

Battleship agent

Training of an agent capable of playing battleship

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Classes: Ship

- A class to represent a ship in the Battleship game.
- **Attributes:**
 - *size* : *int* → The size of the ship
 - *hits* : *list* → A list to keep track of the coordinates where the ship has been hit
 - *x1, y1, x2, y2* : *int* → The coordinates of the ship's position on the grid
 - *orientation* : *str* → The orientation of the ship, either 'horizontal' or 'vertical'
- **Methods:**
 - `place(self, x1, y1, orientation)` → Saves the position and orientation of the ship on the grid
 - `hit(self, x, y, show=False)` → Registers a hit on the ship at the given coordinates, return *True* if the ship is completely hit (sunk), *False* otherwise



Classes: Battleship

- A class to represent the Battleship game environment.
- **Attributes:**
 - *ships : list* → A list of Ship objects representing the ships in the game
 - *opponent_grid : numpy.ndarray* → A 2D array representing the opponent's grid with ship positions (-1 means sea, 1, 2, 3, ... are the ships' indices+1)
 - *player_grid : numpy.ndarray* → A 2D array representing the player's grid with hits and misses (0 means unknown, -3 is a miss, -2 is a hit)
 - *sunken_ships : list* → A list to keep track of the indices of sunken ships
- **Important methods:**
 - *build_ships(self)* → Randomly places ships on the grid. Ships cannot overlap or touch each other
 - *action(self, x, y, last_action)* → Performs an action on the player's grid at the given coordinates and return the reward. It also return *True* if the action result is a hit, *False* otherwise

Other important functions

- `get_q_values(state)` → Retrieves the Q-values for a given state from the Q-table. If the state is not in the Q-table, initializes it with random values. The Q-table is implemented as a dictionary.
- `main` → For each episode, the process executes max 1000 steps. At each step, it can either select a random action with a probability *epsilon* or perform the action with the highest Q-value for the current state (i.e., the *player_grid*). The Q-table is then updated using the following equation:

$$\text{new_q} = (1 - \text{LEARNING_RATE}) * \text{current_q} + \text{LEARNING_RATE} * (\text{reward} + \text{DISCOUNT} * \text{max_future_q})$$

Finally, it saves the average episode reward.

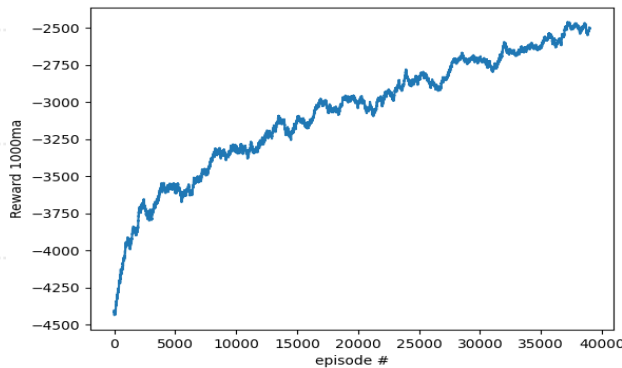
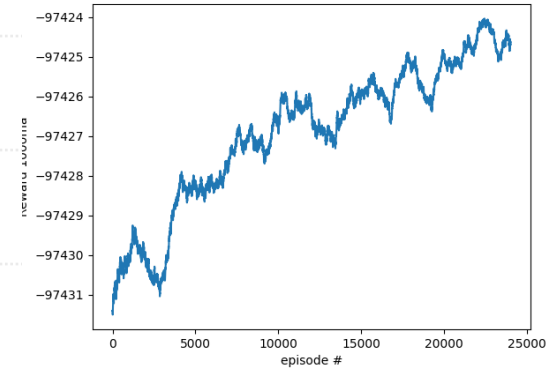


Training parameters

SIZE = 4	<i># grid dimension</i>
HM_EPISODES = 40000	<i># number of episodes</i>
TURN_PENALTY = 50	<i># penalty for each turn</i>
HIT_REWARD = 150	<i># reward for a hit</i>
CONSECUTIVEHIT_REWARD = 50	<i># reward for consecutive hits</i>
CONSECUTIVEMISS_PENALTY = 20	<i># penalty for misses after a hit</i>
SUNK_REWARD = 30	<i># reward for sinking a ship</i>
MISS_PENALTY = 25	<i># penalty for a miss</i>
ALREADY_HIT_PENALTY = 200	<i># penalty for hitting a cell already hit</i>
WIN_REWARD = 1030	<i># reward for winning</i>
ZEROCELLS_REWARD = 20	<i># reward for each remaining zero cell in the grid</i>
epsilon = 0.5	<i># exploration rate</i>
EPSILON_DECAY = 0.99999	<i># exploration rate decay</i>
SHOW_EVERY = 1000	<i># how often to show the game</i>
LEARNING_RATE = 0.1	<i># learning rate</i>
DISCOUNT = 0.9	<i># discount rate</i>

