

Optimal Zone Breadth for Parallel and Distributed MOEA/D with Virtual Overlap Zone and Exclusively Evaluated Mating

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Background

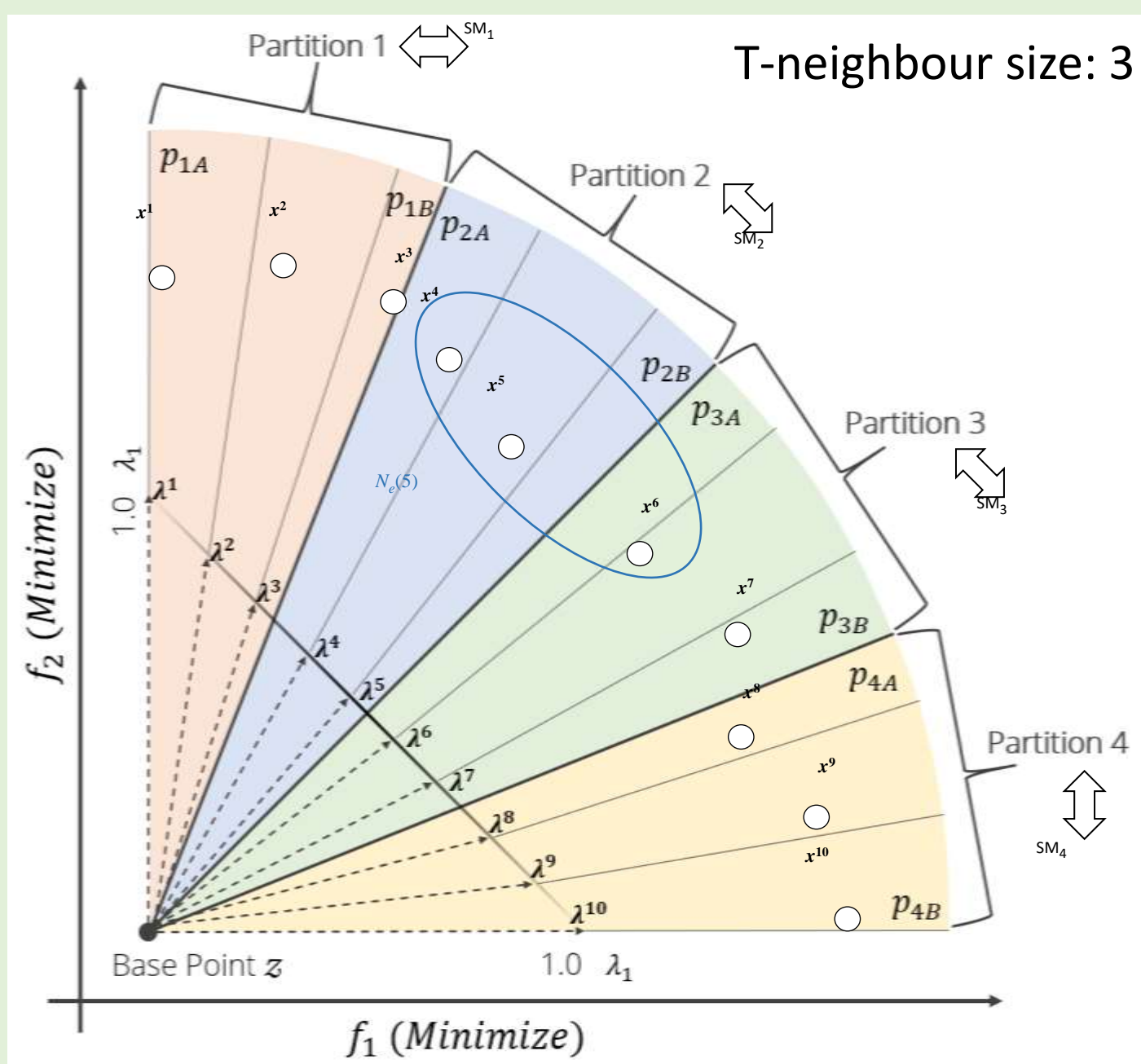


Fig. 1 Standard distribution of solutions in case of four cores.

- [1] Y. Sato, M. Sato, M. Miyakawa, "Parallel and Distributed MOEA/D with Exclusively Evaluated Mating and Migration", Proceedings of 2020 IEEE Congress on Evolutionary Computation (CEC), pp. 1-8, 2020.
[2] M. Sato, Y. Sato, M. Miyakawa, "Inconstant Update of Reference Point Value for Parallel and Distributed MOEA/D", Proceedings of 2021 IEEE Congress on Evolutionary Computation (CEC), pp. 1495-1502, 2021.

