5. CONCLUSIONS

The presence of fuzzy information has complicated the predictive analysis of complex systems under multiple constraints. For engineering applications, it is very meaningful to develop a predictive analysis method facing multiple fuzzy constraints. This paper puts forward an improved multi-constrained fuzzy predictive analysis algorithm. The author explained the normalization, weighting, granularity setting and classic domain of the attributes of multiple constraints, introduced the calculation of the fuzzy distance and fuzzy closeness for the attributes of multiple constraints, and detailed the realization of our algorithm and model multi-constrained fuzzy predictive analysis. The effectiveness and feasibility of our algorithm and model were demonstrated through comparison with relevant data in the literature. The results show that the results of our approach agree with those of the literature. The proposed algorithm and model provide a new solution and support to multi-constrained fuzzy predictive analysis.

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