

# HW3: Solving MDPs

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## 1 Part 1

--- gen\_simple\_world ---

rewards

1.00	-1.00	0.00	0.00
0.00	-1.00	0.00	0.00
0.00	0.00	0.00	0.00
0.00	0.00	0.00	0.00

visualize a random policy

.	.	v	>
<	.	v	^
v	>	<	<
v	^	<	^

--- value iteration ---

Values from Value Iteration

0.00	0.00	0.42	0.44
0.77	0.00	0.45	0.48
0.71	0.59	0.55	0.51
0.66	0.62	0.58	0.54

Optimal Policy

.	.	>	v
^	.	>	v
^	v	<	<
^	<	<	<

--- policy evaluation ---

Optimal Policy

.	.	>	v
^	.	>	v
^	v	<	<
^	<	<	<

Values from Policy Iteration

0.00	0.00	0.42	0.44
0.77	0.00	0.45	0.48
0.71	0.59	0.55	0.51
0.66	0.62	0.58	0.54

--- gen\_simple\_world 2 ---

rewards

1.00	-1.00	0.00	5.00
0.00	-1.00	0.00	0.00
0.00	0.00	0.00	0.00
-1.00	0.00	2.00	0.00

visualize a random policy

.	v	>	.
^	^	v	>
^	>	^	^
.	v	v	v

--- value iteration ---

Values from Value Iteration

0.00	13.05	13.68	0.00
13.41	16.71	18.24	18.19
13.84	18.76	20.72	19.76
0.00	20.94	21.40	21.18

Optimal Policy

.	>	v	.
v	v	v	v
^	>	v	v
.	>	v	<

--- policy evaluation ---

Optimal Policy

.	>	v	.
v	v	v	v
^	>	v	v
.	>	v	<

Values from Policy Iteration

0.00	13.05	13.68	0.00
13.41	16.71	18.24	18.19
13.84	18.76	20.72	19.76
0.00	20.94	21.40	21.18

--- gen\_simple\_world 3 ---

rewards

5.00	-5.00	-2.00	-2.00
-2.00	-2.00	-2.00	-2.00
-2.00	-2.00	-2.00	-100.00
-2.00	-2.00	-2.00	10.00

visualize a random policy

```
.      v      ^      >
^      ^      v      v
<      >      <      >
<      v      >      .
```

--- value iteration ---

Values from Value Iteration

0.00	3.68	-2.23	-4.37
4.01	0.80	-1.73	-5.59
0.86	0.70	-0.97	-1.94
0.86	3.98	8.22	0.00

Optimal Policy

```
.      <      <      <
^      <      <      ^
^      v      <      <
>      >      >      .
```

--- policy evaluation ---

Optimal Policy

```
.      <      <      <
^      <      <      ^
^      v      <      <
>      >      >      .
```

Values from Policy Iteration

0.00	3.68	-2.23	-4.37
4.01	0.80	-1.73	-5.59
0.86	0.70	-0.97	-1.94
0.86	3.98	8.22	0.00

--- gen\_simple\_world 4 ---

rewards

5.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00
-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00
-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00
-1.00	-1.00	-1.00	-5.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00
-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00
-1.00	-1.00	-1.00	-1.00	-1.00	-100.00	-1.00	-1.00	-1.00	-1.00
-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00
-1.00	-1.00	2.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00
-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	5.00	-1.00
-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	-1.00	10.00

visualize a random policy

```

.      v      <      v      v      v      <      >      >      <
v      ^      >      <      >      >      >      ^      v      v
^      ^      v      ^      >      <      <      ^      <      v
<      >      ^      ^      v      <      >      <      >      ^
>      ^      v      v      v      >      >      ^      >      >
<      >      <      ^      <      v      ^      >      >      <
<      >      ^      <      v      <      <      <      v      ^
v      v      >      >      ^      v      >      >      v      <
v      <      >      <      >      <      <      ^      ^      ^
v      v      v      ^      >      v      <      >      ^      .

```

--- value iteration ---

Values from Value Iteration

0.00	4.04	1.86	0.05	-1.47	-2.73	-3.61	-3.29	-2.87	-2.72
4.04	2.15	0.49	-0.98	-2.23	-3.18	-3.03	-2.25	-1.68	-1.54
1.86	0.49	-0.89	-2.16	-3.10	-3.04	-2.06	-0.98	-0.26	-0.17
0.05	-0.95	-1.98	-2.86	-3.26	-2.30	-0.82	0.53	1.42	1.42
-1.44	-2.05	-1.89	-2.36	-2.95	-2.75	0.67	2.30	3.42	3.23
-2.31	-1.46	-0.47	-1.13	-2.00	0.62	2.44	4.36	5.84	5.29
-1.47	-0.14	1.39	0.32	0.84	2.47	4.49	6.74	8.78	7.59
-0.53	1.35	0.43	1.62	2.33	4.37	6.74	9.44	12.46	10.13
-1.33	0.04	1.59	1.57	3.43	5.84	8.78	12.46	10.50	12.83
-2.13	-1.05	0.14	1.44	3.23	5.29	7.59	10.13	12.83	0.00

Optimal Policy

```

.      <      <      <      <      <      <      v      v      v
^      ^      <      <      <      <      v      v      v      v
^      ^      <      ^      <      >      v      v      v      v
^      ^      <      <      >      >      v      v      v      v
^      ^      v      v      v      ^      >      v      v      v
>      v      v      v      <      v      >      v      v      v
>      >      v      v      v      v      v      v      v      v
>      >      >      <      >      >      >      v      v      v
>      >      ^      >      >      >      >      >      >      <
>      >      >      >      >      >      >      >      ^      .

