

```
In [9]: import warnings ; warnings.filterwarnings('ignore')
        import gymnasium as gym, gym walk # Changed gym to gymnasium as gym is unmaint
        import numpy as np
        import random
        import warnings
        warnings.filterwarnings('ignore', category=DeprecationWarning)
        np.set printoptions(suppress=True)
        random.seed(123); np.random.seed(123)
In [3]: pip install gymnasium git+https://github.com/mimoralea/gym-walk#egg=gym-walk #
      Collecting gym-walk
        Cloning https://github.com/mimoralea/gym-walk to /tmp/pip-install-61d6pn9g/gy
      m-walk 2947cd5ce1c04fe6bc1314bd2ceea149
        Running command git clone --filter=blob:none --quiet https://github.com/mimor
      alea/gym-walk /tmp/pip-install-61d6pn9g/gym-walk 2947cd5ce1c04fe6bc1314bd2ceeal
      49
        Resolved https://github.com/mimoralea/gym-walk to commit b915b94cf2ad16f8833a
      1ad92ea94e88159279f5
        Preparing metadata (setup.py) ... done
      Requirement already satisfied: gymnasium in /usr/local/lib/python3.12/dist-pack
      ages (1.2.0)
      Requirement already satisfied: numpy>=1.21.0 in /usr/local/lib/python3.12/dist-
      packages (from gymnasium) (2.0.2)
      Requirement already satisfied: cloudpickle>=1.2.0 in /usr/local/lib/python3.12/
      dist-packages (from gymnasium) (3.1.1)
      Requirement already satisfied: typing-extensions>=4.3.0 in /usr/local/lib/pytho
      n3.12/dist-packages (from gymnasium) (4.15.0)
      Requirement already satisfied: farama-notifications>=0.0.1 in /usr/local/lib/py
      thon3.12/dist-packages (from gymnasium) (0.0.4)
      Requirement already satisfied: gym in /usr/local/lib/python3.12/dist-packages
       (from qym-walk) (0.25.2)
      Requirement already satisfied: gym-notices>=0.0.4 in /usr/local/lib/python3.12/
      dist-packages (from gym->gym-walk) (0.1.0)
      Building wheels for collected packages: gym-walk
        Building wheel for gym-walk (setup.py) ... done
        Created wheel for gym-walk: filename=gym walk-0.0.2-py3-none-any.whl size=537
      7 sha256=98a907f73e0980c04b67aa7b5a2dc2f70c2f1dbfdf20f6bbf951d078d62315fa
        Stored in directory: /tmp/pip-ephem-wheel-cache-ye92yqv5/wheels/bf/23/e5/a94b
      e4a90dd18f7ce958c21f192276cb01ef0daaf2bc66583b
      Successfully built gym-walk
      Installing collected packages: gym-walk
      Successfully installed gym-walk-0.0.2
In [4]: def print policy(pi, P, action symbols=('<', 'v', '>', '^'), n cols=4, title='
            print(title)
```

```
in [4]: def print_poticy(p1, P, action_symbols=( < , v , > , ), n_cots=4, titte=
    print(title)
    arrs = {k:v for k,v in enumerate(action_symbols)}
    for s in range(len(P)):
        a = pi(s)
        print("| ", end="")
        if np.all([done for action in P[s].values() for _, _, _, done in actic
```

```
print("".rjust(9), end=" ")
                 else:
                      print(str(s).zfill(2), arrs[a].rjust(6), end=" ")
                  if (s + 1) % n cols == 0: print("|")
 In [5]: def print state value function(V, P, n cols=4, prec=3, title='State-value func
             print(title)
             for s in range(len(P)):
                 v = V[s]
                 print("| ", end="")
                 if np.all([done for action in P[s].values() for _, _, _, done in actio
                      print("".rjust(9), end=" ")
                 else:
                      print(str(s).zfill(2), '{}'.format(np.round(v, prec)).rjust(6), en
                 if (s + 1) % n cols == 0: print("|")
In [20]: def probability success(env, pi, goal state, n episodes=100, max steps=200):
             random.seed(123); np.random.seed(123) # Removed env.seed(123)
             results = []
             for in range(n episodes):
                 state, info = env.reset(seed=123) # Added seed here
                 done, steps = False, 0
                 while not done and steps < max steps:</pre>
                      state, _, done, _, h = env.step(pi(state)) # Added _ for truncated
                      steps += 1
                  results.append(state == goal state)
             return np.sum(results)/len(results)
In [21]: def mean return(env, pi, n episodes=100, max steps=200):
             random.seed(123); np.random.seed(123) # Removed env.seed(123)
             results = []
             for in range(n episodes):
                 state, info = env.reset(seed=123) # Added seed here
                 done, steps = False, 0
                 results.append(0.0)
                 while not done and steps < max steps:</pre>
                      state, reward, done, , = env.step(pi(state)) # Added for trur
                      results[-1] += reward
                      steps += 1
                  results.append(0.0) # Added this line to match the original code's log
             return np.mean(results)
```

