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Paolo Priore, Borja Ponte, Javier Puente, Alberto Gómez

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Learning-based scheduling of flexible manufacturing systems using ensemble methods

Paolo Priore^{1*}, Borja Ponte², Javier Puente¹, Alberto Gómez¹

¹ Departamento de Administración de Empresas, Escuela Politécnica de Ingeniería de Gijón, Universidad de Oviedo, {priore, jpuente, albertogomez}@uniovi.es

² Department for People and Organisations, The Open University Business School, Michael Young Building, Walton Hall, MK7 6AA, Milton Keynes, UK, borja.ponte-blanco@open.ac.uk

*corresponding author.

Abstract

Dispatching rules are commonly applied to schedule jobs in Flexible Manufacturing Systems (FMSs). However, the suitability of these rules relies heavily on the state of the system; hence, there is no single rule that always outperforms the others. In this scenario, machine learning techniques, such as support vector machines (SVMs), inductive learning-based decision trees (DTs), backpropagation neural networks (BPNs), and case based-reasoning (CBR), offer a powerful approach for dynamic scheduling, as they help managers identify the most appropriate rule in each moment. Nonetheless, different machine learning algorithms may provide different recommendations. In this research, we take the analysis one step further by employing ensemble methods, which are designed to select the most reliable recommendations over time. Specifically, we compare the behaviour of the *bagging*, *boosting*, and *stacking* methods. Building on the aforementioned machine learning algorithms, our results reveal that ensemble methods enhance the dynamic performance of the FMS. Through a simulation study, we show that this new approach results in an improvement of key performance metrics (namely, mean tardiness and mean flow time) over existing dispatching rules and the individual use of each machine learning algorithm.

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