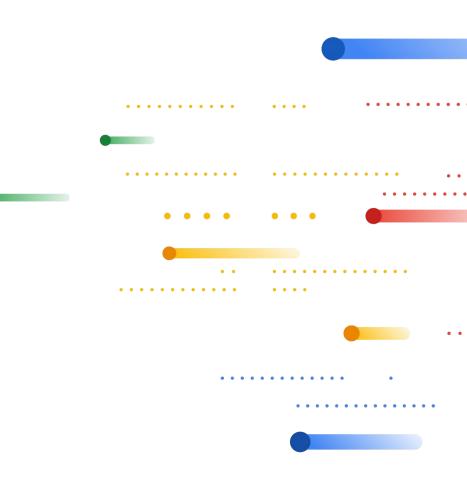
The Value Function Polytope In Reinforcement Learning

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Problematic

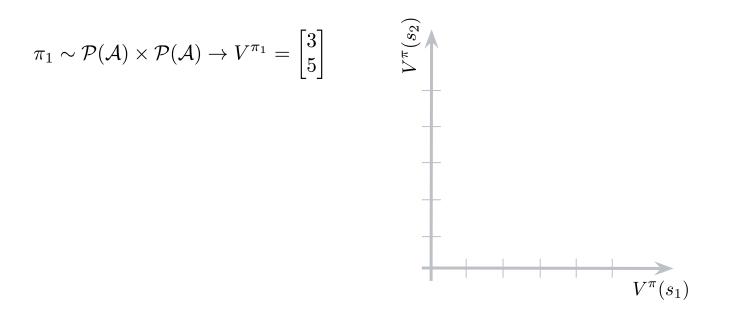
Question:

What is the geometry of the space of possible value functions for a given Markov decision process ?

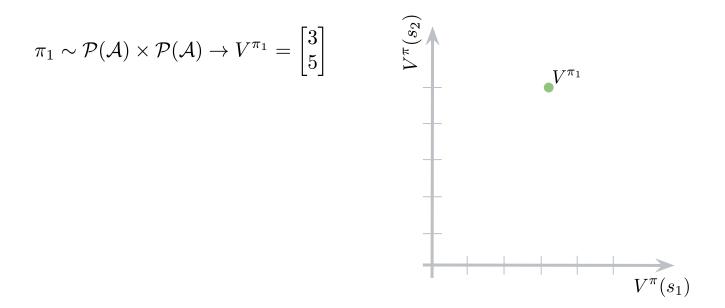
Motivation:

- Relationship between policy space and value function space
- Better understanding of the dynamics of existing algorithms
- New formalism of representation learning in RL

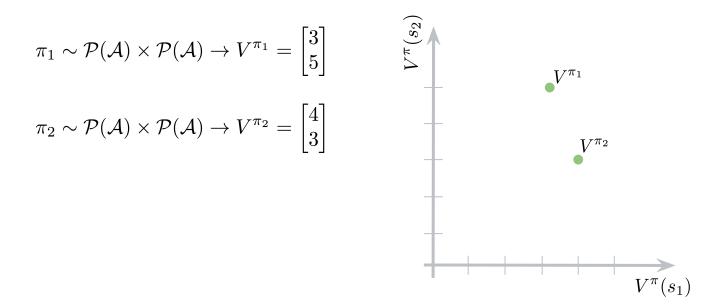
Consider a Markov decision process with 2 states: R, P, γ