## FrozenLake MDP

```
In [11]: env = gym.make('FrozenLake-v1')
# P = env.env.P # Removed this line as it's no longer needed
init_state = env.reset()
goal_state = 15
LEFT, DOWN, RIGHT, UP = range(4)
```

```
Out[15]: {0: {0: [(0.333333333333333, 0, 0.0, False),
     (0.3333333333333333, 0, 0.0, False),
     1: [(0.333333333333333, 0, 0.0, False),
     (0.3333333333333333, 4, 0.0, False),
     2: [(0.333333333333333, 4, 0.0, False),
     (0.3333333333333333, 0, 0.0, False)],
    3: [(0.333333333333333, 1, 0.0, False),
     (0.333333333333333, 0, 0.0, False),
     (0.333333333333333, 0, 0.0, False),
     1: [(0.333333333333333, 0, 0.0, False),
     (0.333333333333333, 5, 0.0, True),
     2: [(0.333333333333333, 5, 0.0, True),
     3: [(0.333333333333333, 2, 0.0, False),
     (0.333333333333333, 0, 0.0, False)]},
    (0.3333333333333333, 6, 0.0, False)],
    (0.333333333333333, 6, 0.0, False),
     2: [(0.333333333333333, 6, 0.0, False),
     (0.333333333333333, 2, 0.0, False),
     (0.333333333333333, 7, 0.0, True),
     2: [(0.333333333333333, 7, 0.0, True),
     4: {0: [(0.3333333333333333, 0, 0.0, False),
     1: [(0.333333333333333, 4, 0.0, False),
     (0.333333333333333, 8, 0.0, False),
```

```
2: [(0.333333333333333, 8, 0.0, False),
 (0.3333333333333333, 5, 0.0, True),
 (0.3333333333333333, 0, 0.0, False)],
3: [(0.33333333333333, 5, 0.0, True),
 (0.3333333333333333, 0, 0.0, False),
 5: {0: [(1.0, 5, 0, True)],
1: [(1.0, 5, 0, True)],
2: [(1.0, 5, 0, True)],
3: [(1.0, 5, 0, True)]},
(0.333333333333333, 5, 0.0, True),
 1: [(0.333333333333333, 5, 0.0, True),
 (0.333333333333333, 7, 0.0, True)],
2: [(0.333333333333333, 10, 0.0, False),
 (0.333333333333333, 7, 0.0, True),
 3: [(0.333333333333333, 7, 0.0, True),
 (0.333333333333333, 5, 0.0, True)]},
7: {0: [(1.0, 7, 0, True)],
1: [(1.0, 7, 0, True)],
2: [(1.0, 7, 0, True)],
3: [(1.0, 7, 0, True)]},
(0.3333333333333333, 8, 0.0, False),
 1: [(0.333333333333333, 8, 0.0, False),
 (0.3333333333333333, 9, 0.0, False)],
(0.333333333333333, 9, 0.0, False),
 3: [(0.333333333333333, 9, 0.0, False),
 1: [(0.333333333333333, 8, 0.0, False),
 2: [(0.333333333333333, 13, 0.0, False),
 (0.333333333333333, 5, 0.0, True)],
3: [(0.333333333333333, 10, 0.0, False),
 (0.3333333333333333, 5, 0.0, True),
 (0.333333333333333, 8, 0.0, False)]},
10: {0: [(0.333333333333333, 6, 0.0, False),
 (0.3333333333333333, 9, 0.0, False),
