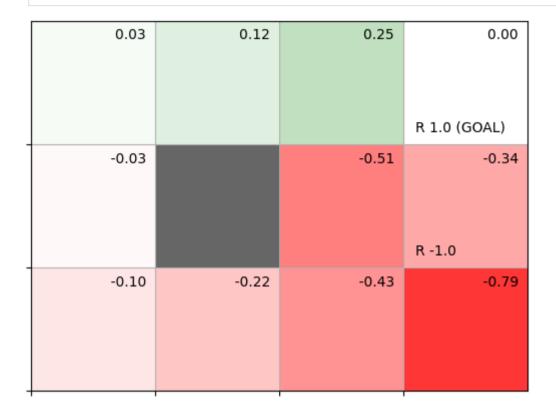
Policy Evaluation

```
In [7]:
         from collections import defaultdict
         import numpy as np
         from common.gridworld import GridWorld
         class RandomAgent:
             def __init__(self):
                 self.gamma = 0.9
                 self.batch_size = 4
                 random actions = \{0: 0.25, 1:0.25, 2:0.25, 3:0.25\}
                 self.pi = defaultdict(lambda: random actions)
                 self.V = defaultdict(lambda: 0)
                 self.cnts = defaultdict(lambda: 0)
                 self.memory = []
             def get_action(self, state):
                 action_probs = self.pi[state]
                 actions = list(action_probs.keys())
                 probs = list(action probs.values())
                 return np.random.choice(actions, p=probs)
             def add(self, state, action, reward):
                 data = [state, action, reward]
                 self.memory.append(data)
             def reset(self):
                 self.memory.clear()
             def eval(self):
                 G = \emptyset
                 for data in reversed(self.memory): # 역방향으로(reversed) 따라기
                     state, action, reward = data
                     G = self.gamma * G + reward
                     self.cnts[state] += 1
                     self.V[state] += (G - self.V[state]) / self.cnts[state]
```

```
In [8]:
        env = GridWorld()
        agent = RandomAgent()
        episodes = 1000
        for episode in range(episodes):
            state = env.reset()
            agent.reset()
            while True:
                action = agent.get_action(state)
                                                    # 행동 선택
                next_state, reward, done = env.step(action) # 행동 수행
               agent.add(state, action, reward) # (상태, 행동, 보상) 저장
                if done: # 목표에 도달 시
                   agent.eval() # 몬테카를로 방식으로 가치 함수 갱신
                           # 다음 에피소드 시작
               state = next_state
```

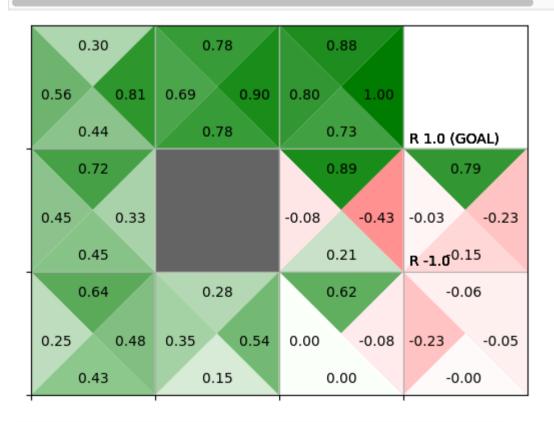
env.render_v(agent.V)



Policy control

```
In [9]:
         import numpy as np
         from collections import defaultdict
         from common.gridworld import GridWorld
         def greedy_probs(Q, state, epsilon=0, action_size=4):
             qs = [Q[(state, action)] for action in range(action size)]
             max_action = np.argmax(qs)
             base_prob = epsilon / action_size
             action_probs = {action: base_prob for action in range(action_size
             action_probs[max_action] += (1 - epsilon)
             return action_probs
         class McAgent:
             def __init__(self):
                 self.gamma = 0.9
                 self.epsilon = 0.1
                 self.alpha = 0.1
                 self.action size = 4
                 random_actions = \{0:0.25, 1:0.25, 2:0.25, 3:0.25\}
                 self.pi = defaultdict(lambda: random_actions)
                 self.Q = defaultdict(lambda: 0)
                 self.memory = []
             def get_action(self, state):
                 action_probs = self.pi[state]
                 actions = list(action_probs.keys())
                 probs = list(action_probs.values())
                 return np.random.choice(actions, p=probs)
```

```
def add(self, state, action, reward):
        data = (state, action, reward)
        self.memory.append(data)
    def reset(self):
        self.memory.clear()
    def update(self):
        G = 0
        for data in reversed(self.memory):
            state, action, reward = data
            G = self.gamma * G + reward
            key = (state, action)
            self.Q[key] += (G-self.Q[key]) * self.alpha
            self.pi[state] = greedy_probs(self.Q, state, self.epsilor
env = GridWorld()
agent = McAgent()
episodes = 10000
for episode in range(episodes):
    state = env.reset()
    agent.reset()
    while True:
        action = agent.get_action(state)
        next_state, reward, done = env.step(action)
        agent.add(state, action, reward)
        if done:
            agent.update()
            break
        state = next_state
env.render_q(agent.Q)
```



→	→	→	R 1.0 (GOAL)
1		↑	1
1	→	↑	R -1.0 ↓