

# GRIDWORLD & The Value function

BLOF 498/598 P1

# Markov Decision Process (MDP)

States	$S$	where we are
actions	$a$	what we choose (next state)
rewards	$r$	gain/loss from $(s, a)$

Markov Property : a state tells us everything we need to know (memoryless)

$s_0, a_0, r_1, s_1, a_1, r_2, s_2, \dots, s_T$

# Making Decisions in an MDP

Assume: MDP is deterministic  
Process terminates (@ time  $T$ )

Let's say we are at state  $s_0$ .

max Reward

$$= \max_a \left\{ \underbrace{r_1}_{\text{immediate reward}} + \underbrace{\sum_{i=2}^T r_i}_{\text{future reward}} \right\} = \max_a \left\{ r_1 + \underbrace{V(s_1)}_{\text{Value Function}} \right\}$$

# Gridworld

13	14	15	<sup>finish</sup> 16
9	10	11	12
5	6	7	8
<sup>start</sup> 1	2	3	4

Compute  $V(s)$   
for every square

If I know  $V(s)$ ,  
then I can  
always find the  
shortest path.

get from start to finish in the  
fewest number of steps.

How do I compute  $V(s)$ ?

$r_i = -1$  except at the finish.

States  $S = \{1, 2, \dots, 16\}$

↑  
start

↑  
terminal

13	14	15	16
9	10	11	12
5	6	7	8
1	2	3	4

actions are defined for every state

Approach: Monte Carlo Method

1. Pick a state.
2. Take a random walk  
until I reach the  
finish
3. Count steps.
4. Repeat.

5, 6, 10, 9, 5, 6, 7, 11, 15, 16

$$V(5) = -9$$

$V(s)$

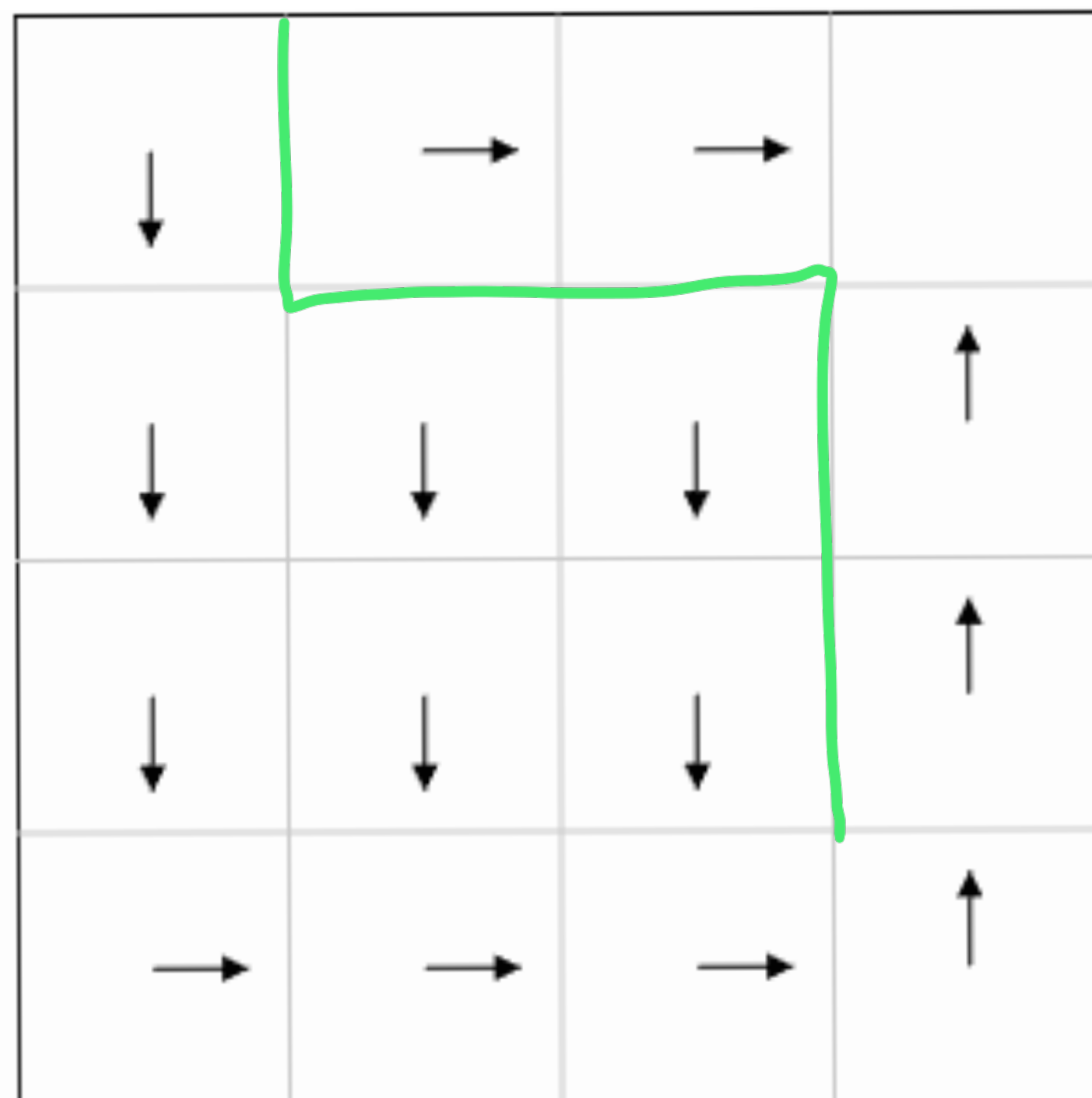
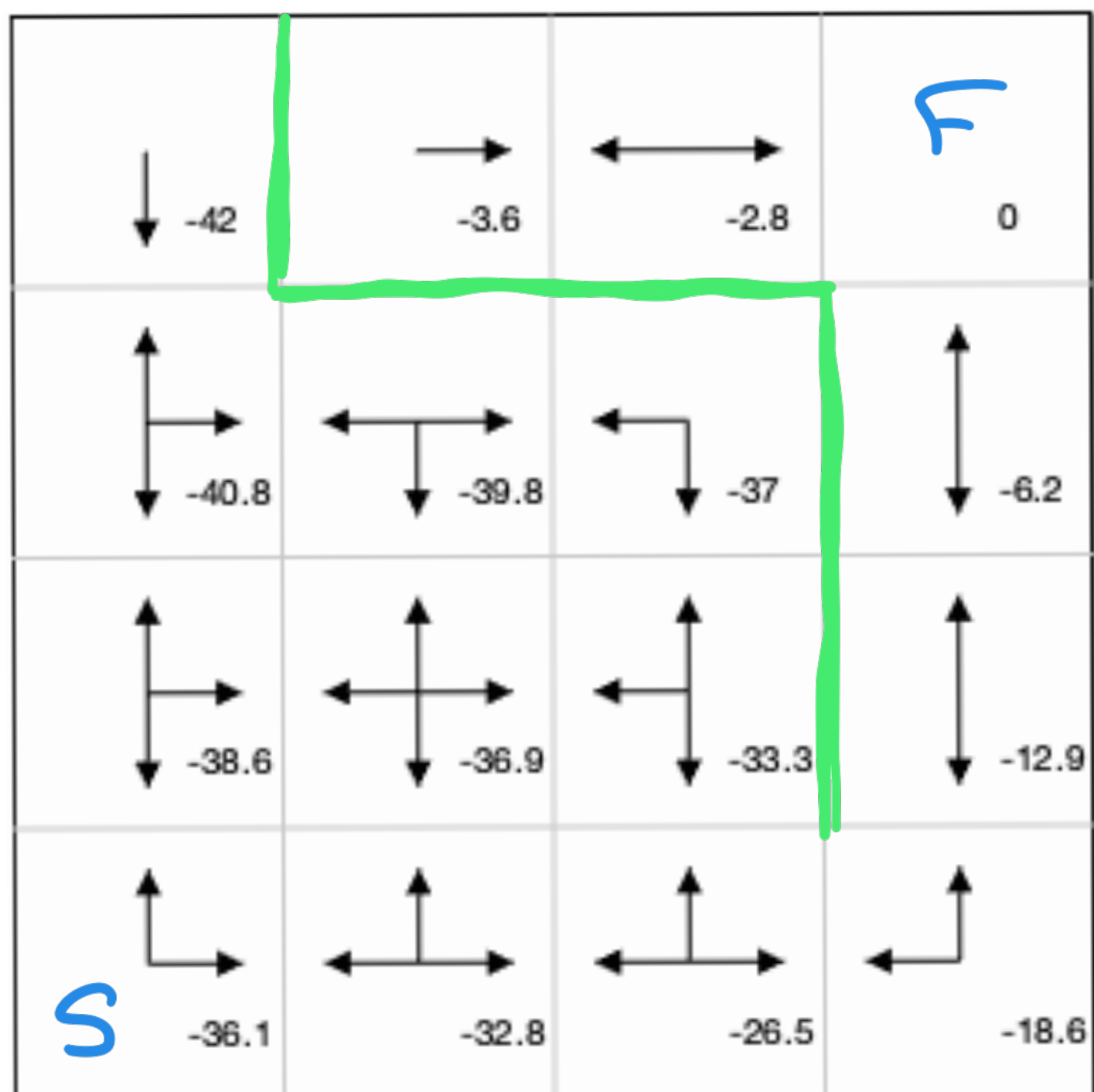
			Finish 0

start

Policy Improvement




$V(s)$



$V(s)$

