## Policy improvement

$$greedy(\pi) \ge \pi$$

 $\pi_1$ 

$$\pi_1 \leq \operatorname{greedy}(\pi_1)$$

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Does this improvement process converge?

Yes!

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 (2)



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$$\Longrightarrow \boxed{\operatorname{greedy}(\pi_*) = \pi_*}$$

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### Goal

Irrespective of the dynamics of the MDP, find the policy that maximize the discounted reward sum (value) for all states in the MDP.

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Method: iterative greedy policy improvement

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