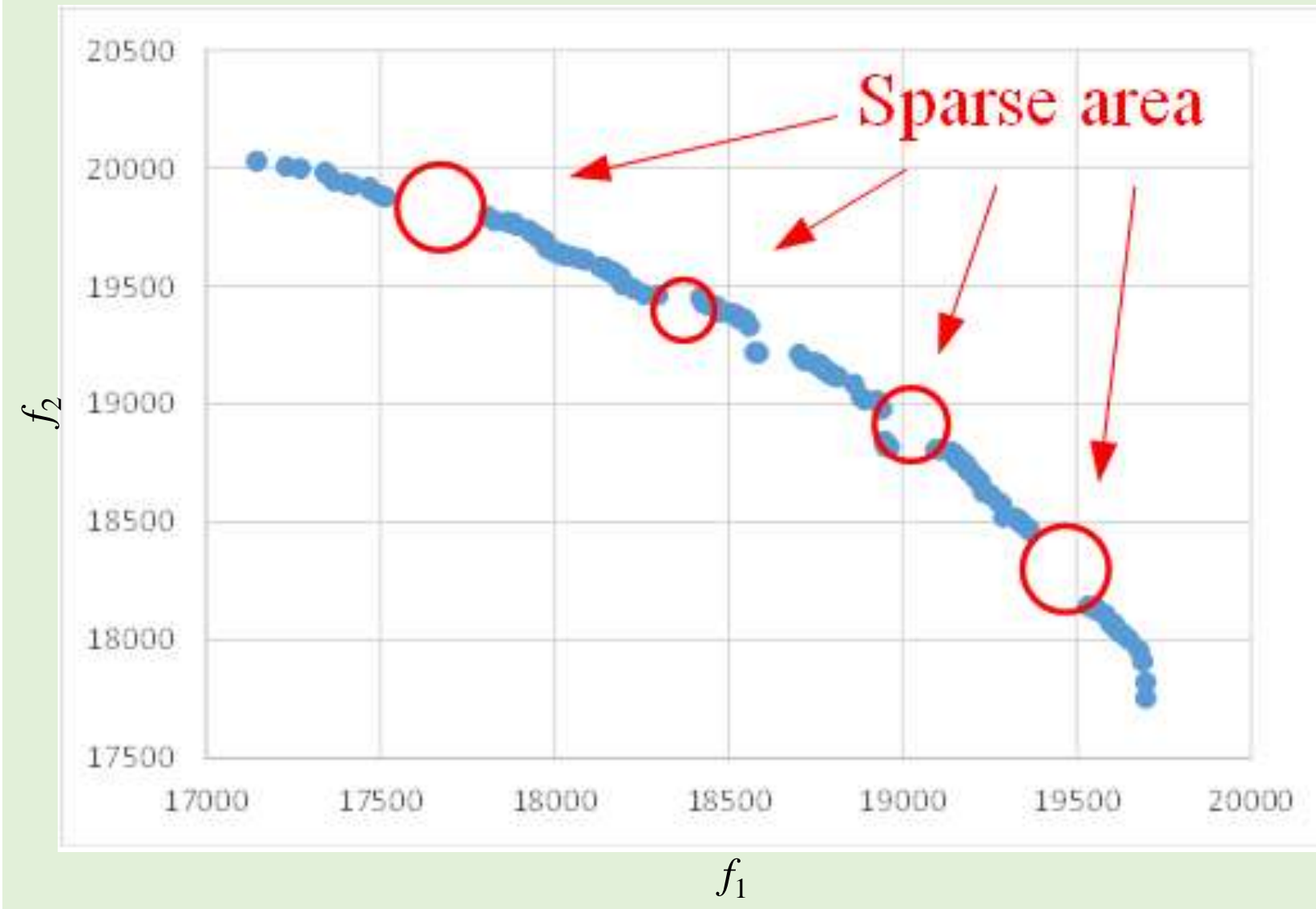


Fig. 2 Space area of Pareto optimal front in standard parallelization.

The standard parallel MOEA/D achieves acceleration by *assigning weight vectors to different computational cores* for simultaneous calculations.

- This method *substantially accelerates significantly* more than the Non-parallelized MOEA/D.
- On the other hand, *the solution distribution becomes sparse because the T-neighborhood is divided at the partition boundary* as shown in Fig. 2.

