

# Эволюционные алгоритмы для задачи составления энергетически эффективного расписания на одном процессоре

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# Report structure

- ▶ Problem Statement
- ▶ Previous Research and Our Results
- ▶ Local Search
- ▶ Evolutionary Algorithms
- ▶ Computational Experiment

$1|r_j, d_j|E$

## Input Data

$J = \{1, \dots, n\}$  is the set of jobs.

$W_j$  is the volume of job  $j$ .

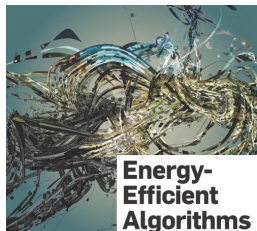
$r_j$  is the release date of job  $j$ .

$d_j$  is the deadline of job  $j$ .

Preemptions are disallowed.

## Agreeable Release Dates and Deadlines

For any two jobs  $i$  and  $j$ , relation  $r_i < r_j$  implies  $d_i \leq d_j$ .

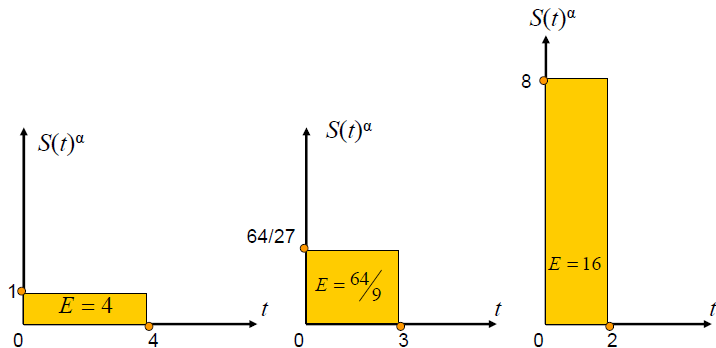


## Homogeneous Model in Speed-scaling

If a processor runs at speed  $s$  then the energy consumption is  $s^\alpha$  units of energy per time unit, where  $\alpha > 1$  is a constant (practical studies show that  $\alpha \leq 3$ ).

It is supposed that a continuous spectrum of processor speeds is available.

The objective is to find a feasible schedule that minimizes the total energy consumption.



## Related Results: Algorithms

### Energy-Efficient Scheduling for Parallel Real-Time Tasks Based on Level-Packing

Kong F. et. al. (SAC'11): two-dimensional strip packing problem, energy consumption assignment

### Energy efficient scheduling of parallel tasks on multiprocessor computers

Li K. (Journal of Supercomputing, 2012): system partitioning, task scheduling, power supplying

### A genetic algorithm for energy-efficiency in job-shop scheduling

Salido M.A. (Int. J. Adv. Manuf. Technol., 2015): genetic algorithm with generational scheme – position-based encoding, problem specific initial population, order crossover, shuffle mutation.

## Related Results: Complexity

$1|pmtn, r_j, d_j|E$  and  $P|agree, r_j, d_j|E$

Yao, Demers, Shenker (1995):  $O(n^2)$  time;

Shioura, Shakhlevich, Strusevich (2015):  $O(n^3)$  time.

$1|r_j, d_j|E$

Antoniadis, Huang (2013): NP-hard,  $2^{5\alpha-4}$ ;

Bampis, Kononov, Letsios et. al. (2018):  $2^{\alpha-1}(1+\varepsilon)^\alpha \tilde{B}^\alpha$ .

# Preemptive and Agreeable instances

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**Algorithm 1** YDS Algorithm (Yao, Demers, Shenker), 1995

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1: While  $\mathcal{J} \neq \emptyset$ :

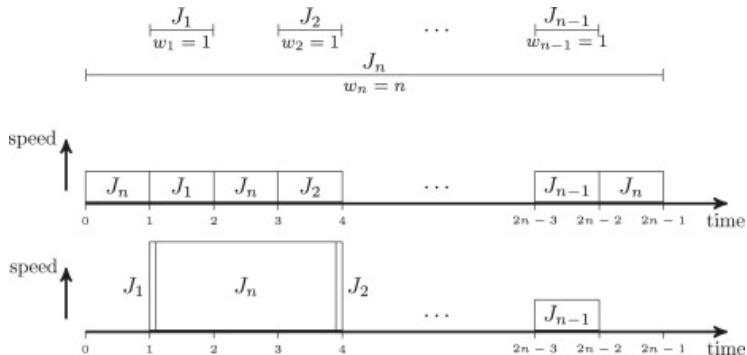
1.1 Let  $[t, t')$  be the interval with maximum density, i.e., that maximizes  $\frac{\sum_{j \in J(t, t')} W_j}{t' - t}$ .

1.2 Process jobs  $i \in J(t, t')$  in interval  $[t, t')$  using the earliest deadline policy with speed equal to the maximum density. Then remove the jobs  $J(t, t')$  from  $\mathcal{J}$ , and adjust the remaining jobs as if the time interval  $[t, t')$  does not exist.