 1: [(0.333333333333333, 9, 0.0, False),
```

```
2: [(0.333333333333333, 14, 0.0, False),
       (0.333333333333333, 11, 0.0, True),
       (0.3333333333333333, 6, 0.0, False)],
      3: [(0.333333333333333, 11, 0.0, True),
       (0.333333333333333, 6, 0.0, False),
       11: {0: [(1.0, 11, 0, True)],
      1: [(1.0, 11, 0, True)],
      2: [(1.0, 11, 0, True)],
      3: [(1.0, 11, 0, True)]},
      12: {0: [(1.0, 12, 0, True)],
      1: [(1.0, 12, 0, True)],
      2: [(1.0, 12, 0, True)],
      3: [(1.0, 12, 0, True)]},
      13: {0: [(0.333333333333333, 9, 0.0, False),
       (0.3333333333333333, 9, 0.0, False)],
      3: [(0.333333333333333, 14, 0.0, False),
       (0.3333333333333333, 9, 0.0, False),
       2: [(0.333333333333333, 14, 0.0, False),
       3: [(0.333333333333333, 15, 1.0, True),
       15: {0: [(1.0, 15, 0, True)],
      1: [(1.0, 15, 0, True)],
      2: [(1.0, 15, 0, True)],
      3: [(1.0, 15, 0, True)]}}
In [13]:
     P = env.unwrapped.P
In [14]:
     init state
Out[14]: (0, {'prob': 1})
In [17]: state, reward, terminated, truncated, info = env.step(RIGHT)
     print("state:{0} - reward:{1} - terminated:{2} - truncated:{3} - info:{4}".for
```

```
In [18]: # Adversarial Policy
         pi frozenlake1 = lambda s: {
             0: RIGHT,
             1: RIGHT,
             2: RIGHT,
             3: RIGHT,
             4: RIGHT,
             5: RIGHT,
             6: RIGHT,
             7: RIGHT,
             8: RIGHT.
             9: RIGHT,
             10:RIGHT,
             11:RIGHT.
             12:RIGHT,
             13:RIGHT,
             14:RIGHT,
             15:RIGHT #Stop
         }[s]
         print("Name:Jeevanesh")
         print("Register Number:
         print policy(pi frozenlake1, P, action symbols=('<', 'v', '>', '^'), n cols=4)
       Name: Jeevanesh
       Register Number:
       Policy:
                            > | 02
| 06
> | 10
> | 14
        00
                 > | 01
                                        > | 03
        | 04
                                        > |
                 > |
                > | 09
        I 08
                                          > |
                  | 13
In [22]: print('Reaches goal {:.2f}%. Obtains an average undiscounted return of {:.4f}.
             probability success(env, pi frozenlake1, goal state=goal state)*100,
             mean return(env, pi frozenlake1)))
```