Proposed Method: Virtual Overlap Zone and Exclusively Evaluated Mating

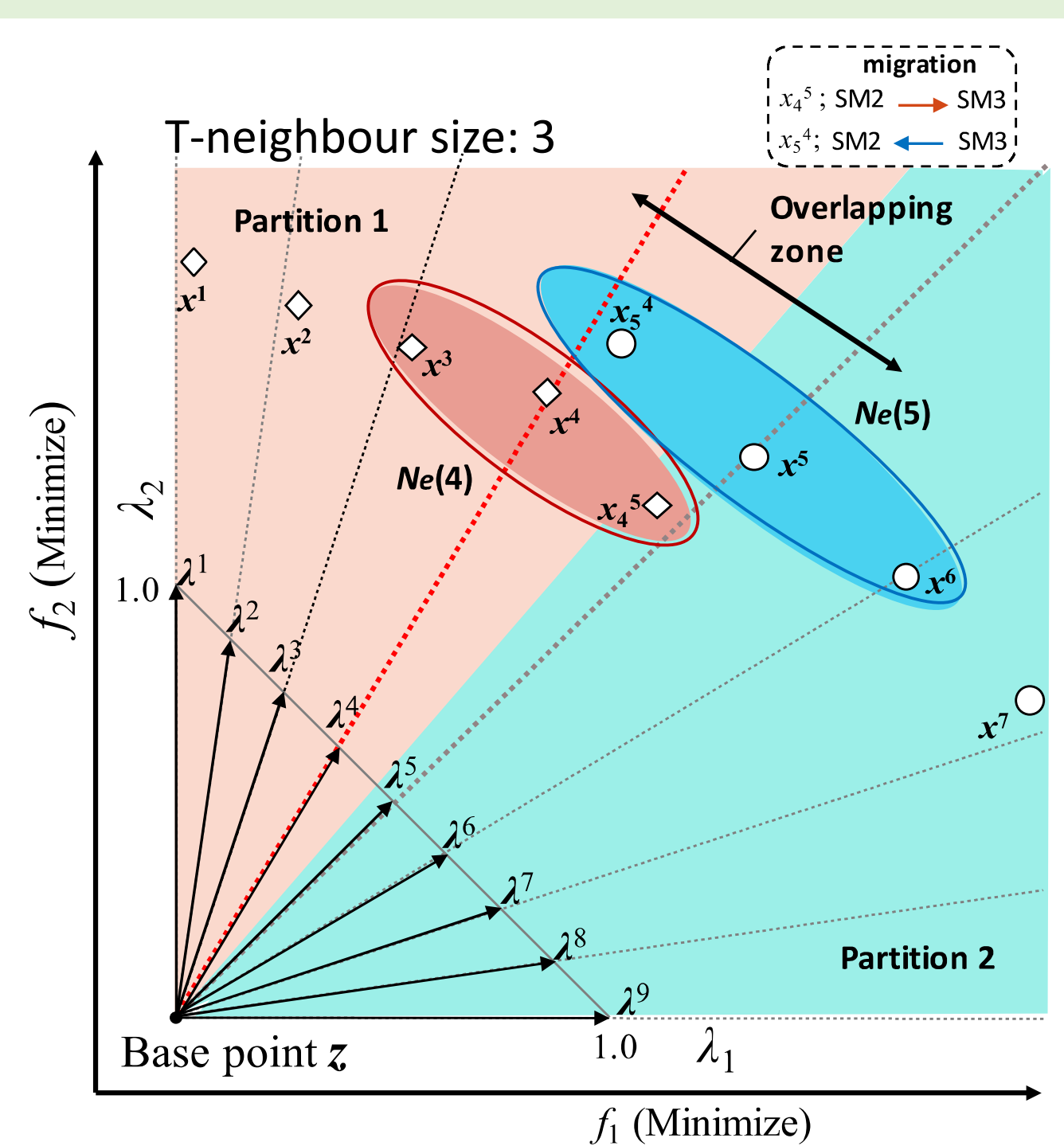


Fig. 3 Exclusively evaluated solutions inside a fold created between partitions.

- Proposed *virtual overlapping method* define the *virtual overlap zone* and *sharing weight near the boundaries between partitions* [1, 2].
- The virtual overlapping method significantly improves the sparse areas in the PF as shown in Fig. 4.
- As the *overlapping zone breadth increases*, the *HV value and execution time both tend to increase* in Fig. 5.

- Our aim is to *investigate and identify indicators for determining the overlapping size*.