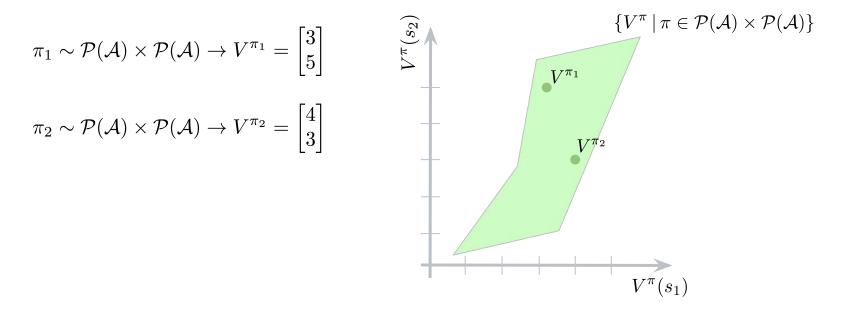
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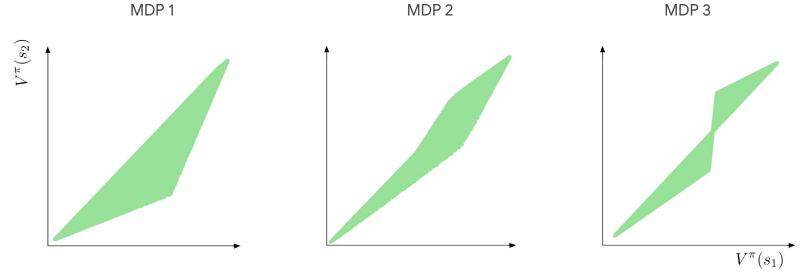
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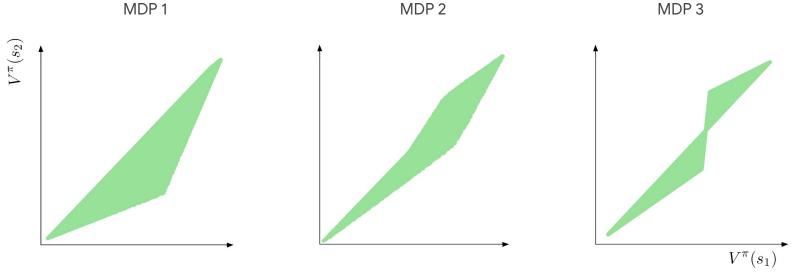


/ Main result



/ Main result

What is the geometry of the space of value functions for a given MDP?



Theorem:

The ensemble of value functions is a possibly non-convex polytope.

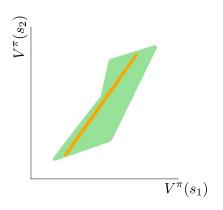
Building blocks

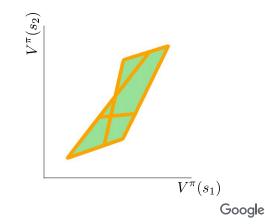
The line Theorem

The value functions of mixtures of two policies that differ in only one state describe a line in value function space.

The boundary Theorem

The boundary of the space of value functions is included in the image of the boundary of the space of policies.





/ Algorithms in the polytope

Value Iteration

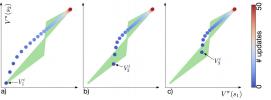


Figure 7. Value iteration dynamics for three initialization points.

Policy Gradient

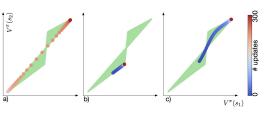


Figure 9. Value functions generated by policy gradient.

Policy Iteration

Figure 8. Policy iteration. The red arrows show the sequence of value functions (blue) generated by the algorithm.

C)

 $V^{\pi}(s_1)$

h١

Policy Gradient + entropy

 (s_1)

Figure 10. Value functions generated by policy gradient with entropy, for three different initialization points.

CEM + CEM-CN

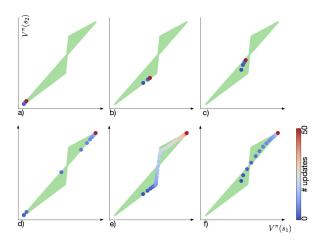


Figure 12. The cross-entropy method without noise (CEM) (a, b, c); with constant noise (CEM-CN) (d, e, f).

/ Ongoing work

- Representation learning in Reinforcement Learning
- New actor-critic algorithms



Poster # 119

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