

RESEARCH

Changsheng Hua

Reinforcement Learning Aided Performance Optimization of Feedback Control systems



Springer Vieweg

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