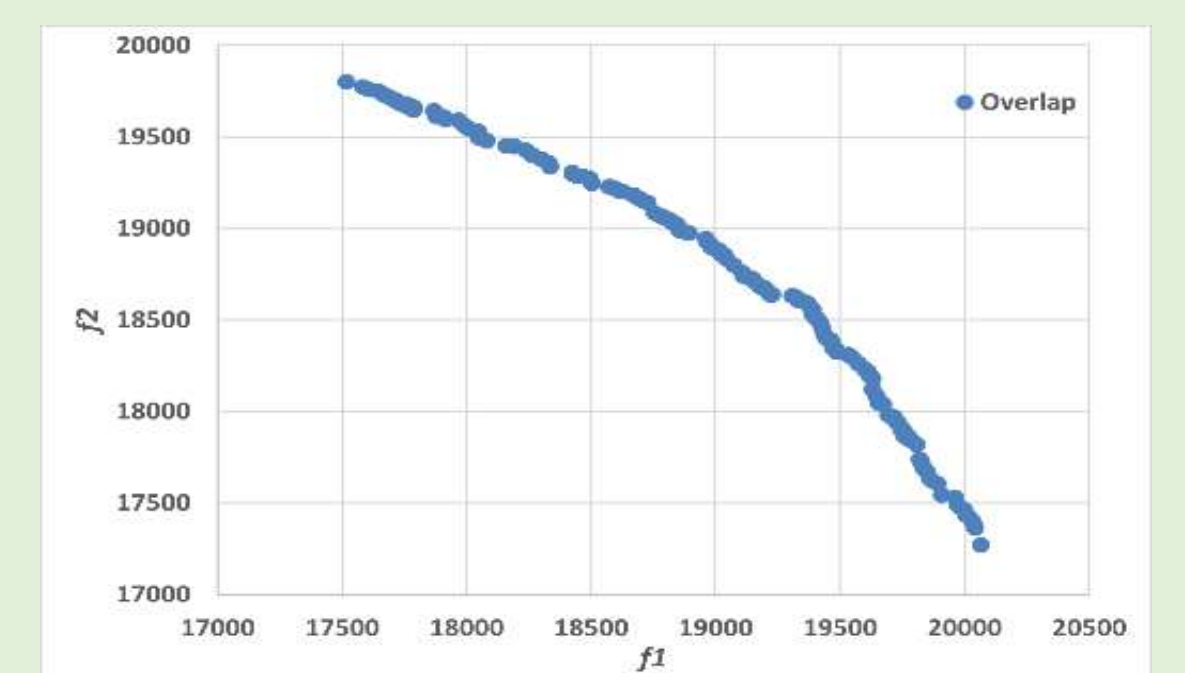


Fig. 4 Pareto optimal front in Virtual Overlapping method.

