## Long-Term Values in Markov Decision Processes, (Co)Algebraically

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- Markov decision processes (MDPs) are state-based models of sequential decision-making under uncertainty
- Applications:
  - Planning
  - Reinforcement learning
  - Insurance and finance
  - . . .
- We restrict to finite, discrete time-homogeneous, infinite-horizon MDPs, with the discounting criterion

#### INTRO - MDPs: probab., state-based systems, with rewards 3/18

Example: A start-up company needs to decide to Advertise or Save money



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A Markov Decision Process (MDP) is coalgebra  $m = \langle u, t \rangle \colon S \to \mathbb{R} \times (\Delta S)^A$  where

- S is finite set of *states* and A is finite set of *actions*
- $u \colon S \to \mathbb{R}$  is *reward* map
- $t: S \to (\Delta S)^A$  is transition map (where  $\Delta S$  is set of prob. distr. on S)

A policy  $\sigma$  is a map  $\sigma \colon S \to A$ 

#### INTRO – Trace Semantics and Long-term Value of a Policy 4/18



Given  $m = \langle u, t \rangle \colon S \to \mathbb{R} \times (\Delta S)^A$ and policy  $\sigma \colon S \to A$  $t_{\sigma} \stackrel{\text{def}}{=} t(s)(\sigma(s))$  $\rightsquigarrow m_{\sigma}^{\sharp}$  by determinization [Jacobs, Silva, Sokolova]

 $ts(s) = (r_0^{\sigma}(s), r_1^{\sigma}(s), r_2^{\sigma}(s), \cdots) \text{ is trace semantics of } m_{\sigma}$  $(r_n^{\sigma}(s) \text{ is expected reward at time } n, \text{ starting from } s)$ 

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Discounting criterion: letting  $0 \le \gamma < 1$  be a *discount factor*, the long-term value of policy  $\sigma$  is  $V^{\sigma} : S \to \mathbb{R}$ :

$$\mathbf{V}^{\sigma}(s) = r_0^{\sigma}(s) + \gamma \cdot r_1^{\sigma}(s) + \dots + \gamma^n \cdot r_n^{\sigma}(s) + \dots$$

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5/18

The optimal value function  $\mathrm{V}^*\colon S\to\mathbb{R}$  of m in state s is given by

$$\mathbf{V}^*(s) = \max_{\sigma} \, \mathbf{V}^{\sigma}(s)$$

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We define:

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Classical facts (cf. (Puterman, 2014)):

- Optimal policy always exists
- Optimal policies need not be unique
- If  $\sigma$  is optimal, then  $\mathbf{V}^{\sigma}=\mathbf{V}^{*}$
- Stationary (memoryless), deterministic policies suffice

## Observations:

- Classic theory uses low-level, analytic methods
- MDPs are coalgebras
- Goal: to develop coalgebraic methods for reasoning about LTVs

The main **contributions**:

Part 1: Value function  $V^{\sigma}$  from *b*-corecursive algebras

Part 2: Coinductive proof of policy improvement theorem

## **PART** 1:

## Value Function Arises from a Universal Property

#### PART 1 – $V^{\sigma}$ as Coalgebra-to-Algebra Morphism

• Following Bellman, the value function has a natural recursive structure:

 $(\mathbf{V}^{\sigma} \text{ from today}) = \text{reward today} + \gamma \cdot (\mathbf{V}^{\sigma} \text{ from tomorrow})$ 

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Our observation: we can re-express (1) as V<sup>σ</sup> being a coalgebra-to-algebra morphism, as in



where  $\alpha_{\gamma} \colon \mathbb{R} \times \mathbb{R} \to \mathbb{R}$  is  $\alpha_{\gamma}(x, y) = x + \gamma \cdot y$  and  $\mathbf{E} \colon \Delta \mathbb{R} \to \mathbb{R}$  is EV

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- Question: is  $\alpha_{\gamma} \circ (id_{\mathbb{R}} \times E)$  a corecursive algebra?
- Consider a more basic question: is algebra α<sub>γ</sub>: ℝ × ℝ → ℝ corecursive?
  By (Capretta et al., 2004), this is equivalent with α<sub>γ</sub> ∘ (id<sub>ℝ</sub> × E) corec.