```

--- policy evaluation ---

Optimal Policy

```

.      <      <      <      <      <      <      v      v      v
^      ^      <      <      <      <      v      v      v      v
^      ^      <      ^      <      >      v      v      v      v
^      ^      <      <      >      >      v      v      v      v
^      ^      v      v      v      ^      >      v      v      v
>      v      v      v      <      v      >      v      v      v
>      >      v      v      v      v      v      v      v      v
>      >      >      <      >      >      >      >      v      v
>      >      ^      >      >      >      >      >      v      <
>      >      >      >      >      >      >      >      ^      .

```

### Values from Policy Iteration

0.00	4.04	1.86	0.05	-1.47	-2.73	-3.61	-3.29	-2.87	-2.72
4.04	2.15	0.49	-0.98	-2.23	-3.18	-3.03	-2.25	-1.68	-1.54
1.86	0.49	-0.89	-2.16	-3.10	-3.04	-2.06	-0.98	-0.26	-0.17
0.05	-0.95	-1.98	-2.86	-3.26	-2.30	-0.82	0.53	1.42	1.42
-1.44	-2.05	-1.89	-2.36	-2.95	-2.75	0.67	2.30	3.42	3.23
-2.31	-1.46	-0.47	-1.13	-2.00	0.62	2.44	4.36	5.84	5.29
-1.47	-0.14	1.39	0.32	0.84	2.47	4.49	6.74	8.78	7.59
-0.52	1.35	0.43	1.62	2.33	4.37	6.74	9.44	12.46	10.13
-1.33	0.04	1.59	1.57	3.43	5.84	8.78	12.46	10.50	12.83
-2.13	-1.05	0.14	1.44	3.23	5.29	7.59	10.13	12.83	0.00

---

## 2 Part 2

### 2.1 Analysis of `gen_simple_world()`

#### 2.1.1 Code Changes:

- $\gamma = 0.95$  (encourages long-term planning).
- **Noise:** 0.1 (introduces some randomness in movement).
- **Terminal states:**  $[0, 1, 5]$ .
- **Rewards:**

$$\begin{bmatrix} 1.00 & -1.00 & 0.00 & 0.00 \\ 0.00 & -1.00 & 0.00 & 0.00 \\ 0.00 & 0.00 & 0.00 & 0.00 \\ 0.00 & 0.00 & 0.00 & 0.00 \end{bmatrix}$$

#### 2.1.2 Model Behavior:

- The agent tries to reach the highest reward while avoiding penalties.
- The **optimal policy** is:

$$\begin{bmatrix} \cdot & \cdot & \rightarrow & \downarrow \\ \uparrow & \cdot & \rightarrow & \downarrow \\ \uparrow & \downarrow & \leftarrow & \leftarrow \\ \uparrow & \leftarrow & \leftarrow & \leftarrow \end{bmatrix}$$

#### Explanation:

- The agent **moves right** initially since it avoids the  $-1$  penalty at  $[1]$ .
- It then moves downwards to collect the most reward while avoiding unnecessary risks.
- The policy is stable, meaning the agent converges to this behavior consistently.

—

## 2.2 Analysis of `gen_simple_world2()`

### 2.2.1 Code Changes:

- $\gamma = 0.95$  (encourages long-term planning).
- Increased **noise** to 0.2 (more randomness).
- Changed **terminal states** to  $[0, 3, 12]$ .
- Adjusted **rewards**:

$$\begin{bmatrix} 1.00 & -1.00 & 0.00 & 5.00 \\ 0.00 & -1.00 & 0.00 & 0.00 \\ 0.00 & 0.00 & 0.00 & 0.00 \\ -1.00 & 0.00 & 2.00 & 0.00 \end{bmatrix}$$

### 2.2.2 Model Behavior:

- The +5 reward at [3] attracts the agent strongly.
- The +2 reward at [14] also influences movement but to a lesser extent.
- The **optimal policy** is:

$$\begin{bmatrix} \cdot & \rightarrow & \downarrow & \cdot \\ \downarrow & \downarrow & \downarrow & \downarrow \\ \uparrow & \rightarrow & \downarrow & \downarrow \\ \cdot & \rightarrow & \downarrow & \leftarrow \end{bmatrix}$$

#### Explanation:

- The agent moves right aggressively toward [3] to maximize rewards.
  - It follows a downward path through [14] to collect the additional +2.
  - Increased **noise** makes actions less deterministic, meaning the agent sometimes explores more than in `gen_simple_world()`.
- 