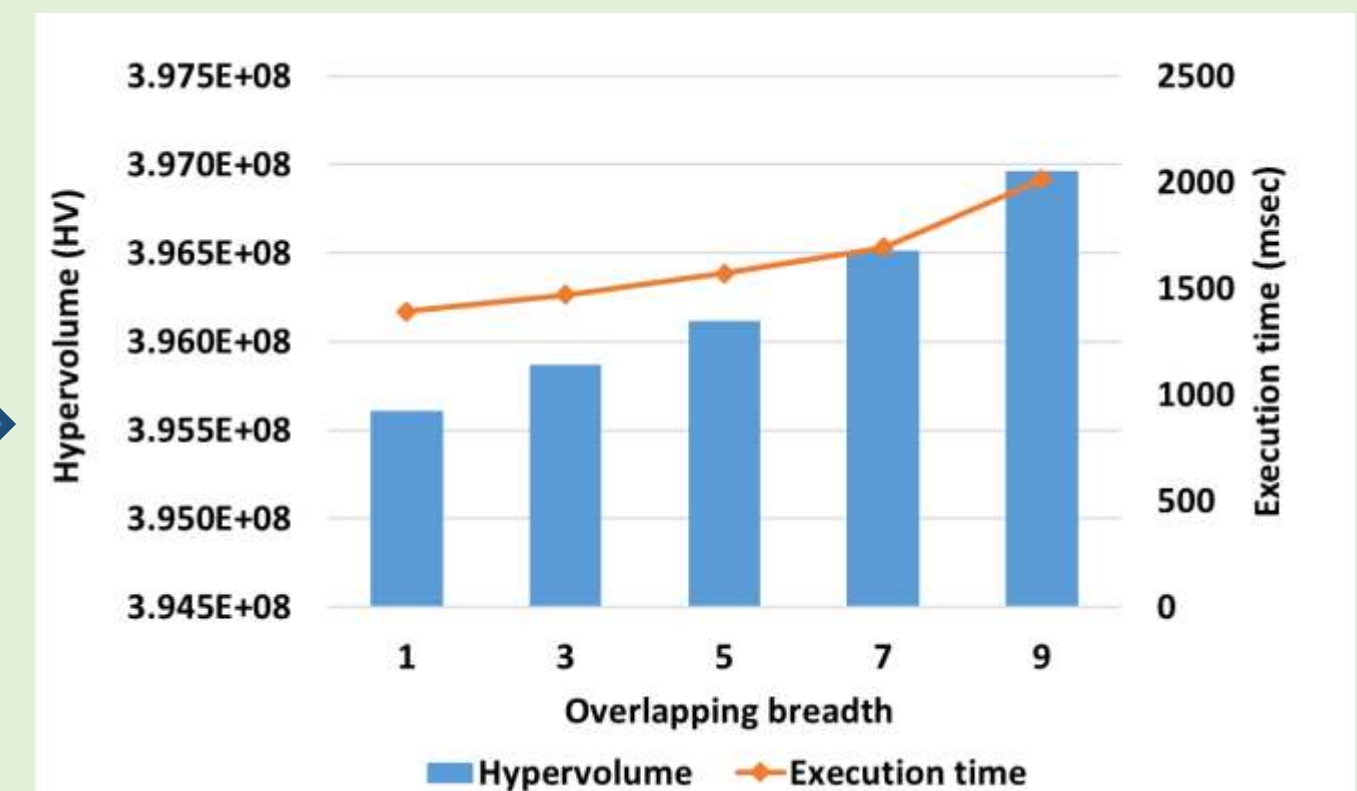


Fig. 5 HV evaluation after 2,500 generations while changing the overlapping zone breadth.

Evaluation

- We investigated the **HV value per unit time** by **varying the degree of parallelism and overlapping breadth** as parameters using the *multi-objective knapsack problem*.

$$\begin{cases} \text{Maximize } f_j(x) = \sum_{i=1}^n p_{ij}x_i \quad (j = 1, 2, \dots, m) \\ \text{Subject to } \sum_{i=1}^n w_{il}x_i \leq c_l \quad (l = 1, 2, \dots, m) \\ c_l = \varphi \sum_{i=1}^n w_{il} \quad (l = 1, 2, \dots, m) \end{cases}$$

Table 1: Experimental parameters for the knapsack problems.

Population Size	320, 480, 640
Parallelism (the number of cores)	8, 12, 16
Timeframe (unit time)	1 second
Number of Objectives	2
T-Neighborhood Size	5, 7, 9
Decomposition Method	Tchebycheff
Crossover Rate	1.0
Mutation Rate	0.05
Knapsack Problem Parameter	
Number of Items	500
Feasibility ratio	0.5

- The investigation of suitable overlapping zone breadths for T-neighborhood sizes of 5, 7, and 9 using 8, 16 cores system. (0* means "without overlapping.")

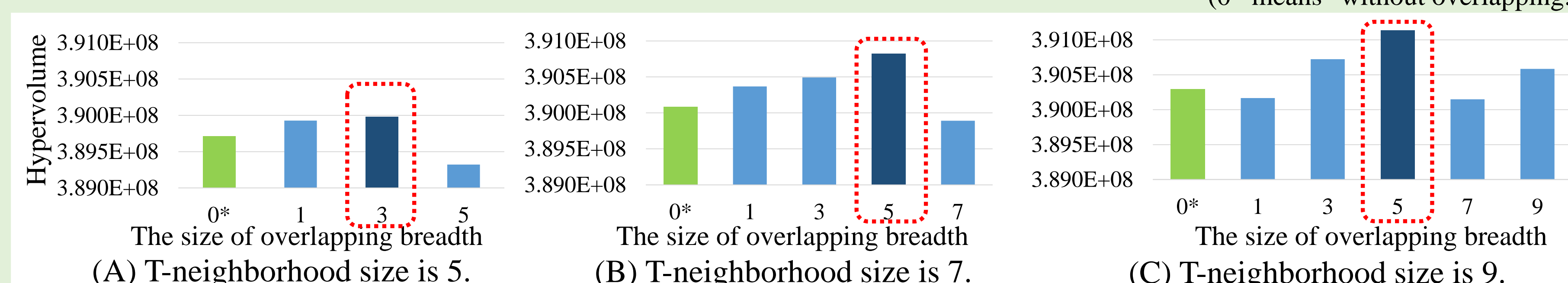


Fig. 6 Comparison of HV values using 8-cores when varying the overlapping zone breadths (1sec execution).