Reaches goal 0.00%. Obtains an average undiscounted return of 0.0000.

## **Policy Evaluation**

```
In [24]: # Code to evaluate the adversarial policy
         V1 = policy evaluation(pi frozenlake1, P)
         print("Name:JEEVANESH S")
         print("Register Number:212222243002")
         print state value function(V1, P, n cols=4, prec=5)
       Name: JEEVANESH S
       Register Number:212222243002
       State-value function:
       | 00 0.0315 | 01 0.02381 | 02 0.04762 | 03
                                                    0.0
       | 04 0.03919 | | 06 0.09524 |
       | 08 0.08608 | 09 0.21905 | 10 0.2381 |
                  | 13 0.41905 | 14 0.61905 |
In [41]: print('Reaches goal with {:.2f}% probability. Obtains a {:.4f} average undisco
             probability success(env, pi frozenlake1, goal state=goal state) * 100,
            mean return(env, pi frozenlake1)))
       Reaches goal with 0.00% probability. Obtains a 0.0000 average undiscounted retu
       rn.
         Policy Improvement
In [32]: def policy improvement(V, P, gamma=1.0):
             Q = np.zeros((len(P), len(P[0])), dtype=np.float64)
            # Write your code here to improve the given policy
            for s in range(len(P)):
              for a in range(len(P[s])):
                for prob, next state, reward, done in P[s][a]:
                  Q[s][a]+=prob*(reward+gamma*V[next state]*(not done))
                  new pi=lambda s:{s:a for s, a in enumerate(np.argmax(Q,axis=1))}[s]
             return new pi
In [28]: pi 2 = policy improvement(V1, P)
         print("Name: JEEVANESH S ")
         print("Register Number: 212222243002")
         print policy(pi 2, P, action symbols=('<', 'v', '>', '^'), n cols=4)
       Name: JEEVANESH S
       Register Number: 212222243002
       Policy:
                            ^ | 02
                                      < | 03
       00
                < | 01
       04
                 <
                           | 06
                                      < |
                 ^ I 09
                            v | 10
       1 08
                                        < |
                  | 13
                           > | 14
                                        v l
In [29]: print('Reaches goal {:.2f}%. Obtains an average undiscounted return of {:.4f}.
            probability success(env, pi 2, goal state=goal state)*100,
            mean return(env, pi 2)))
```

return V

Reaches goal 100.00%. Obtains an average undiscounted return of 0.5000.

```
In [30]: V2 = policy evaluation(pi 2, P)
         print("Name:212222243002")
         print("Register Number:212222243002")
         print state value function(V2, P, n cols=4, prec=5)
       Name: 212222243002
       Register Number:212222243002
       State-value function:
       | 00 0.78049 | 01 0.65854 | 02 0.53659 | 03 0.26829 |
        | 04 0.78049 |
                         | 06 0.41463 |
        | 08 0.78049 | 09 0.78049 | 10 0.70732 |
                   | 13 0.85366 | 14 0.92683 |
In [31]: # comparing the initial and the improved policy
         if (np.sum(V1>=V2)==16):
           print("The Adversarial policy is the better policy")
         elif(np.sum(V2>=V1)==16):
           print("The Improved policy is the better policy")
         else:
           print("Both policies have their merits.")
       The Improved policy is the better policy
         Policy Iteration
In [33]: def policy iteration(P, gamma=1.0, theta=1e-10):
            random actions=np.random.choice(tuple(P[0].keys()),len(P))
            pi = lambda s: {s:a for s, a in enumerate(random actions)}[s]
            while True:
             old pi = {s:pi(s) for s in range(len(P))}
             V = policy evaluation(pi, P,gamma,theta)
             pi = policy improvement(V,P,gamma)
             if old pi == {s:pi(s) for s in range(len(P))}:
               break
            return V, pi
In [34]: optimal V, optimal pi = policy iteration(P)
In [36]: print("Name:JEEVANESH")
         print("Register Number:212222243002")
         print('Optimal policy and state-value function (PI):')
         print_policy(optimal_pi, P, action_symbols=('<', 'v', '>', '^'), n cols=4)
       Name: JEEVANESH
```

Register Number:212222243002

```
In [37]: print('Reaches goal {:.2f}%. Obtains an average undiscounted return of {:.4f}.
    probability_success(env, optimal_pi, goal_state=goal_state)*100,
    mean_return(env, optimal_pi)))
```

Reaches goal 100.00%. Obtains an average undiscounted return of 0.5000.