checkers env

March 15, 2021

```
[1]: # imports
     import numpy as np
     import matplotlib.pyplot as plt
     import random
     from collections import defaultdict
[2]: # define constants
     HEIGHT = 8
     WIDTH = 8
[3]: # intialize board
     def init_board():
         # create board environment
         board = np.zeros((HEIGHT, WIDTH)).astype(int)
         # white pieces
         board[[0,2], 1:8:2] = 1
         board[1,0:7:2] = 1
         # black pieces
         board[[5,7], 0:7:2] = 2
         board[6,1:8:2] = 2
         # return
         return board
[4]: # convert board to a hashable type
     def board_to_state(board):
         res = ''
         for i in range(WIDTH):
             for j in range(HEIGHT):
                 res += str(board[i][j])
             res += '\n'
         return res
[5]: # function to print out the checker board
     \# 0 = empty, 1 = white, 2 = black
     def print_board(board):
        for i in range(WIDTH):
             for j in range(HEIGHT):
```

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[6]: # returns all the valid actions you could take per piece
     # board = 2d array of whole state
     # player = int stating which player you are
     # piece = list of tuples of coordinate of a piece location (i,j)
     # action = (piece location, new piece loction, piece taken)
     def get_actions(board, player):
         # get all pieces
         pieces = np.argwhere(board == player)
         # initialize actions
         valid_actions = []
         # if you are white player
         if player == 1:
             # check if each action is valid
             for i, j in pieces:
                 # can go down?
                 if (i < HEIGHT-1):</pre>
                      # can go left?
                     if (j > 0):
                          # empty space
                          if board[i+1, j-1] == 0:
                              valid_actions.append(((i,j), (i+1,j-1), (-1,-1)))
                          # other player
                          elif board[i+1, j-1] != player:
                              # space to land
                              if (i+2 \le HEIGHT-1) and (j-2 \ge 0) and (board[i+2,j-2]_{\bot}
      →== 0):
                                  valid_actions.append(((i,j), (i+2,j-2), (i+1,j-1)))
                      # can go right?
                      if (j < WIDTH-1):
                          # empty space
                          if board[i+1, j+1] == 0:
                              valid_actions.append(((i,j), (i+1,j+1), (-1,-1)))
                          # other player
                          elif board[i+1, j+1] != player:
                              # space to land
                              if (i+2 \le HEIGHT-1) and (j+2 \le WIDTH-1) and
      \hookrightarrow (board[i+2,j+2] == 0):
```

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valid_actions.append(((i,j), (i+2,j+2), (i+1,j+1)))
   # if you are black player
   elif player == 2:
       # check if each action is valid
       for i,j in pieces:
           # can go up?
           if (i > 0):
               # can go left?
               if (j > 0):
                    # empty space
                    if board[i-1, j-1] == 0:
                        valid_actions.append(((i,j), (i-1,j-1), (-1,-1)))
                    # other player
                    elif board[i-1, j-1] != player:
                        # space to land
                        if (i-2 \ge 0) and (j-2 \ge 0) and (board[i-2, j-2] == 0):
                            valid_actions.append(((i,j), (i-2,j-2), (i-1,j-1)))
               # can go right?
               if (j < WIDTH-1):
                    # empty space
                    if board[i-1, j+1] == 0:
                        valid_actions.append(((i,j), (i-1,j+1),(-1,-1)))
                    # other player
                    elif board[i-1, j+1] != player:
                        # space to land
                        if (i-2 \ge 0) and (j+2 \le WIDTH-1) and (board[i-2,j+2]_{\bot}
\rightarrow == 0):
                            valid_actions.append(((i,j), (i-2,j+2), (i-1,j+1)))
   return valid_actions
```

