

Environment "Move Box" Documentation

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Introduction

The document introduces an environment for multi agent study as two agents are trying to cooperatively move the box to the destination. The code is implemented in python and provided with simple interfaces. The environment is decoupled with learning method so reader can try any algorithms on.

This work is a rewrite of Coordinated Multi-Agent Object Transportation Problems (CMOTPs) problem in

Palmer, Gregory, et al. "Lenient multi-agent deep reinforcement learning." *Proceedings of the 17th International Conference on Autonomous Agents and MultiAgent Systems*. International Foundation for Autonomous Agents and Multiagent Systems, 2018.

This work extend the original domain into POMDP setting and fixed some bugs in the original domain.

Scenario Description

The task of the environment is explained as below

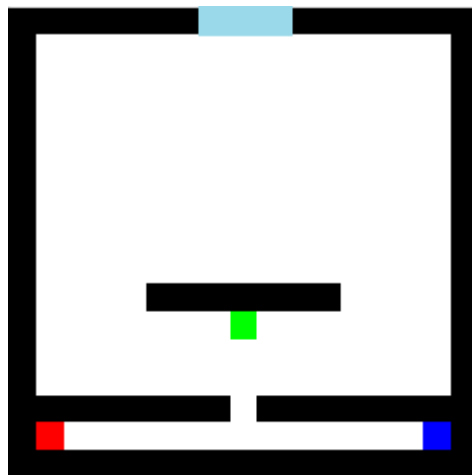


Figure 1

In the environment, two agents (agent 1 and agent2 marked using color red and blue) will firstly find to escape the smaller room at bottom, then they have to find the way to be at the right or

left side close to the box as **green**. When it is at the right or left side of box, it will be glued to the box (to carry it). Only when both the agents are at the side of the box, they enter the joint carrying mode, that when they are taking the same action (go up/down/left/right), they will carry the box to the corresponding direction. If the two agents take different actions (i.e., one goes left one goes right), then the box will not move. The goal is to move the box to the **light blue** area in Figure 1, and each agent will be rewarded for 100, otherwise reward is 0. The joint carrying is as:

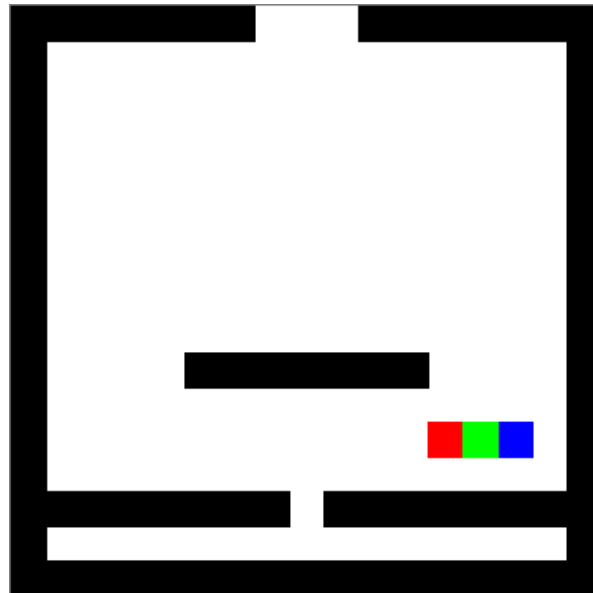


Figure 2

blocks in black are obstacles that agents could not step on and white ones are free spaces, also agents and box cannot stay at the same block. After reaching the goal area, environment will be reset to the initial state as in Figure 1.

The coordinate system is shown as

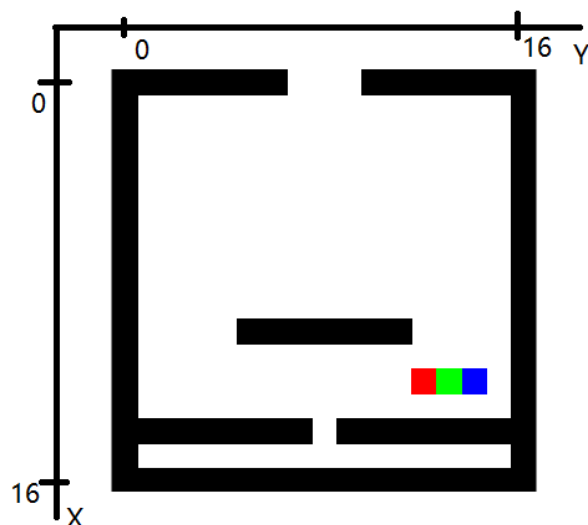


Figure 3

The area is 17*17 large, and box is initially at [11, 8], agent1 at [15, 1], agent 2 at [15, 15].

