Control Theory for Reinforcement Learning — A Tutorial Session

Anders Rantzer

Lund University, Sweden

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Schedule

10:05 Nikolai Matni, University of Pennsylvania Self-tuning regulators and reinforcement learning

10:45 Alexandre Proutiere, KTH, Stockholm Controlling Unknown Discrete Systems

11:10 Anders Rantzer, Lund University Optimization based exploration/exploitation

11:35 Stephen Tu, Google Brain, New York City Concentration Bounds for System Identification

12:00 End