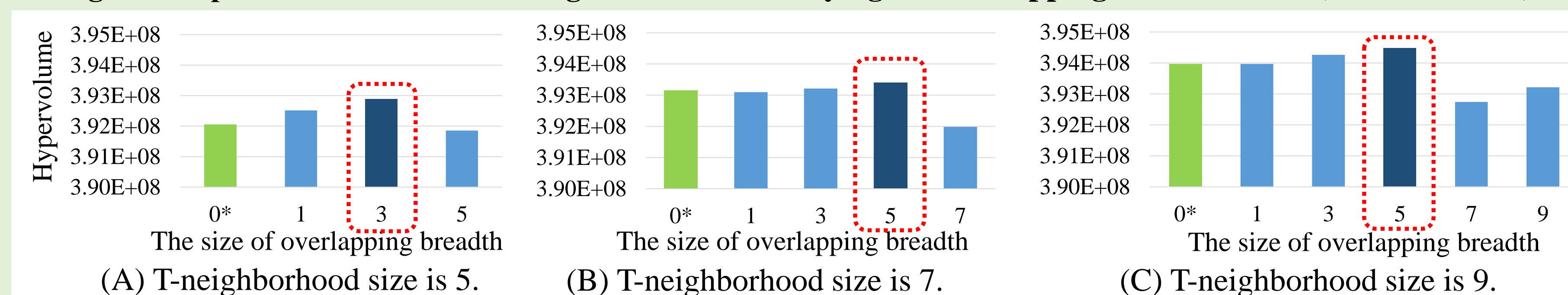


Fig. 7 Comparison of HV values using 16-cores when varying the overlapping zone breadths (1sec execution).

- The investigation of suitable overlapping zone breadths for 8, 12, and 16 cores system with T-neighborhood sizes of 9. (0* means "without overlapping.")

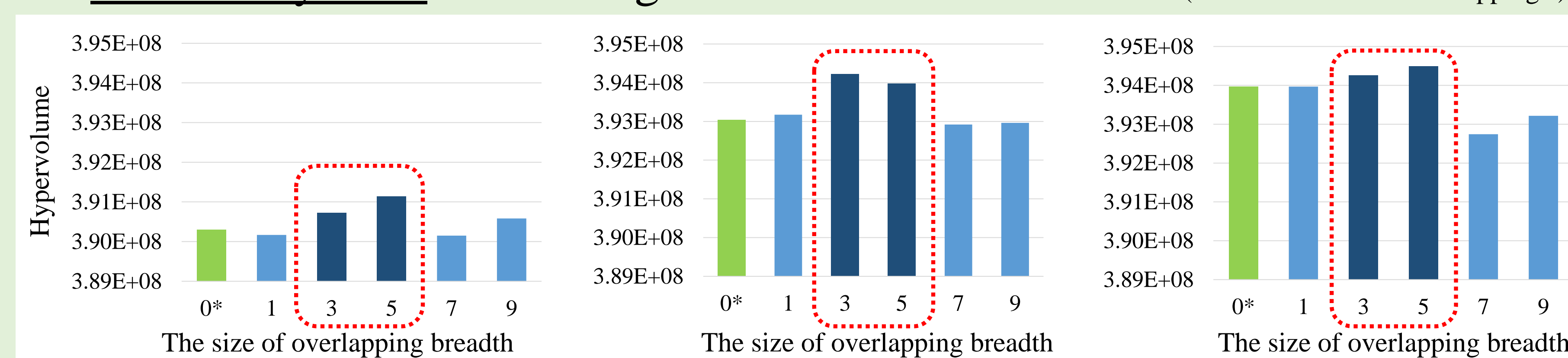


Fig. 8 Comparison of HV values for varying the overlapping zone breadths with T-neighborhood size of 9.

Conclusion

- Setting the **Overlapping Zone Breadth** to **approximately half of the T-neighborhood size** yielded the highest HV values.
- It can be also seen that with an appropriate overlapping zone breadth **setting significantly improves the sparse areas** in the PF, and **the HV values achieved by the overlapping method are higher than the standard parallelization method**.

Future works: (1)The detailed investigation of the three-objective optimization, (2) it is necessary to experiment with another application programs for real-world problems.