

**Supplementary File for “Minimizing energy consumption of flexible assembly systems with tool change processes using hybrid heuristic search”**

*Theorem 1:*  $h_1, h_2, h_3$ , and  $h_5$  are admissible.

*Proof:* Let  $\beta^*$  denote the actual optimal schedule from  $(M, \alpha)$  to a final vertex, i.e.,  $M[\beta^*] > M_F$ . Let  $h^*(M, \alpha)$  denote the total energy consumption of  $\beta^*$ . Let  $E_W(\alpha, \beta^*)$ ,  $E_O(\alpha, \beta^*)$ ,  $E_I(\alpha, \beta^*)$ ,  $E_H(\alpha, \beta^*)$  and  $E_C(\alpha, \beta^*)$  denote the total energy consumption of working, occupied, idle, hold and tool change state from  $\alpha$  to  $\alpha\beta^*$ , respectively. Under  $E_1$ ,  $h^*(M, \alpha) = E_H(\alpha, \beta^*) + E_I(\alpha, \beta^*) + E_C(\alpha, \beta^*)$ , while under  $E_2$ ,  $h^*(M, \alpha) = E_W(\alpha, \beta^*) + E_O(\alpha, \beta^*) + E_I(\alpha, \beta^*) + E_C(\alpha, \beta^*)$ .

Under  $E_2$ ,  $h_1(M, \alpha) = \sum_{p \in P_A \cup P_B} \sum_{j \in \mathbf{N}_{M(p)}} (E_R(p, j, M, \alpha) + E_P(p))$  is the minimum working energy consumption of all parts from  $(M, \alpha)$  to the end, thus  $h_1(M, \alpha) \leq E_W(\alpha, \beta^*)$  under  $E_2$ . Since  $\sum_{t \in \Pi} \kappa(t, M, \alpha)J(t, M, \alpha)$  is only part of the occupied energy,  $\sum_{t \in \Pi} \kappa(t, M, \alpha)J(t, M, \alpha) \leq E_O(\alpha, \beta^*)$ . Since  $\sum_{r_i \in R} \delta(r_i, M, \alpha)G(r_i, M, \alpha)$  is only part of the idle energy,  $\sum_{r_i \in R} \delta(r_i, M, \alpha)G(r_i, M, \alpha) \leq E_I(\alpha, \beta^*)$ . Since  $E_C(\alpha, \beta^*) \geq 0$ ,  $h_5(M, \alpha) = h_1(M, \alpha) + \sum_{r_i \in R} \delta(r_i, M, \alpha)G(r_i, M, \alpha) + \sum_{t \in \Pi} \kappa(t, M, \alpha)J(t, M, \alpha) \leq E_W(\alpha, \beta^*) + E_O(\alpha, \beta^*) + E_I(\alpha, \beta^*) + E_C(\alpha, \beta^*) = h^*(M, \alpha)$ , hence  $h_5(M, \alpha)$  is admissible under  $E_2$ . By the definitions of  $h_1(M, \alpha)$ - $h_5(M, \alpha)$ ,  $h_2(M, \alpha) \leq h_5(M, \alpha)$  and  $h_1(M, \alpha) \leq h_3(M, \alpha) \leq h_5(M, \alpha)$ , hence  $h_1(M, \alpha)$ ,  $h_2(M, \alpha)$  and  $h_3(M, \alpha)$  are admissible under  $E_2$ .

Since  $(h_1(M, \alpha) + \sum_{t \in \Pi} \kappa(t, M, \alpha)J(t, M, \alpha))$  is only part of the hold energy consumption of all parts from  $(M, \alpha)$  to the end,  $h_1(M, \alpha) + \sum_{t \in \Pi} \kappa(t, M, \alpha)J(t, M, \alpha) \leq E_H(\alpha, \beta^*)$ .  $h_5(M, \alpha) \leq E_H(\alpha, \beta^*) + E_I(\alpha, \beta^*) + E_C(\alpha, \beta^*) = h^*(M, \alpha)$ , hence  $h_5(M, \alpha)$  is admissible under  $E_1$ . By the definitions of  $h_1(M, \alpha)$ - $h_5(M, \alpha)$ ,  $h_2(M, \alpha) \leq h_5(M, \alpha)$  and  $h_1(M, \alpha) \leq h_3(M, \alpha) \leq h_5(M, \alpha)$ , hence  $h_1(M, \alpha)$ ,  $h_2(M, \alpha)$  and  $h_3(M, \alpha)$  are admissible under  $E_1$ . ■