2: Return the resulting schedule and its objective value.

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# Preemptive vs Non-Preemptive





# Small Neighborhoods

## Solution encoding

Solutions are encoded as permutations.

Consider a pair of indexes  $i < j$  and correct release dates and deadlines as follows:  $r'_j = \max\{r_i, r_j\}$  and  $d'_i = \min\{d_i, d_j\}$ .

Objective value may be calculated in  $O(n^2)$  time for the given permutation.

## Neighborhoods

*Swap* neighborhood: positions of two jobs are exchanged.

*Insert* neighborhood: inserting a job in some other position.

## Partial Order Between Jobs

Release dates and deadlines give us a partial order between jobs: if  $d_i < r_j$  then job  $i$  must precedes job  $j$ .

We exchange only independent jobs in the neighborhoods.

# Large Neighborhoods

## Optimal Recombination Problem (ORP)

Given two parent solutions  $\pi^1$  and  $\pi^2$ . It is required to find a permutation  $\pi'$  such that:

- (I)  $\pi'_i = \pi_i^1$  or  $\pi'_i = \pi_i^2$  for all  $i = 1, \dots, n$ ;
- (II)  $\pi'$  has the minimum value of objective function  $E(\pi')$  among all permutations that satisfy condition (I).


Optimal recombination may be considered as a best-improving move in a large neighbourhood defined by two parent solutions.

The ORP is NP-hard, but “almost all” instances are polynomially solvable.

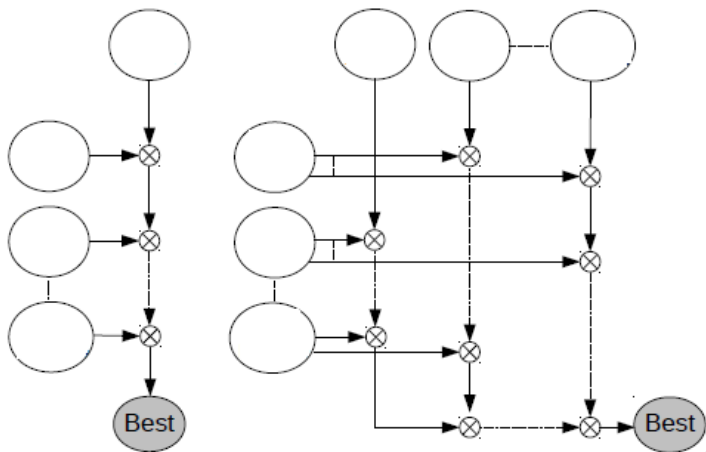
# Population Local Search (PLS)<sup>1</sup> [NUMTA2023]

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- 1: Construct the initial population of  $m$  permutations (feasible in accordance with release dates and deadlines).
  - 2: Apply local search based on swap or insert neighborhood to each permutation.
  - 3: For  $j=1$  to  $m$  perform Steps 4-6:
  - 4: Generate random sequence of permutations  $\pi^1, \dots, \pi^m$ .  
Put  $\pi' = OR(\pi^1, \pi^2)$ .
  - 5: For  $i=3$  to  $m$  construct  
$$\pi' = OR(\pi', \pi^i)$$
  - 6: Improve  $\pi'$  by local search withing swap or insert neighborhood and save as  $\pi'_j$ .
  - 7: Put  $\pi'' = OR(\pi'_1, \pi'_2)$ .
  - 8: For  $j=3$  to  $m$  construct  
$$\pi'' = OR(\pi'', \pi'_j)$$
  - 9: Return  $\pi''$  and  $E(\pi'')$ .
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<sup>1</sup>R. Tinos, D. Whitley, G. Ochoa (2020): A New Generalized Partition Crossover for the Traveling Salesman Problem: Tunneling Between Local Optima 

# Population Local Search (PLS) [NUMTA2023]

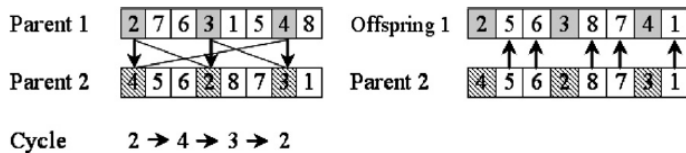


# Genetic Algorithm with OR ( $GA : ORP$ ) [NUMTA2023]

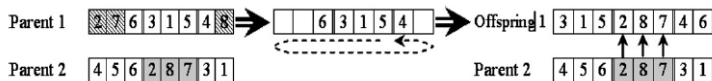
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- 1: Construct the initial population of  $m$  permutations (feasible in accordance with release dates and deadlines).
  - 2: Apply local search based on swap or insert neighborhood to each permutation.
  - 3: Until termination condition is met, perform
    - 3.1 Select two parent permutations  $\pi^1$  and  $\pi^2$ .
    - 3.2 Apply swap or insert mutation to permutations  $\pi^1$  and  $\pi^2$ .
    - 3.3 Put  $\pi' = OR(\pi^1, \pi^2)$ .
    - 3.4 Replace the worst permutation of the population by  $\pi'$ .
  - 4: Improve the record solution by local search withing swap or insert neighborhood.
  - 5: Return the best found solution.
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# Crossover Operators