## 2.3 Analysis of `gen_simple_world3()`

### 2.3.1 Code Changes:

- Reduced  $\gamma$  to 0.85 (short-term planning favored).
- Added a -100 penalty at [11] (huge obstacle).
- Defined **terminal states** as  $[0, 15]$ .
- Adjusted **rewards**:

$$\begin{bmatrix} 5.00 & -5.00 & -2.00 & -2.00 \\ -2.00 & -2.00 & -2.00 & -2.00 \\ -2.00 & -2.00 & -2.00 & -100.00 \\ -2.00 & -2.00 & -2.00 & 10.00 \end{bmatrix}$$

### 2.3.2 Model Behavior:

- The agent strongly avoids [11] due to the huge penalty.
- The goal state at [15] attracts movement.
- The **optimal policy** is:

$$\begin{bmatrix} . & \leftarrow & \leftarrow & \leftarrow \\ \uparrow & \leftarrow & \leftarrow & \uparrow \\ \uparrow & \downarrow & \leftarrow & \leftarrow \\ \rightarrow & \rightarrow & \rightarrow & . \end{bmatrix}$$

#### Explanation:

- The agent starts moving leftward in the first two rows, avoiding negative rewards.
- In the third row, the agent takes a downward step at [2, 1] before continuing left, demonstrating an alternate risk-averse strategy.
- In the last row, the agent moves rightward towards the goal state at [15], avoiding the high penalty at [11].
- Since  $\gamma = 0.85$ , the agent prioritizes safer short-term rewards over long-term optimization.
- The agent **completely avoids state** [11], confirming that the -100 penalty effectively deters it from that path.

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## 2.4 Analysis of gen\_simple\_world4() (10x10 Grid)

### 2.4.1 Code Changes:

- **Larger state space** ( $10 \times 10$  grid).
- Added -100 penalty at [5, 5] (major obstacle).
- Increased **noise** to 0.15 (more randomness).
- Defined **terminal states** as [0, 99].
- $\gamma = 0.9$  (encourages long-term planning).
- Adjusted **rewards**:

$$\begin{bmatrix} 5.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 \\ -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 \\ -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 \\ -1.00 & -1.00 & -1.00 & -5.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 \\ -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 \\ -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -100.00 & -1.00 & -1.00 & -1.00 & -1.00 \\ -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 \\ -1.00 & -1.00 & 2.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 \\ -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & 5.00 & -1.00 \\ -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & -1.00 & 10.00 \end{bmatrix}$$

### 2.4.2 Model Behavior:

- The agent avoids [5, 5] at all costs.
- The goal at [9, 9] is prioritized.
- The **optimal policy** is:

$$\begin{bmatrix} \cdot & \leftarrow & \leftarrow & \leftarrow & \leftarrow & \leftarrow & \leftarrow & \downarrow & \downarrow & \downarrow \\ \uparrow & \uparrow & \leftarrow & \leftarrow & \leftarrow & \leftarrow & \downarrow & \downarrow & \downarrow & \downarrow \\ \uparrow & \uparrow & \leftarrow & \uparrow & \leftarrow & \rightarrow & \downarrow & \downarrow & \downarrow & \downarrow \\ \uparrow & \uparrow & \leftarrow & \leftarrow & \rightarrow & \rightarrow & \downarrow & \downarrow & \downarrow & \downarrow \\ \uparrow & \uparrow & \downarrow & \downarrow & \downarrow & \uparrow & \rightarrow & \downarrow & \downarrow & \downarrow \\ \rightarrow & \downarrow & \downarrow & \downarrow & \leftarrow & \downarrow & \rightarrow & \downarrow & \downarrow & \downarrow \\ \rightarrow & \rightarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow & \downarrow \\ \rightarrow & \rightarrow & \rightarrow & \leftarrow & \rightarrow & \rightarrow & \rightarrow & \rightarrow & \downarrow & \downarrow \\ \rightarrow & \rightarrow & \uparrow & \rightarrow & \rightarrow & \rightarrow & \rightarrow & \rightarrow & \downarrow & \leftarrow \\ \rightarrow & \rightarrow & \rightarrow & \rightarrow & \rightarrow & \rightarrow & \rightarrow & \rightarrow & \uparrow & \cdot \end{bmatrix}$$

Explanation:

- The agent strictly avoids  $[5, 5]$  due to the severe -100 penalty, taking a longer but safer route.
- The goal at  $[9, 9]$  is prioritized, with movement becoming more direct in the lower rows.
- The +5 reward at  $[8, 8]$  and +2 at  $[7, 2]$  influence movement but do not outweigh reaching  $[9, 9]$ .
- Early movement is leftward, mid-movement is downward, and final movement is rightward, optimizing risk avoidance.
- Higher noise (0.15) causes occasional detours, but the main strategy remains stable.
- $\gamma = 0.9$  ensures long-term rewards are considered, preventing loops in suboptimal areas.