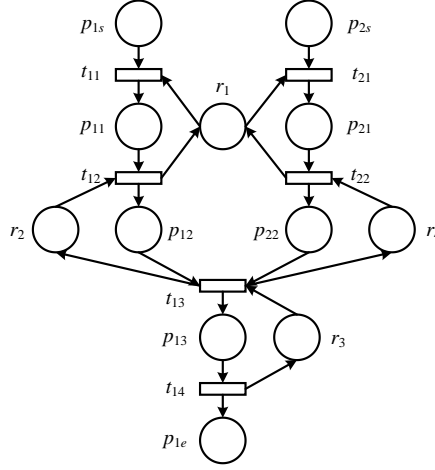


Fig. 1. APNS of FAS in Example 1.

**Example 1:** For the APNS processing two types of parts in Fig. 1, suppose  $M_0 = p_{1s} + p_{2s} + r_1 + r_2 + r_3 + r_4$ . The processing time of operations in places  $p_{11}$ ,  $p_{12}$ ,  $p_{13}$ ,  $p_{21}$  and  $p_{22}$  are 10, 5, 5, 30 and 20, respectively. The energy consumption per unit time in different kinds of states is listed in Table I. For tool change processes,  $\mu = 5$  and  $e_c = 2$ .

TABLE I

ENERGY CONSUMPTION PER UNIT TIME OF RESOURCES IN FIG. 1.

	$r_1$	$r_2$	$r_3$	$r_4$
Hold state	2.00	10.00	2.00	2.00
Working state	2.00	10.00	2.00	2.00
Occupied state	1.00	5.00	1.00	1.00
Idle state	0.20	1.00	0.20	0.20

Let's calculate  $h_4(M_0, \alpha_0)$  under  $E_1$  and  $E_2$ . The first part of  $h_4(M_0, \alpha_0)$  is  $h_1(M_0, \alpha_0)$ .  $E_R(p_{1s}, 1, M_0, \alpha_0) = E_R(p_{2s}, 1, M_0, \alpha_0) = 0$ ,  $E_P(p_{1s}) = d(p_{11}) \times e_0(r_1) + d(p_{12}) \times e_0(r_2) + d(p_{13}) \times e_0(r_3) = 10 \times 2 + 5 \times 10 + 5 \times 2 = 20 + 50 + 10 = 80$  and  $E_P(p_{2s}) = d(p_{21}) \times e_0(r_1) + d(p_{22}) \times e_0(r_4) = 30 \times 2 + 20 \times 2 = 100$ . Then  $h_1(M_0, \alpha_0) = E_R(p_{1s}, 1, M_0, \alpha_0) + E_P(p_{1s}) + E_R(p_{2s}, 1, M_0, \alpha_0) + E_P(p_{2s}) = 180$ .

The rest of  $h_4(M_0, \alpha_0)$  under  $E_1$  is calculated as follows.  $\Pi = \{t_{12}, t_{22}, t_{13}\}$ , for  $t_{12}$  and  $t_{22}$ , since  $M(p_{11}) = M(p_{21}) = 0$ ,  $\theta(t_{12}, M_0, \alpha_0) = \theta(t_{22}, M_0, \alpha_0) = 0$ . For assembly transition  $t_{13}$ , since  $p_{12} \in {}^{(o)}t_{13}$ ,  $p_{1s} \in I(p_{12})$  and  $M(p_{1s}) = 1 > 0$ ,  $\theta(t_{13}, M_0, \alpha_0) = 1$ .

