langle \mathcal{D}, s_0, s_* \rangle$ :

- $\mathcal{D} = \langle \mathcal{F}, \mathcal{A} \rangle$  is a **non-deterministic domain model**, where  $\mathcal{F}$  is a set of *fluents*, and a set of **non-deterministic actions** A. Every  $a \in \mathcal{A}$  consists of  $a = \langle pre, EFFS \rangle$ , where:
  - $\blacksquare$  pre(a) represents the **preconditions**; and
  - EFFS(a) represents the **set of possible effects** of a;
- $\blacksquare$   $s_0$  is the **initial state**;
- $\blacksquare$   $s_*$  is the **goal condition**;

The application of EFFS(a) to a state s generates a set of possible successor states  $SUCCS(s, a) = \{SUCC(s, eff) \mid eff \in EFFS(a)\}.$ 

#### Definition (FOND Planning Solution Policy $\pi$ )

A solution to a FOND planning task  $\Pi$  is a **policy**  $\pi$ , a partial function that maps non-goal states s into actions  $a \in \mathcal{A}$ .

A policy  $\pi$  induces  $\pi$ -trajectories, a non-empty sequence of states  $\langle s^0, s^1, \dots s^k \rangle$ , such that  $s^{i+1} \in SUCCS(s^i, \pi(s^i)), \forall i \in \{0, 1, ..., k-1\}.$ 

A policy  $\pi$  is closed if any  $\pi$ -trajectory starting from  $s_0$  ends either in a goal state or in a state defined in the policy  $\pi$ .

#### **Definition (Strong Policy)**

A policy  $\pi$  is a *strong policy* for  $\Pi$  if it is closed and no  $\pi$ -trajectory passes through a state more than once.

#### **Definition (Strong Cyclic Policy)**

A policy  $\pi$  is a strong cyclic policy for  $\Pi$  if it is closed and any  $\pi$ -trajectory starting from  $s_0$  which does not end in a goal state, ends in a state s' such that exists another  $\pi$ -trajectory starting from s' ending in a goal state.

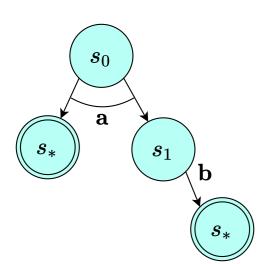


Figure 1. Strong Policy.

Figure 2. Strong Cyclic Policy.

### Evaluation Function $\mathcal{F}$ for FOND

 ${}^{a}g(s)$  represents the search depth from  $s_0$  to  $s_0$ 

We define the f-value<sup>a</sup> of a state s as:

$$f(s) = g(s) + h(s). \tag{1}$$

We evaluate the successor states SUCCS(s, a) using the Evaluation Function  $\mathcal{F}_{\xi}$ .  $\mathcal{F}_{\xi}$  uses a function  $\xi$  to aggregate the f-values of states in SUCCS(s, a):

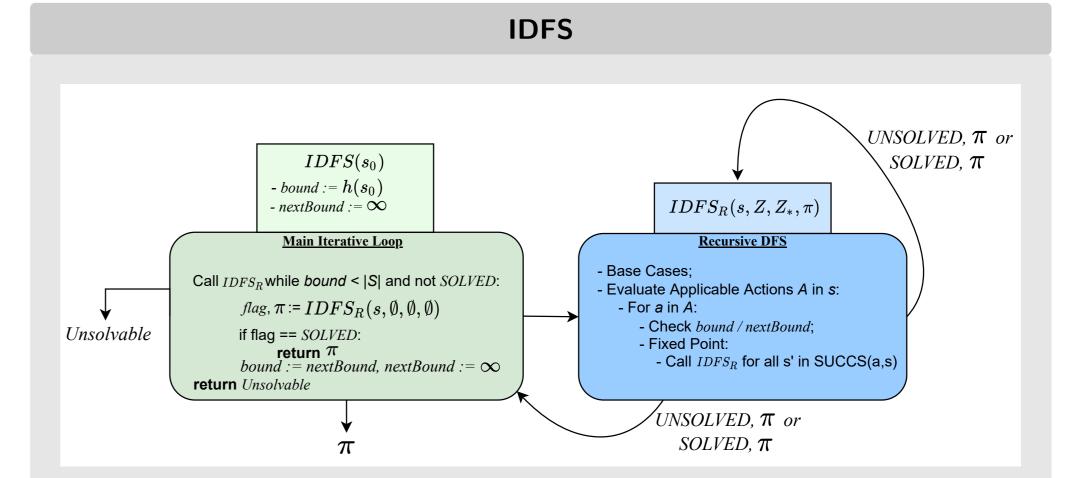
$$\mathcal{F}_{\min}(SUCCS(s,a)) \equiv \min_{s' \in SUCCS(s,a)} f(s')$$
 (2)

$$\mathcal{F}_{\max}(SUCCS(s,a)) \equiv \max_{s' \in SUCCS(s,a)} f(s')$$
(3)

 $\mathcal{F}_{\xi}$  is "pessimistic" when  $\xi = \max$ , whereas it is "optmistic" when  $\xi = \min$ .

