CS394R Reinforcement Learning: Theory and Practice

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BE a reinforcement learner

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 - Way of selecting actions that gets you the most reward

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- S = 4x3 grid
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- Reinforcement Learning in practice (end)

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- Many approaches possible (including evolutionary)
- Book focusses on a particular class of approaches

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• Available on-line

Assignments

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