$\Pi = \{t_{12}, t_{13}, t_{22}\}$ . Since  $|{}^{(o)}t_{12}| = |{}^{(o)}t_{22}| = 1$  and  $M_0(p_{11}) = M_0(p_{21}) = 0$ ,  $\theta(t_{12}, M_0, \alpha_0) = \theta(t_{22}, M_0, \alpha_0) = 0$ .  $E_{O1}(t_{13}, M_0, \alpha_0)$  and  $E_{O2}(t_{13}, M_0, \alpha_0)$  are calculated as follows. Since  $M_0(r_3) > 0$ ,  $T_R(r_3, M_0, \alpha_0) = 0$ .  $T_O(t_{13}, M_0, \alpha_0) = \max\{T_O(p_{12}, M_0, \alpha_0), T_O(p_{22}, M_0, \alpha_0)\} = \max\{d(p_{11}) + d(p_{12}), d(p_{21}) + d(p_{22})\} = \max\{10 + 5, 30 + 20\} = 50$ . Then  $E_{O1}(t_{13}, M_0, \alpha_0) = (e_0(r_2) + e_0(r_4)) \times (\max\{T_O(t_{13}, M_0, \alpha_0), T_R(r_3, M_0, \alpha_0)\} - T_O(t_{13}, M_0, \alpha_0)) = (10 + 2) \times (\max\{50, 0\} - 50) = 0$ , and  $E_{O2}(t_{13}, M_0, \alpha_0) = (T_O(t_{13}, M_0, \alpha_0) - T_O(p_{12}, M_0, \alpha_0)) \times e_0(r_2) + (T_O(t_{13}, M_0, \alpha_0) - T_O(p_{22}, M_0, \alpha_0)) \times e_0(r_4) = (50 - 15) \times 10 + (50 - 50) \times 2 = 350$ . Thus  $h_4(M_0, \alpha_0) = h_1(M_0, \alpha_0) + \theta(t_{13}, M_0, \alpha_0)(E_{O1}(t_{13}, M_0, \alpha_0) + E_{O2}(t_{13}, M_0, \alpha_0)) = 180 + 1 \times (0 + 350) = 530$  under  $E_1$ .

Similarly, under  $E_2$ ,  $E_{O2}(t_{13}, M_0, \alpha_0) = (T_O(t_{13}, M_0, \alpha_0) - T_O(p_{12}, M_0, \alpha_0)) \times e_1(r_2) + (T_O(t_{13}, M_0, \alpha_0) -$

$T_O(p_{22}, M_0, \alpha_0) \times e_1(r_4) = (50 - 15) \times 5 + (50 - 50) \times 1 = 175$ , hence  $h_4(M_0, \alpha_0) = 180 + 1 \times (0 + 175) = 355$ .

Let's calculate  $h_6(M_0, \alpha_0)$  under  $E_1$  and  $E_2$ . Since  ${}^{(o)}(r_1^\bullet) = \emptyset$ ,  $\delta(r_1, M_0, \alpha_0) = 0$ . Since  ${}^{(o)}(r_2^\bullet) = \{p_{11}\}$ ,  ${}^{(o)}(r_4^\bullet) = \{p_{21}\}$ ,  ${}^{(o)}(r_3^\bullet) = \{p_{12}, p_{22}\}$ , and  $M(p_{11}) = M(p_{12}) = M(p_{21}) = M(p_{22}) = 0$ ,  $\delta(r_2, M_0, \alpha_0) = \delta(r_3, M_0, \alpha_0) = \delta(r_4, M_0, \alpha_0) = 0$ . Then  $h_2(M_0, \alpha_0) = h_1(M_0, \alpha_0)$  under both  $E_1$  and  $E_2$ .  $h_6(M_0, \alpha_0) = h_4(M_0, \alpha_0) = 530$  under  $E_1$  and  $h_6(M_0, \alpha_0) = 355$  under  $E_2$ .

$$\mathbf{T}(\alpha) = \begin{pmatrix} 35 & 50 & 60 & 70 \\ 0 & 35 & 60 & \end{pmatrix}$$

The energy consumption of a feasible schedule  $\alpha = t_{21}t_{22}t_{11}t_{12}t_{13}t_{14}$  is calculated as follows. It can be checked that  $M_0[\alpha > M_F]$ . To calculate the energy consumption during  $\alpha$ , matrix  $\mathbf{T}(\alpha)$  is generated. Let  $q_{ij}$  denote the  $j$ -th operation for  $q_i$ , then the Gantt chart of  $\alpha$  is shown in Fig. 2.

