```
SAI DISHA .D,231057026 number of rows=7, number of columns=9, terminal
states=[(1,2)], holes=[(2,4)], dummy states=[(2,3),(3,1),(3,3)]
import itertools
import numpy as np
import pandas as pd
import random
import matplotlib.pyplot as plt
class GridWorldEnv:
  def \underline{\phantom{a}}init\underline{\phantom{a}}(self, N = 7, M = 9):
   # States of a gridworld
    self.N = N
    self.M = M
    # state space
    self.observation space = list(itertools.product(range(self.N),
range(self.M)))
    # action space
    self.action space = [(0,1), (0,-1), (1,0), (-1,0)]
    self.terminal states = [(1,2)]
    self.holes = [(2,4)]
    self.dummy state = [(2,3),(3,1),(3,3)]
    self.reset()
  def reset(self):
    self.state = (0,0)
    self.is terminated = False
    self.total reward = 0
  def get transition probaility(self, start state, action, end state):
    if start_state in self.terminal_states:
      return 0
    if start state in self.holes:
      return 0
    expected state = tuple(np.array(start state) + np.array(action))
    if expected state == end state and end state not in
self.dummy state:
      return 1
    if expected state not in self.observation space and start state ==
end state:
      return 1
    if expected state in self.dummy state and start state ==
end state:
      return 1
```

```
return 0
  def get reward(self, start state, action, end state):
    if end state in self.terminal states:
      return 10
    else:
      return -1
  def step(self, action):
    if self.state in self.terminal_states:
      self.is terminated = True
      reward = np.nan
      return self.state, reward, self.is terminated
    current state = self.state
    \max \text{ prob} = 0
    for possible_state in self.observation_space:
      p = self.get transition probaility(current state, action,
possible_state)
      if p > max prob:
        \max prob = p
        next state = possible state
    reward = self.get reward(current state, action, next state)
    self.state = next state
    self.total_reward += reward
    return self.state, reward, self.is terminated
class RandomActionAgent:
 def __init__(self, env):
    self.env = env
  def policy(self):
    action = random.choices(self.env.action space)[0]
    return
  def train(self):
    pass
env = GridWorldEnv()
env.state
(0, 0)
class DPAgent:
  def init (self, env):
```

```
self.env = env
    self.qamma = 1
    self.v = dict(zip(self.env.observation space,
np.zeros(self.env.N*self.env.M)))
    self.is trained = False
  def policy(self):
    if not self.is trained:
      action = random.choices(self.env.action space)[0]
    else:
      s = self.env.state
      max = -np.inf
      for a in self.env.action space:
        term = 0
        for s prime in self.env.observation_space:
          term+= self.env.get transition probaility(s, a, s prime) *
(self.env.get reward(s, a, s prime) + self.gamma * self.v[s prime])
        if term > max:
          max = term
          action = a
    return action
  def train(self, iter limit = 1000):
    print("performing training...")
    self.v = dict(zip(self.env.observation space,
np.zeros(self.env.N*self.env.M)))
    iter = 0
    while iter< iter_limit:</pre>
      for s in self.env.observation space:
        max = -np.inf
        for a in self.env.action_space:
          term2 = 0
          for s prime in self.env.observation space:
            term2+= self.env.get transition probaility(s, a, s prime)
* (self.env.get reward(s, a, s prime) + self.gamma*self.v[s prime])
          if term2 > max:
            max = term2
        self.v[s] = max
      iter+=1
    self.is trained = True
    print(np.array(list(self.v.values())).reshape(self.env.N,
self.env.M))
```

Optimal state values through value iteration

```
env = GridWorldEnv(7,9)
dp agent = DPAgent(env)
dp agent.train(iter limit = 100)
env.reset()
while not env.is terminated:
 current state = env.state
 action = dp agent.policy()
 next_state, reward, is_terminated = env.step(action)
 # print(current state, action, reward, next state)
performing training...
[[ 8. 9. 10. 9. 8.
                     7. 6.
                            5.
                                4.1
 [ 9. 10. 0. 10. 9. 8. 7.
                            6. 5.]
 [8, 9, 10, 9, 0, 7, 6,
                            5. 4.1
 [7. 8. 9. 8. 5. 6. 5. 4. 3.]
 [ 6. 7. 8. 7. 6. 5. 4. 3. 2.]
 [5. 6. 7. 6. 5. 4. 3.
                            2. 1.1
 [4. 5. 6. 5. 4. 3. 2. 1. 0.]]
```

What are the average total rewards that the agent gets training?

```
env.total_reward
8
```

What are the average total rewards that the agent gets without training?

```
env2 = GridWorldEnv(7,9)
dp_agent = DPAgent(env2)

while not env2.is_terminated:
    current_state = env2.state
    action = dp_agent.policy()
    next_state, reward, is_terminated = env2.step(action)
    # print(current_state, action, reward, next_state)

env2.total_reward
-62
```

Create a heatmap of optimal state values

```
v = np.array(list(dp_agent.v.values())).reshape(env.N, env.M)
plt.imshow(v, cmap='hot')
```

```
plt.colorbar()
plt.show()
```

