





Markov Decision Processes

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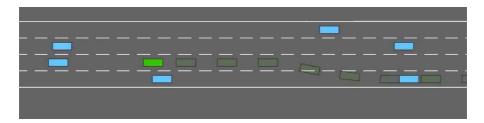
Objective: maximise $V = \mathbb{E}\left[\sum_{t=0}^{\infty} \gamma^t r_t\right]$



Motivation — Example

The highway-env environment (





We want to handle stochasticity.



Online Planning

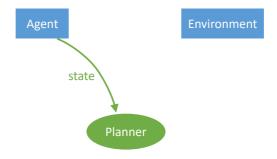
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Agent Environment

Planner

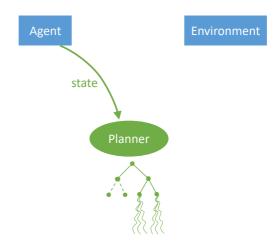


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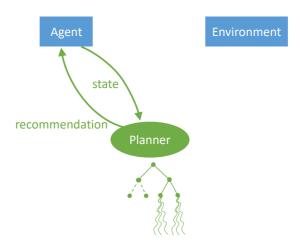


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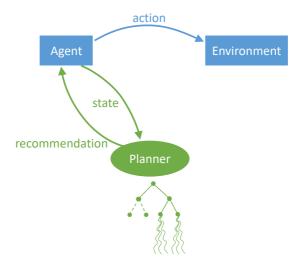


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