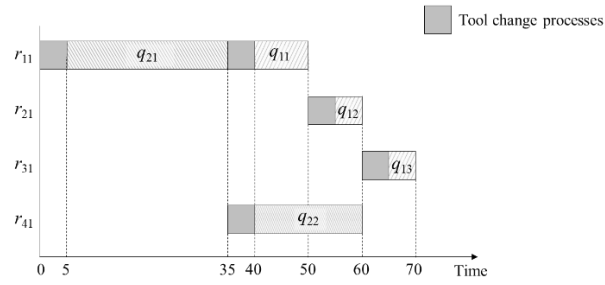


Fig. 2. Gantt chart of schedule  $\alpha$

Under  $E_1$ ,  $E_H(\alpha) = e_0(r_1)T_H(\alpha, p_{11}) + e_0(r_2)T_H(\alpha, p_{12}) + e_0(r_3)T_H(\alpha, p_{13}) + e_0(r_1)T_H(\alpha, p_{21}) + e_0(r_4)T_H(\alpha, p_{22}) = 2 \times 10 + 10 \times 5 + 2 \times 5 + 2 \times 30 + 2 \times 20 = 180$ . From Fig. 2, it is easy to find that the idle time of  $r_1$ ,  $r_2$ ,  $r_3$  and  $r_4$  are 20, 60, 60 and 45, respectively. Thus,  $E_I(\alpha) = 20 \times 0.2 + 60 \times 1 + 60 \times 0.2 + 45 \times 0.2 = 85$ . Tool change processes occur before  $q_{11}$ ,  $q_{12}$ ,  $q_{13}$ ,  $q_{21}$  and  $q_{22}$ , thus  $E_C(\alpha) = 50$ . Then  $E_1(\alpha) = E_H(\alpha) + E_I(\alpha) + E_C(\alpha) = 180 + 85 + 50 = 315$ . Since there is no occupied time during  $\alpha$  under  $E_2$ ,  $E_2(\alpha) = E_1(\alpha) = 315$ .

Let  $h_a^*(M_0, \alpha_0)$  and  $h_b^*(M_0, \alpha_0)$  be the actual optimal total energy consumption from  $(M_0, \alpha_0)$  to  $M_F$  under  $E_1$  and  $E_2$ . For a feasible solution  $\alpha$ ,  $E_1(\alpha) \geq h_a^*(M_0, \alpha_0)$  and  $E_2(\alpha) \geq h_b^*(M_0, \alpha_0)$ . Since  $h_6(M_0, \alpha_0) = h_4(M_0, \alpha_0) > E_1(\alpha) \geq h_a^*(M_0, \alpha_0)$  under  $E_1$  and  $h_6(M_0, \alpha_0) = h_4(M_0, \alpha_0) > E_2(\alpha) \geq h_b^*(M_0, \alpha_0)$  under  $E_2$ ,  $h_4$  and  $h_6$  are not admissible under both  $E_1$  and  $E_2$ .

**Input:** An APNS( $N, M_0$ ) and  $high, max\_vertexes, max\_size, max\_top$ ;

**Output:** a feasible transition sequence  $\alpha$ ;

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1:  $OPEN = \{(M_0, \alpha_0)\}$ ;  $CLOSED = \emptyset$ ;  $bottom\_depth = 0$ ;  $top\_depth = high$ ; /* initialization */
2: while ( $OPEN \neq \emptyset$ ) do
3:   if ( $\varphi(bottom\_depth) = 0$ )
4:      $top\_depth++$ ;  $bottom\_depth++$ ;
5:   end if
6:   select a vertex  $(M, \alpha)$  from  $OPEN$  with minimum  $f(M, \alpha)$  and  $|\alpha| < top\_depth$ ;
7:    $OPEN = OPEN \setminus \{(M, \alpha)\}$ ;  $CLOSED = CLOSED \cup \{(M, \alpha)\}$ ;
8:   sort elements in  $\Xi(M, C)$  in ascending order according to their  $A(t, M, \alpha)$ ;
9:    $\nu(\alpha) = 0$ ; /*  $\nu(\alpha)$  counts the number of successor vertexes of  $(M, \alpha)$  */
10:  while ( $\Xi(M, C) \neq \emptyset$ ) do
11:    let  $t \in \Xi(M, C)$  be the first transition in  $\Xi(M, C)$ ;
12:     $\Xi(M, C) = \Xi(M, C) \setminus \{t\}$ ;
13:    let  $M[t > M_1$ , and  $\alpha_1 = \alpha t$ ;
14:    if ( $M_1 = M_F$ )
15:      return  $\alpha_1$ ;
16:    end if
17:    if ( $\exists$  a node  $(M_1, \alpha_2)$  in  $OPEN$  satisfying  $B(M_1, \alpha_1) \leq B(M_1, \alpha_2)$ )
18:       $OPEN = (OPEN \setminus \{(M_1, \alpha_2)\}) \cup \{(M_1, \alpha_1)\}$ ;
19:    else if ( $\nexists$  a node with  $M_1$  in  $OPEN \cup CLOSED$  or  $\exists$  a node  $(M_1, \alpha_2)$  in  $CLOSED$  satisfying  $B(M_1, \alpha_1) < B(M_1, \alpha_2)$ )
20:      if ( $\varphi(|\alpha_1|) < max\_size$  or  $f(M_1, \alpha_1) < \tilde{f}(\alpha_1)$ )
21:         $OPEN = OPEN \cup \{(M_1, \alpha_1)\}$ ;  $\nu(\alpha)++$ ;
22:      end if
23:      if ( $\varphi(top\_depth) \geq max\_top$ )
24:        discard vertexes at depth  $bottom\_depth$ ;
25:         $top\_depth++$ ;  $bottom\_depth++$ ;
26:      end if
27:      if ( $\nu(\alpha) \geq max\_vertexes$ ), break;
28:    end if
29:  end if-else if
30: end while
31: end while
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*Theorem 2:* D<sup>2</sup>WS can always output a solution