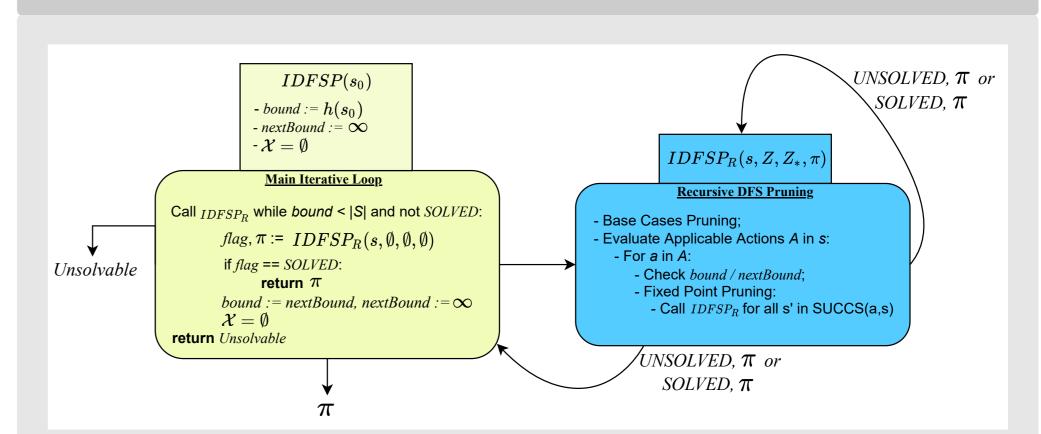
#### **Iterative DFS for FOND Planning**

- IDFS performs a series of **bounded** depth-first searches;
- IDFS produces a strong cyclic policy in a **bottom-up way**;
  - It only adds an action to the policy if it determines that the resulting policy with the additional action has the potential to become a strong cyclic policy without exceeding the current search-depth bound.



- Z ancestors of s;  $Z_*$  ancestors of s that has achieved  $s_*$  through a  $\pi$ -trajectory;
- Base Cases: Check if  $s \models s_*$  OR  $s \in Z_*$  OR  $\pi(s) \neq \bot$ . If so, it returns *SOLVED*; And, if  $s \in (Z \setminus Z_*)$  (s visited before), if so, it returns UNSOLVED.
- **Evaluate Applicable Actions in** s: Check bound for SUCCS(s, a).
- If  $\mathcal{F}_{\varepsilon}(SUCCS(s, a)) > bound$ ) and  $Z_{*} = \emptyset$ : **discards** a on s, and proceeds to the next action.
- **2** If if g(n) + 1 > bound: **discards** a on s, and proceeds to the next action.
- **■** Fixed Point:
- It maps  $s \mapsto a$  to  $\pi$  **ONLY** if **ALL** recursive calls on states of SUCCS(s, a) returned SOLVED;
- $\blacksquare$  If not, it **discards** a on s, and proceeds to the next action.

## IDFS Pruning (IDFSP)



IDFSP **prunes** non-promising states. E.g., it prunes states whose all successor states for all applicable actions have  $\mathcal{F}$  greater than the current bound.

#### **Experiments and Evaluation**

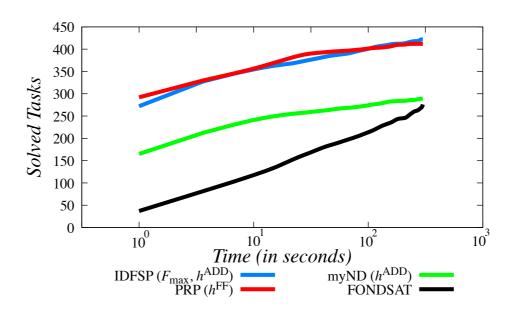
We use two FOND planning benchmarks: IPC-FOND and NEW-FOND.

Our planner is called PALADINUS.

We compare PALADINUS with: MYND, PRP, and FONDSAT.

${ m JS}$ with: MYND, PRP, and FC	ONDSAT.
Planner	Solved Tasks (#590)
PALADINUS $\overline{\text{IDFSP}}$ ( $\mathcal{F}_{\min}, h^{\text{max}}$ )	337
$\text{PALADINUS } \underline{\text{IDFSP}}_{(\mathcal{F}_{\min}, \; h^{\text{FF}})}$	406
$\text{PALADINUS IDFSP}_{(\mathcal{F}_{\min}, \; h^{\text{ADD}})}$	411
PALADINUS $\overline{\text{IDFSP}}$ ( $\mathcal{F}_{\text{max}}$ , $h^{\text{max}}$ )	334
$\text{PALADINUS IDFSP}_{(\mathcal{F}_{\max}, \; h^{\text{ff}})}$	380
$\text{PALADINUS IDFSP}_{(\mathcal{F}_{\text{max}}, \; h^{\text{ADD}})}$	422
FONDSAT	276
$\operatorname{PRP}\left(h^{\scriptscriptstyle{\mathrm{MAX}}}\right)$	292
$PRP_{(h^{\text{ff}})}$	412
$PRP(h^{ADD})$	389
$\mathrm{MYND}_{(h^{\mathrm{MAX}})}$	180
$\mathrm{MYND}_{(h^{\mathrm{FF}})}$	265
$\mathrm{MYND}_{(h^{\mathrm{ADD}})}$	289

Table 1. Overall coverage results.



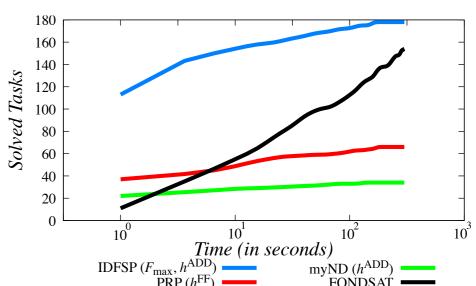


Figure 3. All Benchmarks.

Figure 4. NEW-FOND Benchmarks.

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