The Permutation Problem
Using a Unit-Capacity Robot

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Abstract
In this paper, we consider the permutation problem using a unit-capacity robot. In particular, we consider an approach to solve the problem. This approach is based on an explicit reduction from the problem to the satisfiability problem.

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The permutation problem using a unit-capacity robot (PPCR) arises in nuclear power plant fuel renewal (see [1, 2]). The PPCR can be considered as a particular pickup and delivery problem [3]. The PPCR is NP-complete [2]. In this paper, we consider an approach to solve PPCR. This approach is based on an explicit reduction from PPCR to the satisfiability problem.
In PPCR, it is assumed that the directed graph
\[ G = (V, A) \]
describes all possible moves. We consider a set of
\[ P = \{1, 2, \ldots, p\} \]
pieces that have to be moved over a set \( W \subseteq V \) of particular locations. It is assumed that any location contains at most one piece.

Each piece from \( P \) possesses a specific type. For any type, there exists only one node in \( V \) having a piece of this type. We assume that, in the initial state, each piece is assigned to a given location and, in the final state, each location must contain a piece of a given type. Is there a sequence of \( n \leq N \) states from the initial state to the final state?

Pieces are handled by a unit-capacity robot. It is assumed that such a robot can only move one piece at a time. Also, it is assumed that each piece can be moved only once.

We need two functions for initial and final states:
\[ I : V \to P \cup \{0\}, \]
\[ F : V \to P \cup \{0\}. \]
For any \( j \) such that \( 1 \leq j \leq |V| \), let \( I[j] = 0 \) if and only if, in the initial state, node \( j \) containing no piece. Let \( I[j] = k, 0 < k \leq p, 1 \leq j \leq |V| \), if and only if, in the initial state, node \( j \) having piece of type \( k \). We assume that \( F[j] = 0, 1 \leq j \leq |V| \), if and only if, in the final state, node \( j \) containing no piece. Let \( F[j] = k, 0 < k \leq p, 1 \leq j \leq |V| \), if and only if, in the final state, node \( j \) having piece of type \( k \).

\[ \varphi[1] = \bigwedge_{1 \leq j \leq |V|} x[1, j, k]; \]
\[ \varphi[2] = \bigwedge_{1 \leq j \leq |V|} x[N, j, k]; \]
\[ \varphi[3] = \bigwedge_{1 \leq i \leq N} (\neg x[i, j, k[1]] \lor \neg x[i, j, k[2]]); \]
The permutation problem using a unit-capacity robot

\[ \varphi[4] = \bigwedge_{1 \leq i \leq N} \bigvee_{0 \leq k \leq p} x[i, j, k]; \]

\[ \varphi[5] = \bigwedge_{1 \leq i \leq N-1} \bigvee_{1 \leq j[1] < j[2] \leq |V|} (-s[i, j[1]] \lor -s[i, j[2]]); \]

\[ \varphi[6] = \bigwedge_{1 \leq i \leq N-1} \bigvee_{1 \leq j \leq |V|} s[i, j]; \]

\[ \varphi[7] = \bigwedge_{1 \leq i \leq N-1} \bigvee_{1 \leq j[1] < j[2] \leq |V|} (-t[i, j[1]] \lor -t[i, j[2]]); \]

\[ \varphi[8] = \bigwedge_{1 \leq i \leq N-1} \bigvee_{1 \leq j \leq |V|} t[i, j]; \]

\[ \varphi[9] = \bigwedge_{1 \leq i \leq N-1} (-s[i, j] \lor -x[i, j, 0]); \]

\[ \varphi[10] = \bigwedge_{1 \leq i \leq N-1} (-s[i, j] \lor x[i, j, 0]); \]

\[ \varphi[11] = \bigwedge_{1 \leq i \leq N-1} (-t[i, j] \lor x[i, j, 0]); \]

\[ \varphi[12] = \bigwedge_{1 \leq i \leq N-1} (-t[i, j] \lor -x[i + 1, j, 0]); \]

\[ \varphi[13] = \bigwedge_{1 \leq i \leq N-1} (-s[i, j[1]] \lor -t[i, j[2]]); \]

\[ \varphi[14] = \bigwedge_{1 \leq k \leq p} (-y[k, i[1]] \lor -y[k, i[2]]); \]

\[ \varphi[15] = \bigwedge_{1 \leq k \leq p} \bigvee_{1 \leq i \leq N-1} y[k, i]; \]

\[ \varphi[16] = \bigwedge_{1 \leq k \leq p} \bigvee_{1 \leq i \leq N-1} (-s[i, j] \lor -y[k, i] \lor x[i, j, k]); \]
\( \varphi[17] = \bigwedge_{1 \leq k \leq p} \bigwedge_{1 \leq i \leq N-1} \bigwedge_{1 \leq j \leq |V|} (\neg t[i, j] \lor \neg y[k, i] \lor x[i + 1, j, k]); \)

\[
\eta = \bigwedge_{r=1}^{17} \varphi[r].
\]

Using standard transformations, we can obtain an explicit transformation \( \eta \) into \( \zeta \) such that

\[
\eta \leftrightarrow \zeta
\]

and \( \zeta \) is a 3-CNF. Clearly, \( \zeta \) gives us an explicit reduction from PPCR to 3SAT.

To obtain optimal solutions of PPCR we use genetic algorithms OA[1] (see [4]), OA[2] (see [5]), and OA[3] (see [6]) for the satisfiability problem. We have used heterogeneous cluster. Each test was runned on a cluster of at least 100 nodes. Selected experimental results are given in Table 1.

<table>
<thead>
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<th>time</th>
<th>average</th>
<th>max</th>
<th>best</th>
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<tr>
<td>OA[1]</td>
<td>17.52 min</td>
<td>1.97 hr</td>
<td>3.12 min</td>
</tr>
<tr>
<td>OA[2]</td>
<td>15.89 min</td>
<td>2.45 hr</td>
<td>1.43 min</td>
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<tr>
<td>OA[3]</td>
<td>11.54 min</td>
<td>27.34 min</td>
<td>29 sec</td>
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</table>

Table 1: Experimental results for different genetic algorithms.

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