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• In particular, let X be a set of "variables"  $\{x_0, x_1, x_2, \ldots\}$ , and fix  $\gamma$ .

$$\begin{cases} (3) & \underbrace{X_n = a_n + \gamma \cdot x_{n+1}}_{(n = 0, 1, 2, \ldots)} \end{cases} & \longleftrightarrow & \text{coalgebra } f \colon X \to \mathbb{R} \times X \\ & \underbrace{x_n = a_n + \gamma \cdot x_{n+1}}_{(n = 0, 1, 2, \ldots)} \end{cases} & \longleftrightarrow & f^{\dagger} \text{ s.t. } f^{\dagger} = \alpha_{\gamma} \cdot (\text{id}_{\mathbb{R}} \times f^{\dagger}) \cdot f \end{cases}$$

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• However, if  $(a_n)_n$  is bounded then (3) has a unique **bounded** solution

• To get uniqueness  $\Rightarrow$  incorporate boundedness information

#### PART 1 – Need for Boundedness

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- **Definition** A *b*-category (C, B) is a category C with a subcollection of "bounded" morphism B s.t.  $(f \in B \Rightarrow f \circ g \in B)$

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- **Definition** Let  $(C, \mathcal{B})$  *b*-category, *F* endofunctor on C,  $\alpha \colon FA \to A$  an *F*-algebra. Then  $\alpha$  is a *b*-corecursive algebra (*bca*) if



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- **<u>Proposition</u>**  $\alpha_{\gamma}$  is a bca in (Met, B) for H
- Then by b-version of (Capretta et al., 2004) result:
  <u>Proposition</u> α<sub>γ</sub> ∘ (ℝ × E) is a bca in (Met, B) for H ∘ Δ

#### Part 1: Value function $V^{\sigma}$ from *b*-corecursive algebras $\checkmark$

Part 2: Coinductive proof of policy improvement theorem

## **PART 2:**

Correctness of Policy Iteration via Contraction (Co)Induction

• Suppose  $\sigma$  is policy  $\rightsquigarrow$  find  $V^{\sigma}$ 

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Theorem (Howard, 1960)

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- If  $\sigma' = \sigma$ , then  $\sigma$  is optimal.
- Policy Iteration: start with any  $\sigma$ , iteratively obtain  $\sigma', \sigma'', \sigma''', \ldots$ , and continue until there is fixpoint  $\Rightarrow$  this outputs an *optimal policy*

### Policy Improvement Theorem (Howard, 1960)

 $P_{\sigma'} \cdot \mathbf{V}^{\sigma} \ge P_{\sigma} \cdot \mathbf{V}^{\sigma} \quad \Rightarrow \quad \mathbf{V}^{\sigma'} \ge \mathbf{V}^{\sigma}$ 

**Policy Improvement Theorem** (Howard, 1960)  $P_{\sigma'} \cdot V^{\sigma} \ge P_{\sigma} \cdot V^{\sigma} \implies V^{\sigma'} \ge V^{\sigma}$ 

Recall

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- $\Rightarrow$  antecedent holds
- New proof using Contraction (Co)Induction

## Contraction (Co)Induction Theorem

Let M be a non-empty, complete, ordered (i.e., there is partial order  $\leq$  s.t.  $\uparrow x = \{y \mid x \leq y\}$  and  $\downarrow x = \{y \mid y \leq x\}$  are closed) metric space. If  $f: M \to M$  is contractive and order-preserving, then the (unique) fixpoint  $x^*$  is

- least pre-fixpoint (if  $f(x) \le x$ , then  $x^* \le x$ ),
- greatest post-fixpoint (if  $x \leq f(x)$ , then  $x \leq x^*$ ).

Cf. Metric Coinduction (Kozen & Ruozzi, 2009) and (Denardo, 1967).

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Proof of Policy Improvement (i.e.,  $P_{\sigma'} \cdot V^{\sigma} \ge P_{\sigma} \cdot V^{\sigma} \Rightarrow V^{\sigma'} \ge V^{\sigma}$ ). Apply theorem to  $\Psi_{\pi} \colon \mathbb{R}^{S} \to \mathbb{R}^{S}$  (contractive and order-preserving  $\checkmark$ )

$$\Psi_{\pi}(v) = u + \gamma P_{\pi}v,$$
 and  $V^{\pi}$  is its fixpoint.

 $\text{Thus} \quad P_{\sigma'} \cdot \mathbf{V}^{\sigma} \geq P_{\sigma} \cdot \mathbf{V}^{\sigma} \ \Rightarrow \ \Psi_{\sigma'}(\mathbf{V}^{\sigma}) \geq \Psi_{\sigma}(\mathbf{V}^{\sigma}) = \mathbf{V}^{\sigma} \ \Rightarrow \ \mathbf{V}^{\sigma'} \geq \mathbf{V}^{\sigma} \,.$ 

## Contributions:

- Value functions  $\mathbf{V}^{\sigma}$  and  $\mathbf{V}^{*}$  from b-corecursive algebras
- Coinductive proof of policy improvement theorem

Future work:

- Generalize the setting (e.g., to stochastic games)
- Make connections with related literature:

Combining semantics of computation and game theory (Pavlovic, 2009) Coalgebraic formulation of infinite games (Abramsky & Winschel, 2017) Open games (Hedges, Ghani, Winschel, Zahn, 2018)

• Investigate contraction coinduction further and look for other applications (e.g., in social choice)

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