```
import numpy as np
import random
import itertools

# blackjack

# observation (=state):
    # triple ( integer, integer )
    # 1. integer: the julyer's score (12 ~ 21)
    # 2. integer: the dealer's card score of upside (1 ~ 10)
    # 3. integer: 1 if the player has at least an ace, and 0 otherwise

# action
    # 0: hit
    # 1: stay
    # doesn't allow double down, surrender and split

# step types
STEPTYPE_FIRST = 0
STEPTYPE_MID = 1
STEPTYPE_MAST = 2

cardset = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 10, 10, 10]
deck = None

def shuffle_deck():
    global deck
    # card deck (we don't care the suite, but, for gui game in future) - 3 sets
    deck = \
        list(itertools.product(range(4), cardset)) \
        + list(itertools.product(range(4), cardset)) \
        + list(itertools.product(range(4), cardset))
        random.shuffle(deck)

shuffle_deck()
```

```
# the table of policy to access with the three indices
Q = np.random.uniform(size=(10, 10, 2, 2))
```

```
dealer = None # dealer's hands
player = None # player's hands
# reset the environment
    global dealer, player
    shuffle_deck()
    dealer = [ deck.pop(), deck.pop() ]
    player = [ deck.pop(), deck.pop() ]
    dealer_score = dealer[0][1]
    if player[0][1] == 1 and player[1][1] == 1:
        player_score = 1
        has_ace =
    elif player[0][1] == 1:
        player_score = 11 + player[1][1]
        has_ace =
    elif player[1][1] == 1:
        player_score = 11 + player[0][1]
        has ace =
```

```
player\_score = player[0][1] + player[1][1]
        while player_score
            player.append(deck.pop())
            player_score += player[-1][1]
        has_ace =
    # 1st step
    return { 'observation': (player_score, dealer_score, has_ace),
              reward': 0., 'step_type': STEPTYPE_FIRST }
import random
epsilon = 0.01
    idx = (observ[0] - 12, observ[1] - 1, observ[2])
    # epsilon-soft greedy policy
    if random.random() < epsilon:</pre>
        return 1 if Q[idx][0] < Q[idx][1] else 0</pre>
    global player, dealer
    player_score, dealer_open, has_ace = step['observation']
    # has ace is used to check if the player has
    game_stop = Fals
        player.append(deck.pop())
        player_score += player[-1][1]
        if player_score == 21:
            game_stop =
        elif player_score > 21:
            if has_ace == 1:
               player_score -= 10
                has_ace =
                game_stop = True
                busted = T
        game_stop = True
    if busted:
       return { 'observation': (player_score, dealer_open, has_ace),
                 'reward': -1., 'step_type': STEPTYPE_LAST }
    if game_stop:
       dealer_has_ace = Fals
        dealer_busted = False
        # examine dealer's hands
        if dealer[0][1] == 1 and dealer[1][1] == 1:
```

```
dealer score
            dealer_has_ace =
        elif dealer[0][1] == 1:
    dealer_score = 11. + dealer[1][1]
            dealer_has_ace =
        elif dealer[1][1] == 1:
            dealer_score = 11. + dealer[0][1]
            dealer_has_ace =
            dealer_score = dealer[0][1] + dealer[1][1]
            dealer_has_ace =
        while dealer_score < 17:</pre>
            dealer.append(deck.pop())
            dealer_score += dealer[-1][1]
            if dealer_score > 21:
                if dealer_has_ace:
                    dealer_score -=
                    dealer_has_ace = False
                    dealer_busted = True
        if dealer_busted:
            reward = 1
            if player_score > dealer_score:
                reward = 1
            elif player_score < dealer_score:</pre>
                reward = -1.
                reward = 0.
        return { 'observation': (player_score, dealer_score, has_ace),
                  'reward': reward, 'step_type': STEPTYPE_LAST }
        return { 'observation': (player_score, dealer_open, has_ace),
                 'reward': 0., 'step_type': STEPTYPE_MID }
def generate_episode(policy_func=get_eps_soft_action):
    episode = list()
    step = generate_start_step()
    episode.append(step)
    while step['step_type'] != STEPTYPE_LAST:
       action = policy_func(step)
       step = generate_next_step(step, action)
       episode.append(step)
        actions.append(action)
    return episode, actions
test = generate_episode()
test
([{'observation': (18, 8, 0), 'reward': 0.0, 'step_type': 0},
 {'observation': (27, 8, 0), 'reward': -1.0, 'step_type': 2}],
def in_episode(epi, observ, action):
    for s, a in zip(*epi):
```

```
example-mc-on-policy.ipynb
# monte-carlo exploring-start policy learning
maxiter =
gamma = 1
epsilon = 0.4
N = np.zeros((10, 10, 2, 2), dtype='float32')
SUM = np.zeros((10, 10, 2, 2), dtype='float32')
Q = np.random.uniform(size=(10, 10, 2, 2))
    episode = generate_episode()
    last_step = episode[0].pop()
    while len(episode[0]) > 0:
        G = gamma * G + last_step['reward']
        last_step = episode[0].pop()
        last_action = episode[1].pop()
        # exploring-start estimation: if the state appears for the first time
        observ = last_step['observation']
                                              1, observ[2], last_action)
         if not in_episode(episode, observ, last_action):
            SUM[idx] += G
            Q[idx] = SUM[idx] / N[idx]
import pandas as pd
wo_ace = pd.DataFrame(np.zeros((10, 10)),
```

```
wo_ace.loc[row + 12, col + 1] = v
```

```
wo_ace
```

```
2
             3
                  4
                           6
                                   8
                                        9
                                             10
21
20
19
18
17
16
15
14
13
12
     0
                                             0
```

```
# without ace
w_ace = pd.DataFrame(np.zeros((10, 10)),
                            columns = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10],
index = [21, 20, 19, 18, 17, 16, 15, 14, 13, 12], dtype='int32')
for row in range(10):
```

```
v = 1 if Q[row, col, 1, 0] < Q[row, col, 1, 1] else
w_ace.loc[row + 12, col + 1] = v</pre>
w_ace
                                                                       10
21
20
19
18
12
```