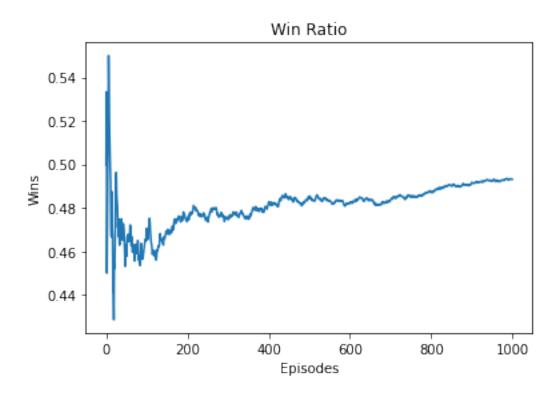
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[7]: # white player always moves randomly
     def random_move(board, player):
         # get all actions
         actions = get_actions(board, player)
         # return terminal if no moves left
         if len(actions) == 0:
             return board, False
         # select random action
         action = random.choice(actions)
         # update board
         board[action[0]] = 0
         board[action[1]] = player
         # if it was a take move
         if (action[2] != (-1,-1)):
             board[action[2]] = 0
         # return new board
         return board, True
```

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[8]: # given the current board state, and all valid actions, pick one
      # episilon greedy strategy
      def pi(q, state, actions, epsilon):
          # get all state action values
          moves = [(q[state, action], action) for action in actions]
          random.shuffle(moves)
          # randomly choose an action
          if np.random.random() < epsilon:</pre>
              return moves[0][1]
          # pick best action (random tie breaks)
          else:
              return max(moves)[1]
 [9]: # have the agent move according to a give policy
      def agent_move(board, action):
          # determine reward
          reward = 0
          # update board
          board[action[0]] = 0
          board[action[1]] = 2
          # if it was a jump move
          if (action[2] != (-1,-1)):
              board[action[2]] = 0
              reward = 1
          return board, reward
[10]: # determines a winner based on how many pieces left
      def find_winner(board):
          white = np.sum(board == 1)
          black = np.sum(board == 2)
          if white > black:
              return 1
          elif black > white:
              return 2
          else:
              return 0
[11]: # simulate an episode
      def simulate_episode(q, epsilon, alpha, gamma):
          # intialize board
          board, playing = init_board(), True
          # terminal state / no valid moves left
          player = np.random.randint(1,3) # who starts?
          while playing:
              # white move
              if player == 1:
                  board, playing = random_move(board, 1)
```

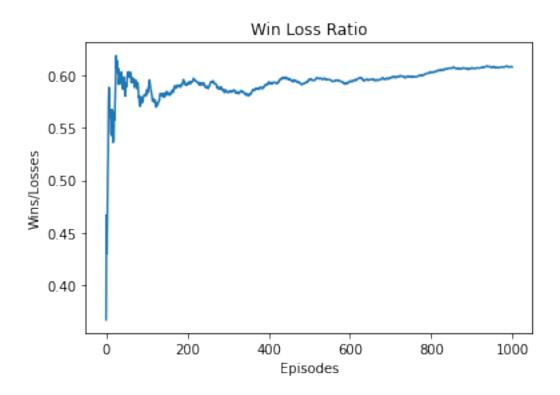
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# black move
              elif player == 2:
                  # current state
                  state = board_to_state(board)
                  # get all actions
                  actions = get_actions(board, 2)
                  # break if no moves left or white has run out of moves
                  if len(actions) == 0:
                      break
                  # select action according to pi
                  action = pi(q, state, actions, epsilon)
                  # observe R, S'
                  board, reward = agent_move(board, action)
                  next_state = board_to_state(board)
                  # update state-action value function
                  q[state, action] += alpha * (reward + gamma * max([q[next_state, a]_
       →for a in actions]) - q[state, action])
              # alternate players
              player = (player \% 2) + 1
          return find_winner(board)
[12]: # define parameters
      num_exp = 5
      num_eps = 1001
      alpha = 0.1
      gamma = 0.9
      epsilon = 0.05
[13]: # define state-action value function
      q = defaultdict(lambda: 0)
[14]: # define tracking variables
      total wins = 0
      total loss = 0
      total_ties = 0
      avg_win_ratio = np.zeros(num_eps)
      avg_win_ratio_no_ties = np.zeros(num_eps)
      # go through several experiments
      for e in range(num_exp):
          # qo through several episodes
          wins = 0
          loss = 0
          ties = 0
          win_ratio = np.zeros(num_eps)
```

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win_ratio_no_ties = np.zeros(num_eps)
   for t in range(num_eps):
       winner = simulate_episode(q, epsilon, alpha, gamma)
        if (winner == 2):
            wins += 1
        elif (winner == 1):
           loss += 1
       else:
           ties += 1
       win_ratio[t] = wins/(t+1)
       win_ratio_no_ties[t] = wins/(wins+loss+1)
    # accumulate counts
   total wins += wins
   total_loss += loss
   total_ties += ties
   # accumulate ratios
   avg_win_ratio += win_ratio
   avg_win_ratio_no_ties += win_ratio_no_ties
# average over all experiments
avg_win_ratio /= num_exp
avg_win_ratio_no_ties /= num_exp
```

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[15]: # plot win ratio
plt.plot(avg_win_ratio[1:])
plt.title('Win Ratio')
plt.xlabel('Episodes')
plt.ylabel('Wins')
plt.savefig('./q_wr.png', bbox_inches='tight')
plt.show()
```



```
[16]: # plot win ratio
plt.plot(avg_win_ratio_no_ties[1:])
plt.title('Win Loss Ratio')
plt.xlabel('Episodes')
plt.ylabel('Wins/Losses')
plt.savefig('./q_wrnt.png', bbox_inches='tight')
plt.show()
```