*Proof:* All vertexes in D<sup>2</sup>WS are generated under a DAP, which prohibit transition firings that lead APNS from safe markings to deadlock. Thus all markings of generated vertexes are safe. That is to say, from any markings except for  $M_F$ , there is at least one enabled transition that can fire and lead to a safe marking.

There are two termination conditions for D<sup>2</sup>WS: the final marking is reached or  $OPEN = \emptyset$ . If D<sup>2</sup>WS terminates with the final marking reached, then D<sup>2</sup>WS yields a solution. Assume the final marking has not been reached but D<sup>2</sup>WS terminates with  $OPEN = \emptyset$ . Then, all vertexes in the current window (upon termination of the algorithm) are explored. Let  $(M, \alpha)$  be an explored vertex with the deepest depth among all the vertexes in the current window. Since  $M_F \neq M$ , there must be a vertex  $(M_1, \alpha_1)$  in the current window generated from  $(M, \alpha)$  and  $|\alpha_1| = |\alpha| + 1$ .  $(M_1, \alpha_1)$  will be kept in the current window unless there is a vertex at the depth of  $|\alpha_1|$  with marking  $M_1$  or the number of successor vertexes of  $(M, \alpha)$  is more than  $max\_size$ . That is a contradiction to the assumption that  $(M, \alpha)$  is an explored vertex with the deepest depth among all vertexes in the current window. Thus, D<sup>2</sup>WS can always end with the final marking. ■

The parameters of the search windows are randomly distributed in the range of  $[2, 5]$  for  $max\_vertexes$ ,  $[2, 7]$  for  $high$ , and  $[2, 5]$  for  $max\_size$ .  $max\_top$  is always 3 times as big as  $max\_size$ . The parameters of search windows used in the first experiment are shown in II

TABLE II  
PARAMETERS OF SEARCH WINDOWS.

Search windows	<i>high</i>	<i>max_vertexes</i>	<i>max_size</i>	<i>max_top</i>
SW01	3	3	3	9
SW02	6	4	3	9
SW03	5	4	2	6
SW04	2	2	3	9
SW05	4	2	3	9
SW06	2	2	5	15
SW07	3	4	2	6
SW08	4	2	2	6
SW09	4	5	4	12
SW10	4	4	3	9

TABLE III

PARAMETERS OF SEARCH WINDOWS USED IN THE SECOND EXPERIMENT.

	Search windows	<i>high</i>	<i>max_vertexes</i>	<i>max_size</i>	<i>max_top</i>
Group 1	GW01	2	2	2	6
	GW02	2	2	3	9
	GW03	2	3	3	9
	GW04	3	3	3	9
Group 2	GW05	3	3	4	12
	GW06	3	4	4	12
	GW07	4	4	4	12
	GW08	4	4	5	15
Group 3	GW09	4	5	5	15
	GW10	5	5	5	15
	GW11	5	5	6	18
	GW12	5	6	6	18
Group 4	GW13	6	6	6	18
	GW14	6	6	7	21
	GW15	6	7	7	21
	GW16	7	7	7	21

TABLE IV

RESULTS OF NONPARAMETRIC FRIEDMAN'S TESTS (SIGNIFICANCE LEVEL = 0.05).

