

## ACCURACY–TIME TRADE-OFF IN RECONSTRUCTION OF THE HEAT TRANSFER COEFFICIENT $\kappa$ USING ABC AND ACO ALGORITHMS

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The accurate reconstruction of the heat transfer coefficient  $\kappa$  under the fourth-kind boundary condition remains one of the key inverse problems in heat conduction modeling, especially in applications related to the cooling of castings and imperfect thermal contact between the casting and the mold [1,2,3,4]. In engineering practice, however, high solution accuracy alone is not sufficient. Equally important is the computational cost incurred to achieve such accuracy. This very problem serves as the starting point of this work. Instead of another classic comparison of optimization algorithm effectiveness, an assessment of their utility from the perspective of the trade-off between reconstruction error and computation time was proposed [5,6].

In the study, two swarm algorithms, Artificial Bee Colony (ABC) and Ant Colony Optimization (ACO) [7,8], were considered and applied to the identification of the  $\kappa$  coefficient in the inverse heat conduction problem using temperature data. The calculations were performed in an original numerical environment based on the finite element method. As a quality criterion, the L2 error functional and the reconstruction error of the parameter  $\kappa$  relative to the reference value were adopted. The analysis was conducted for different population sizes, numbers of iterations, and levels of noise in the reference data. A key element of the work is the comparison of reconstruction quality with the time required to obtain a solution, and the determination of computational configurations that are most cost-effective from a practical standpoint.

The adopted approach allows moving away from the simplified criterion of "minimum error" and toward a more realistic approach to identifying the parameter  $\kappa$ . In many industrial tasks, a solution that is slightly less accurate but significantly faster can be more valuable than the numerically best result, even if computationally expensive. It is expected that the conducted analysis will identify the parameter ranges in which increasing the number of iterations or population size ceases to yield a proportional improvement in quality, and will also show which of the ana-

lyzed algorithms more frequently lead to solutions that dominate in the accuracy-time trade-off. In this context, the work contributes not only a comparison of two metaheuristics but, more importantly, a practical criterion for their conscious selection for tasks involving the reconstruction of thermal contact conditions [9].

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