

Width and Inference Based Planners: *SIW*, *BFS(f)*, and *PROBE*

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Introduction

We entered the planners *SIW*, *BFS(f)*, and *PROBE* to the *agile-track* of the 2014 International Planning Competition, and an anytime planner for the *satisficing track* that runs both *SIW* and *BFS(f)*. *SIW* and *BFS(f)* are classical planners that make use of the notion of width (Lipovetzky and Geffner 2012), while *PROBE* is a standard best-first search planner that augments the expansion of a node by throwing an “intelligent” probe which either reaches the goal or terminates quickly in low polynomial time (Lipovetzky and Geffner 2011). The basic building block of *SIW* is the Iterative Width Procedure (*IW*) for achieving one atomic goal at a time. *IW* runs in time exponential in the problem width by performing a sequence of pruned breadth first searches. The planner *BFS(f)* integrates a *novelty* measure from *IW* with helpful-actions, landmarks and delete-relaxation heuristics in a Greedy Best-First search.

In the following sections we introduce the basic notions of the algorithms and the implementation.

SIW: Iterated Width Search

The algorithm *Iterated Width*, or *IW*, consists of a sequence of calls $IW(i)$ for $i = 0, 1, \dots, |F|$ until the problem is solved. Each iteration $IW(i)$ is a breadth-first search that prunes right away states that do not pass a *novelty* test; namely, for a state s in $IW(i)$ *not* to be pruned there must be a tuple t of at most i atoms such that s is the first state generated in the search that makes t true. The time complexities of $IW(i)$ and *IW* are $O(n^i)$ and $O(n^w)$ respectively where n is $|F|$ and w is the problem width. The width of existing domains is low for atomic goals, and indeed, 89% of the benchmarks can be solved by $IW(2)$ when the goal is set to any one of the atoms in the goal (Lipovetzky and Geffner 2012). The width of the benchmark domains with conjunctive goals, however, is not low in general, yet such problems can be serialized.

The algorithm *Serialized Iterative Width*, or *SIW*, uses *IW* for serializing a problem into subproblems and for solving the subproblems. *SIW* uses *IW* to greedily achieve one atomic goal at a time until all atomic goals are achieved

jointly. In between, atomic goals may be undone, but after each invocation of *IW*, each of the previously achieved goals must hold. *SIW* will thus never call *IW* more than $|G|$ times where $|G|$ is the number of atomic goals. *SIW* compares surprisingly well to a baseline heuristic search planner based on greedy best-first search and the h_{add} heuristic (Bonet and Geffner 2001), but does not approach the performance of the most recent planners. Nonetheless, *SIW* competes well in domains with no dead-ends and simple serializations.

BFS(f): Novelty Best-First Search

While the blind-search *SIW* procedure competes well with a greedy best-first planner using the additive heuristic, neither planner is state-of-the-art. For this, we developed a standard *forward-search best-first planner* guided by an evaluation function that combines the notions of novelty and helpful actions (Lipovetzky and Geffner 2012; Hoffmann and Nebel 2001). In this planner, called *BFS(f)* (Lipovetzky and Geffner 2012), ties are broken lexicographically by two other measures: (1) the number of subgoals not yet achieved up to a node in the search, and (2) the additive heuristic, h_{add} . The additive heuristic is delayed for non-helpful actions.

PROBE

PROBE is a complete, standard greedy best first search (GBFS) STRIPS planner using the standard additive heuristic (Bonet and Geffner 2001), with just *one change*: when a state is selected for expansion, it first launches a *probe* from the state to the goal (Lipovetzky and Geffner 2011). If the probe reaches the goal, the problem is solved and the solution is returned. Otherwise, the states expanded by probe are added to the open list, and control returns to the GBFS loop. The crucial and only novel part in the planning algorithm is the definition and computation of the probes.

PROBE is built using an early-version of an automated planning toolkit that supports the implementation details of the width-based algorithms. The only difference with respect to *PROBE-IPC7* is that the anytime procedure is disabled, as we are only concerned with the first solution.

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Implementation Notes

The planners *SIW*, *BFS(f)* have been implemented using the automated planning toolkit `lwaptk`¹. The toolkit is an extensible C++ framework that decouples search and heuristic algorithms from PDDL parsing and grounding modules, by relying on planner “agnostic” data structures to represent (ground) fluents and actions. We consider `lwaptk` to be a valuable contribution in itself since it allows to develop, relying on a collection of readily available implementations of search algorithms and planning heuristics, planners which are independent from specific parsing modules and grounding algorithms. If the planner is to be acquiring descriptions of planning tasks from PDDL specifications, the toolkit provides the means to plug into the planner either FF (Hoffmann and Nebel 2001) or FAST-DOWNWARD (Helmert 2006) parsers. Alternatively, and more interestingly, the planner can be embedded into complex applications, *directly*, if the “host” application is written in C++, or *indirectly* when the host is written in an interpreted language, such as PYTHON, by wrapping the planner with suitably generated marshalling code.

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¹Source code available from <https://github.com/miquelramirez/lwaptk>.