Data for nonparametric Friedman's tests (significance level = 0.05)	$p$ -value
The average results found by six heuristic functions under $E_1$	0.000
The best results found by six heuristic functions under $E_1$	0.000
The average results found by six heuristic functions under $E_1$	0.000
The best results found by six heuristic functions under $E_2$	0.000
The average results found by $h_4$ under $E_1$ using four groups of search windows	0.000
The best results found by $h_4$ under $E_1$ using four groups of search windows	0.000
The average results found by $h_6$ under $E_2$ using four groups of search windows	0.000
The best results found by $h_6$ under $E_2$ using four groups of search windows	0.000

The multiple comparisons method used in this paper can be found in (M. Hollander and D. A. Wolfe, *Nonparametric Statistical Methods*. New York, NY, USA: Wiley, 1973.)

TABLE V

MULTIPLE COMPARISON RESULTS USING THE AVERAGE SOLUTIONS FOUND BY SIX HEURISTIC FUNCTIONS UNDER  $E_1$ .

$h_i$	$h_j$	$p$ -value
1	2	0.713
1	3	0.026
1	4	0.000
1	5	0.001
1	6	0.000
2	3	0.000
2	4	0.000
2	5	0.000
2	6	0.000
3	4	0.000
3	5	0.920
3	6	0.000
4	5	0.000
4	6	1.000
5	6	0.000

TABLE VI

MULTIPLE COMPARISON RESULTS USING THE BEST SOLUTIONS FOUND BY SIX HEURISTIC FUNCTIONS UNDER  $E_1$ .

$h_i$	$h_j$	$p$ -value
1	2	0.990
1	3	0.023
1	4	0.000
1	5	0.777
1	6	0.000
2	3	0.003
2	4	0.000
2	5	0.398
2	6	0.000
3	4	0.000
3	5	0.425
3	6	0.000
4	5	0.000
4	6	0.724
5	6	0.000



TABLE VII

MULTIPLE COMPARISON RESULTS USING THE AVERAGE SOLUTIONS FOUND BY SIX HEURISTIC FUNCTIONS UNDER  $E_2$ .

$h_i$	$h_j$	$p$ -value
1	2	0.065
1	3	0.000
1	4	0.000
1	5	0.000
1	6	0.000
2	3	0.000
2	4	0.000
2	5	0.182
2	6	0.000
3	4	0.000
3	5	0.277
3	6	0.000
4	5	0.000
4	6	0.686
5	6	0.000

TABLE VIII

MULTIPLE COMPARISON RESULTS USING THE BEST SOLUTIONS FOUND BY SIX HEURISTIC FUNCTIONS UNDER  $E_2$ .

$h_i$	$h_j$	$p$ -value
1	2	0.944
1	3	0.000
1	4	0.006
1	5	0.547
1	6	0.000
2	3	0.000
2	4	0.079
2	5	0.110
2	6	0.000
3	4	0.000
3	5	0.003
3	6	0.000
4	5	0.000
4	6	0.548
5	6	0.000

TABLE IX

MULTIPLE COMPARISON RESULTS USING AVERAGE SOLUTIONS FOUND BY FOUR GROUPS OF SEARCH WINDOWS AND  $h_4$  UNDER  $E_1$ .

Group $i$	Group $j$	$p$ -value
1	2	0.000
1	3	0.000
1	4	0.000
2	3	0.000
2	4	0.000
3	4	0.986

TABLE X

MULTIPLE COMPARISON RESULTS USING THE BEST SOLUTIONS FOUND BY FOUR GROUPS OF SEARCH WINDOWS AND  $h_4$  UNDER  $E_1$ .

Group $i$	Group $j$	$p$ -value
1	2	0.000
1	3	0.000
1	4	0.000
2	3	0.027
2	4	0.032
3	4	1.000

TABLE XI

MULTIPLE COMPARISON RESULTS USING AVERAGE SOLUTIONS FOUND BY FOUR GROUPS OF SEARCH WINDOWS AND  $h_6$  UNDER  $E_2$ .

Group $i$	Group $j$	$p$ -value
1	2	0.000
1	3	0.000
1	4	0.000
2	3	1.000
2	4	0.004
3	4	0.004

TABLE XII

MULTIPLE COMPARISON RESULTS USING THE BEST SOLUTIONS FOUND BY FOUR GROUPS OF SEARCH WINDOWS AND  $h_6$  UNDER  $E_2$ .

Group $i$	Group $j$	$p$ -value
1	2	0.005
1	3	0.107
1	4	0.467
2	3	0.644
2	4	0.000
3	4	0.002