```
[17]: # print counts
print('Total wins:', total_wins)
print('Total loss:', total_loss)
print('Total ties:', total_ties)
```

Total wins: 2468 Total loss: 1585 Total ties: 952

1 Monte Carlo Tree Search

```
[18]: from copy import deepcopy

[19]: # randomly simulate game and return winner
def rollout(board, player):
    playing = True
    while playing:
        board, playing = random_move(board, player)
        player = (player % 2) + 1
    return find_winner(board)
```

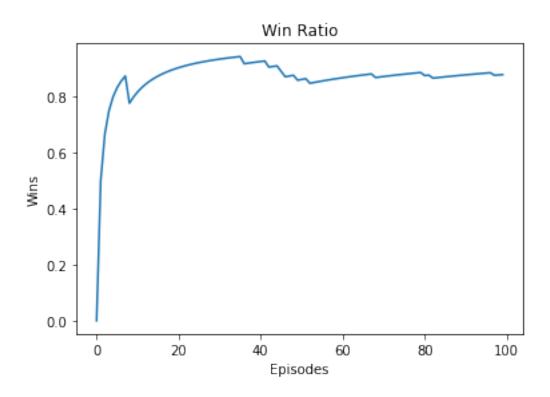
```
[20]: # use montecarlo search to find best move
      def pi(board, actions, num_trials):
          action_vals = []
          # try all actions
          for action in actions:
              # take action
              new_board = deepcopy(board)
              new_board, _ = agent_move(new_board, action)
              # estimate end reward
              reward = 0
              for i in range(num trials):
                  winner = rollout(new_board, 1)
                  if winner == 1:
                      reward -= 1
                  elif winner == 2:
                      reward += 1
              action_vals.append(reward/num_trials)
          # return best action
          return actions[np.argmax(action_vals)]
```

```
[21]: # simulate an episode
      def simulate_episode(num_trials):
          # intialize board
          board, playing = init_board(), True
          # terminal state / no valid moves left
          player = np.random.randint(1,3) # who starts?
          while playing:
              # white move
              if player == 1:
                  board, playing = random_move(board, 1)
              # black move
              elif player == 2:
                  # current state
                  state = board_to_state(board)
                  # get all actions
                  actions = get_actions(board, 2)
                  # break if no moves left or white has run out of moves
                  if len(actions) == 0:
                      break
                  # select action according to pi
                  action = pi(board, actions, num_trials)
                  # observe R, S'
                  board, _ = agent_move(board, action)
              # alternate players
              player = (player \% 2) + 1
          return find_winner(board)
```

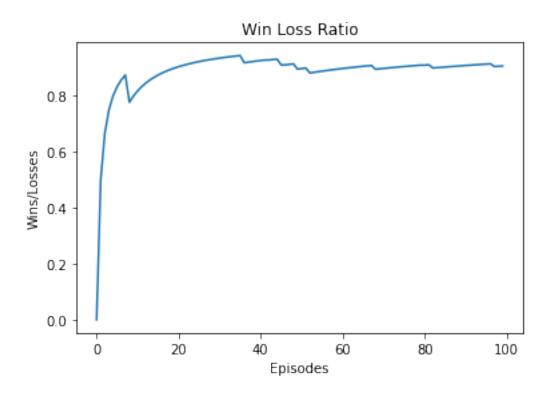
```
[22]: # define parameters
num_eps = 100
num_trials = 5
```

```
[23]: # go through several episodes
      wins = 0
      loss = 0
      ties = 0
      win_ratio = np.zeros(num_eps)
      win_ratio_no_ties = np.zeros(num_eps)
      for t in range(num_eps):
          winner = simulate_episode(num_trials)
          if (winner == 2):
              wins += 1
          elif (winner == 1):
              loss += 1
          else:
              ties += 1
          win_ratio[t] = wins/(t+1)
          win_ratio_no_ties[t] = wins/(wins+loss+1)
```

```
[24]: # plot win ratio
plt.plot(win_ratio)
plt.title('Win Ratio')
plt.xlabel('Episodes')
plt.ylabel('Wins')
plt.savefig('./mcts_wr.png', bbox_inches='tight')
plt.show()
```



```
[25]: # plot win ratio
plt.plot(win_ratio_no_ties)
plt.title('Win Loss Ratio')
plt.xlabel('Episodes')
plt.ylabel('Wins/Losses')
plt.savefig('./mcts_wrnt.png', bbox_inches='tight')
plt.show()
```



```
[26]: # print counts
print('Total wins:', wins)
print('Total loss:', loss)
print('Total ties:', ties)
```

Total wins: 88
Total loss: 8
Total ties: 4