## Cycle Crossover (CX)

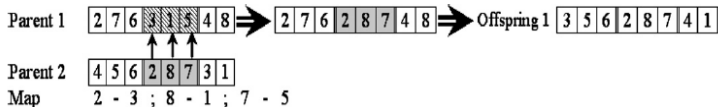


## Order Crossover (OX)

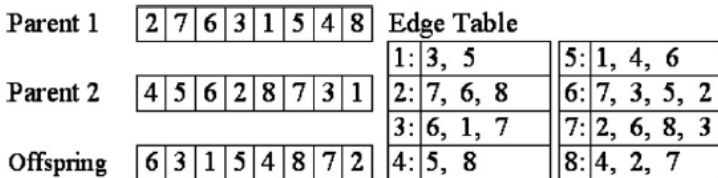


# Crossover Operators

## Partially Mapped Crossover (PMX)



## Edge Recombination (EX)



# Genetic Algorithm with Generational Scheme (GAGs)

- 1: Construct the initial population of  $m$  permutations.
- 2: Apply local search based on swap or insert neighborhood to each permutation.
- 3: Until termination condition is met, perform
  - for  $i \leftarrow 1$  to  $\beta m$ 
    - 2.1 Select two parent permutations  $\pi^1$  and  $\pi^2$ .
    - 2.2 Construct  $(\pi^{1'}, \pi^{2'}) = \text{Cross}(\pi^1, \pi^2)$ .
    - 2.3 Apply insert mutation to permutations  $\pi^{1'}$  and  $\pi^{2'}$ .
    - 2.4 Compute the objective value of the offspring.
- 4: Return the best found solution.



# Adaptive Technique

$$\phi_a = \begin{cases} w_1, & \text{if the new solution is a new global best,} \\ w_2, & \text{if the new solution is better than the current one,} \\ w_3, & \text{if the new solution is accepted,} \\ w_4, & \text{if the new solution is rejected.} \end{cases}$$

$$\rho_a = \lambda\rho_a + (1 - \lambda)\phi_a.$$

# Computational Experiment: Input Data

Number of jobs  $n = 50$  and  $n = 100$ .

Parameter  $\alpha = 2$  and  $\alpha = 3$ .

Release date  $r_j$  is selected randomly from interval  $[0, 20]$ .

Deadline  $d_j$  is generated randomly from  $[r_j + 1, r_j + 11]$ .

Volume  $W_j$  is chosen randomly from  $[5, 15]$ .

# Computational Experiment: Relative Deviation from Lower Bound

Series	<i>PLS</i>			<i>GA : ORP</i>		
	min	aver	max	min	aver	max
$S_{\alpha=2, n=50}$	0	<b>0,9</b>	6,1	0	<b>1,7</b>	8,2
$S_{\alpha=3, n=50}$	0	<b>2,1</b>	10,1	0	<b>3,2</b>	12,7
$S_{\alpha=2, n=100}$	1,7	<b>3,1</b>	6,1	2,8	<b>5,2</b>	7,9
$S_{\alpha=3, n=100}$	0,5	<b>5,8</b>	15,1	0,8	<b>8,8</b>	17,3

Series	<i>GA : Adapt</i>			<i>GA : PMX</i>		
	min	aver	max	min	aver	max
$S_{\alpha=2, n=50}$	0	<b>0,6</b>	3,1	0	<b>0,7</b>	3,7
$S_{\alpha=3, n=50}$	0	<b>1,9</b>	9,5	0	<b>2,0</b>	9,7
$S_{\alpha=2, n=100}$	1,5	<b>3,1</b>	5,5	1,7	<b>3,1</b>	5,9
$S_{\alpha=3, n=100}$	0,3	<b>5,7</b>	12,5	0,3	<b>5,9</b>	12,3

Series	<i>GA : Mut</i>		
	min	aver	max
$S_{\alpha=2, n = 50}$	0	<b>1,9</b>	9,0
$S_{\alpha=3, n = 50}$	0	<b>3,9</b>	9,8
$S_{\alpha=2, n = 100}$	3,2	<b>6,0</b>	9,5
$S_{\alpha=3, n = 100}$	2,1	<b>9,1</b>	12,7

## Conclusion and Further Research

- ▶ We proposed and investigated the evolutionary algorithm with various operators and schemes for the single processor speed scaling scheduling problem.
- ▶ Experimental evaluation on instances of different structures shown that the algorithms demonstrate competitive results.
- ▶ Further research can be undertaken to various optimized and/or randomized recombination operators and comparison of them in context of the presented algorithms and their composition; generalize to the case of several processors.

Thank you for your attention!