There are 4 possible actions for the two agents go up/down/left/right, using a integer as 0/1/2/3. When the agent moves, it will move to the free space of 1 cell near it. However, when there is a block or the other agent is in its way, the action won't work.

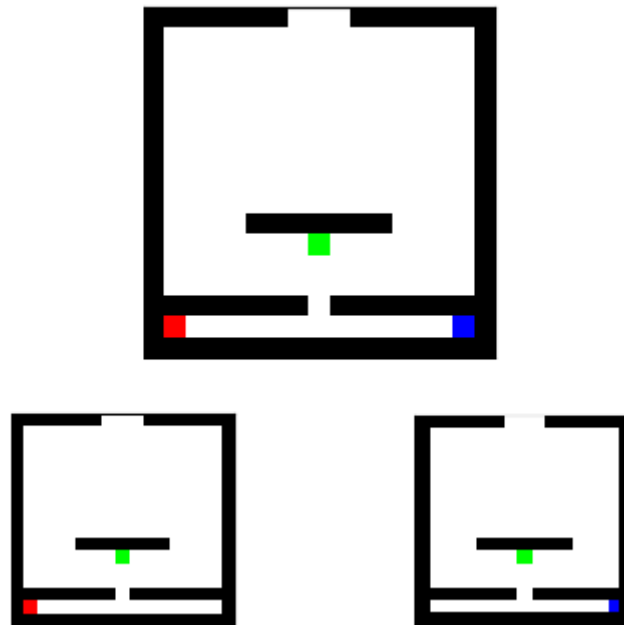


Figure4

An example of observation is shown in Figure 4. On the top is the global view with full observation, but usually we do not use it for the POMDP problem. On the bottom from left to right are observations of agent 1, agent 2. As shown in the graph, agent 1 can only observe itself and the box, so as agent 2.

The observation is an image, with form (width = 17, height = 17, rgb_channel = 3)

Class Organization

The environment is programmed in python 3.6 and has very simple interfaces. The file contains the environment is "env_MoveBox.py" and a class " EnvMoveBox" is defined.

The class has the following interfaces:

```
__init__(self)
```

```

reset(self)
obs_list = get_obs(self)
obs = get_global_obs(self)
reward_list, = step(self, action_list)
plot_scene(self)
render(self)

```

```
__init__(self)
```

initialize the object, important variables are

```
agt1_pos
```

```
agt2_pos
```

```
box_pos
```

```
get_obs(self):
```

This function return a list of current visions of each agent as an image, with form (width = 17, height = 17, rgb_channel = 3) as shown in Figure 8 bottom. obs_list = [obs1, obs2]

```
step(self, action_list):
```

The function updates the environment by feeding the actions for agent1 and agent2. The actions are denoted by integers 0/1/2/3. action_list = [action1, action2]. The return is reward_list = [reward1, reward2].

```
reset(self):
```

This function relocates the two agents and the box to the initial position

```
plot_scene(self):
```

This function plots the current state of the whole environment, and uses two subplots to show the views of each agent. This function can be useful in single step debug.

```
render(self)
```

This function plot scene in animation, call in each step

Example

Here is an example using test function, and it is in "test_MoveBox.py". The actions for agents are random chosen, click and run.

```

from env_MoveBox import EnvMoveBox
import random

```

```

env = EnvMoveBox()
max_iter = 100000
for i in range(max_iter):
    print("iter= ", i)

```

```
#env.plot_scene()
env.render()
action1 = random.randint(0, 3)
action2 = random.randint(0, 3)
reward_list = env.step([action1, action2])
print('reward', reward_list[0])
if reward_list[0] > 0:
    print('reset')
```