# COMP9414 Artificial Intelligence

## GridWorldEnv Environment User Guide

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Purpose: This guide provides comprehensive documentation for the GridWorldEnv

class used in Assignment 2

## 1 Overview

The GridWorldEnv class implements a grid world environment for reinforcement learning experiments. This guide covers installation, usage, and troubleshooting for the environment provided in env.py.

# 2 Environment Specifications

The environment implements an  $11 \times 11$  grid world with the following characteristics:

- Grid Size:  $11 \times 11$  cells
- State Space: Discrete 2D coordinates (x, y) where  $x, y \in \{0, 1, ..., 10\}$
- Action Space: 4 discrete actions
- Goal Position: (10, 10) bottom-right corner
- Starting Position: Random (avoiding obstacles and goal)
- Episode Termination: Only when goal is reached

# 2.1 Coordinate System

The environment uses a standard 2D coordinate system:

- $\circ$  Origin (0,0) is at the top-left corner
- X-axis increases downward (rows)
- Y-axis increases rightward (columns)

### 2.2 Obstacles

The environment contains 10 fixed obstacles arranged in two patterns:

### L-shaped pattern (top-left area):

 $\circ$  Positions: (2,2), (2,3), (2,4), (3,2), (4,2)

Cross pattern (centre):

 $\circ$  Positions: (5,4), (5,5), (5,6), (4,5), (6,5)

## 2.3 Action Space

Action	Value	Effect
Up	0	Decrease x by 1
Down	1	Increase x by 1
Left	2	Decrease y by 1
Right	3	Increase y by 1

### 2.4 Reward Structure

Event	Reward
Reach Goal	+25
Hit Obstacle	-10
Normal Step	-1

# 3 Installation and Setup

## 3.1 Prerequisites

Ensure you have Python 3.7+ and the following packages installed:

```
pip install numpy pygame
```

### 3.2 File Structure

Your project directory should contain:

## 4 API Reference

### 4.1 Class: GridWorldEnv

#### 4.1.1 Constructor

```
GridWorldEnv(seed=None)
```

#### Parameters:

- seed (int, optional): Random seed for reproducibility

### Example:

```
env = GridWorldEnv(seed=42) # Reproducible environment
```

### $4.1.2 \operatorname{reset}()$

```
reset() -> Tuple[int, int]
```

Resets the environment to a new episode.

### Returns:

- state: Initial position as tuple (x, y)

## Example:

```
state = env.reset() # Returns e.g., (3, 7)
```

## 4.1.3 step(action)

```
step(action: int) -> Tuple[Tuple[int, int], float, bool, dict]
```

Executes one environment step.

#### Parameters:

```
- action: Integer in \{0, 1, 2, 3\}
```

#### Returns:

- next\_state: New position (x,y)
- reward: Reward for this step
- done: Whether episode is finished
- info: Additional information (empty dict)

#### Example:

```
next_state, reward, done, info = env.step(0) # Move up
```

### $4.1.4 \quad \text{render}()$

```
render(delay=0.1, mode="human", episode=1,
learning_type="Q-learning",
availability=None, accuracy=None)
```

Displays the environment using Pygame.

### Parameters:

- delay: Time delay between frames (seconds)
- mode: Rendering mode (unused, for compatibility)
- episode: Current episode number for display

- learning\_type: Algorithm name for display
- availability: Teacher availability (0.0-1.0)
- accuracy: Teacher accuracy (0.0-1.0)

### Example:

```
env.render(delay=0.1, episode=50, learning_type="SARSA")
```

## 4.1.5 close()

```
close() -> None
```

Closes the Pygame window and cleans up resources.

#### Example:

```
env.close() # Always call when done
```

## 5 Usage Examples

## 5.1 Jupyter Notebook Setup

Since your solution must be submitted as a Jupyter notebook (.ipynb), start your notebook with:

```
# Import required libraries
  import numpy as np
  import pandas as pd
  import matplotlib.pyplot as plt
  import seaborn as sns
  from tqdm import tqdm
  # Import the environment
  from env import GridWorldEnv
9
  # Set random seed for reproducibility
11
  np.random.seed(42)
12
13
  # Configure plotting for notebook
14
  %matplotlib inline
15
  |plt.style.use('default')
```

## 5.2 Basic Episode

```
from env import GridWorldEnv
import numpy as np

# Create environment
env = GridWorldEnv(seed=42)
```

```
# Run one episode
  state = env.reset()
  done = False
  total\_reward = 0
10
  while not done:
12
       # Random action (replace with your policy)
       action = np.random.randint(0, 4)
14
15
       # Take action
16
       next_state, reward, done, _ = env.step(action)
17
       total_reward += reward
19
       # Optional: Render
20
       env.render(delay=0.1)
21
22
       state = next_state
23
24
  print(f"Episode finished with total reward: {total_reward}")
25
  env.close()
```

### 5.3 Teacher-Student Framework

```
# Rendering with teacher information
env.render(
delay=0.05,
episode=episode,
learning_type="Q-learning with Teacher",
availability=0.7, # 70% availability
accuracy=0.9 # 90% accuracy

8 )
```

## 6 Common Issues and Solutions

Issue	Solution
ModuleNotFoundError: No	Install pygame: pip install pygame
module named 'pygame'	
FileNotFoundError:	Ensure the images/ folder exists in your
'images/grid_agent.png'	project directory with all three PNG files
Display issues in WSL/SSH	Set environment variable: export
	SDL_VIDEODRIVER=dummy or run with-
	out rendering
Window not responding	Ensure you're calling env.render() in
	your loop; Pygame requires regular event
	processing
Slow training with rendering	Only render periodically (e.g., every 50-
	100 episodes) or increase the delay param-
	eter

# 7 Important Notes

- Actions that would move the agent outside grid boundaries are ignored
- Actions that would move the agent into an obstacle are ignored, but the agent still receives the obstacle penalty (-10)
- The episode only terminates when the goal is reached; there is no maximum step limit in the environment itself
- Always use seed parameter for reproducible experiments
- The rendering is optional and can be disabled for faster training

## 7.1 Jupyter Notebook Requirements

- Submit your solution as a single .ipynb file
- Ensure all cells have been run and outputs are visible
- Include markdown cells to explain your approach
- Use %matplotlib inline for plots to display in the notebook
- The notebook should be self-contained and runnable from top to bottom
- Do not use env.render() in the final submission as it requires pygame display

## 8 Integration with Assignment Tasks

When implementing your reinforcement learning algorithms:

- 1. Initialise your Q-table with dimensions: (11, 11, 4)
- 2. Access Q-values using: q\_table[state[0], state[1], action]
- 3. Track metrics as specified in the assignment requirements
- 4. The environment automatically handles boundaries and obstacles