A Hyper-Heuristic Using GRASP with Path-Relinking: A Case Study of the Nurse Rostering Problem

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ABSTRACT

The goal of hyper-heuristics is to design and choose heuristics to solve complex problems. The primary motivation behind the hyper-heuristics is to generalize the solving ability of the heuristics. In this paper, the authors propose a Hyper-heuristic using GRASP with Path-Relinking (HyGrasPr). HyGrasPr generates heuristic sequences to produce solutions within an iterative procedure. The procedure of HyGrasPr consists of three phases, namely the construction phase, the local search phase, and the path-relinking phase. To show the performance of the HyGrasPr, the authors use the nurse rostering problem as a case study. The authors use an existing simulated annealing based hyper-heuristic as a baseline. The experimental results indicate that HyGrasPr can achieve better solutions than SAHH within the same running time and the path-relinking phase is effective for the framework of HyGrasPr.

Keywords: GRASP, Heuristics, Hyper-Heuristics, Nurse Rostering Problem, Path-relinking

INTRODUCTION

Hyper-heuristics aim to design general solving technologies for various problems by choosing existing heuristics (Burke, Hyde, Kendall, Ochoa, Ozcan, & Woodward, 2010). In contrast to meta-heuristics focused on the domain knowledge, hyper-heuristics tend to produce the High Level Heuristics (HLHs) for guiding the Low Level Heuristics (LLHs) (Burke, Hyde, Kendall, Ochoa, Ozcan, & Qu, 2010). The high level heuristics are referred to the heuristics designed by algorithm experts over the problem domains while the LLHs are referred to the heuristics designed by the problem domain experts. Since the domain knowledge is necessary for a particular problem and is hard to explore by an algorithm designer (Ochaoa, Qu, & Burke, 2009), the primary motivation behind the hyper-heuristics is to help the algorithm designers to jump out of the limit from the problem domain and to produce general

DOI: 10.4018/jitr.2011040103
approaches. Based on the ability of general problem solving, hyper-heuristics have been applied to many kinds of problems, especially NP-hard problems, such as the timetabling (Burke, McCollum, Meisels, Petrovic, & Qu, 2007; Qu & Burke, 2009), the cutting stock (Terashima-Martin, Moran-Saavedre, & Ross, 2005), the workforce scheduling (Remde, Cowling, Dahal, & Colledge, 2006; Remde, Dahal, Cowling, & Colledge, 2009) and the p-median (Ren, Jiang, Xuan, & Luo, 2010).

In general, the goal of a hyper-heuristic is to design HLH to find an optimal LLH sequence, which can generate optimal solutions to the problems. As one kind of heuristics, most of hyper-heuristics draw on the experiments from the existing meta-heuristics, e.g., a simulated annealing based hyper-heuristic (Dowsland, Soubeiga, & Burke, 2007) and a genetic algorithm based hyper-heuristic (Ross, Martin-Blazquez, Schulenburg, & Hart, 2003). However, the kinds of hyper-heuristics are much fewer than those of meta-heuristics. The insufficiency of hyper-heuristics has limited the development of hyper-heuristics (Burke, Hyde, Kendall, Ochoa, Ozcan, & Woodward, 2010).

Motivated by the success of this algorithm in meta-heuristics, we propose a Hyper-heuristic using GRASP with Path-Relinking (HyGrasPr) in this paper. Our algorithm, HyGrasPr, generates LLH sequences to produce solutions in an iterative procedure. In each iteration, HyGrasPr builds an initial LLH sequence and applies a local search operator to find a relatively good LLH sequence. To avoid to be trapped as a local optimal LLH sequence, the path-relinking strategy is applied to obtain potential good solutions. To show the experimental results of HyGrasPr, we take the nurse rostering problem as a case study. On this problem, an existing simulated annealing based hyper-heuristic (SAHH) is employed as an experiment baseline. The results indicate that HyGrasPr can achieve better solutions than SAHH within the same running time.

The rest of this paper is organized as follows. First, we give the background of our work. We then propose the details of our HyGrasPr and present the experiments results with a case study on the nurse rostering problem.

**BACKGROUND**

**Hyper-Heuristics and Meta-Heuristics**

Hyper-heuristic technology is able to handle a wide range of problem domains rather than current meta-heuristic technology concentrated on a particular problem or a narrow class of problems (Burke, Kendall, Newall, Hart, Ross, & Schulenburg, 2003). A solution of a meta-heuristic is a structure abstracted from the problem domain; on the other hand, a solution of a hyper-heuristic is a sequence of LLHs. Many hyper-heuristics are based on the mechanism from meta-heuristics, e.g., hyper-heuristics based on tabu search for timetabling and rostering (Burke, Kendall, & Soubeiga, 2003), simulated annealing for determining shipper sizes (Dowsland et al., 2007), for automated planograms (Bai & Kendall, 2005), genetic algorithm (bin-packing) (Ross et al., 2003), and for 2D-regular cutting stock problems (Terashima-Martin et al., 2005), genetic programming for two dimensional strip packing (Burke, Hyde, Kendall, Ochoa, Ozcan, & Qu, 2010), and ant colony optimization for project presentation scheduling (Burke, Kendall, Silva, O’Brien, & Soubeiga, 2005) and for p-median by Ren et al. (2010). Moreover, Bai, Burke and Kendall (2008) investigate both the meta-heuristics and hyper-heuristics for fresh produce inventory control and shelf space allocation. Considering the similarity
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