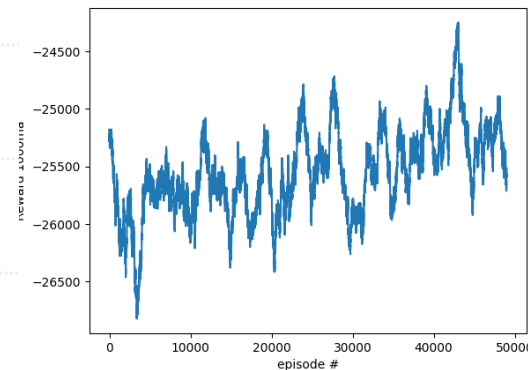
Plots

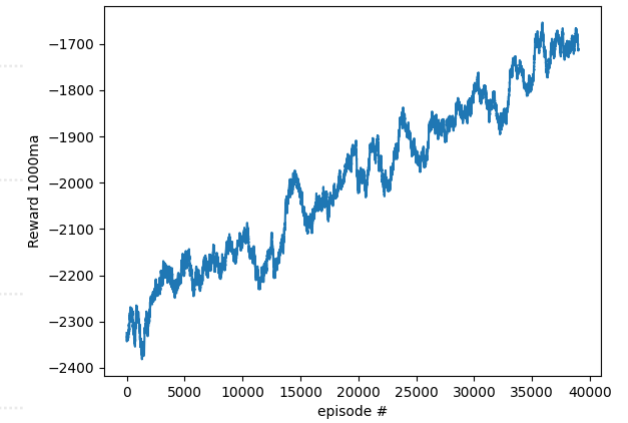
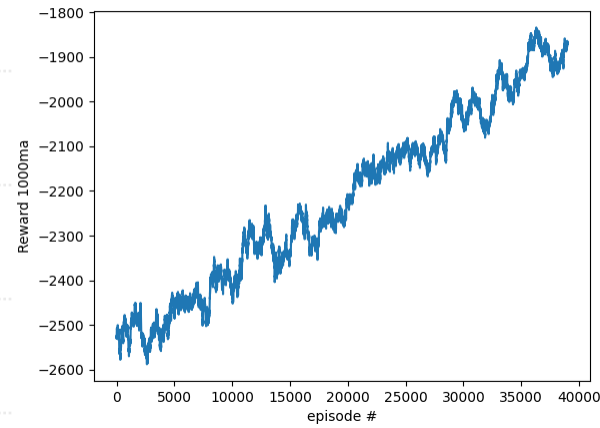
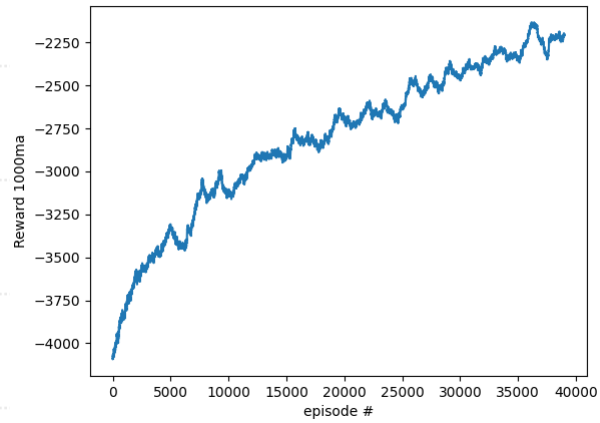
- 1000 rounds (mandatory)
- $\text{TURN_PENALTY} = \begin{cases} 0.5 & \text{if not win} \\ 0 & \text{otherwise} \end{cases}$



- More episodes (25k → 40k)
- Higher ALREADY_HIT_PENALTY (100 → 200)
- Introduced SUNK_REWARD
- Higher TURN_PENALTY (0.5 → 5)
- Higher WIN_REWARD (1000 → 1030)
- Higher MISS_PENALTY (5 → 25)
- Introduced ZEROCELLS_REWARD, CONSECUTIVEHIT_REWARD and CONSECUTIVEMISS_PENALTY

- 7x7 grids





- Three consecutive training on the same Q-table
- Higher CONSECUTIVEMISS_PENALTY (15 \rightarrow 20)
- Higher HIT_REWARD (100 \rightarrow 150)
- No diagonals contact between ships (less possible states)



BattleshipAgent class

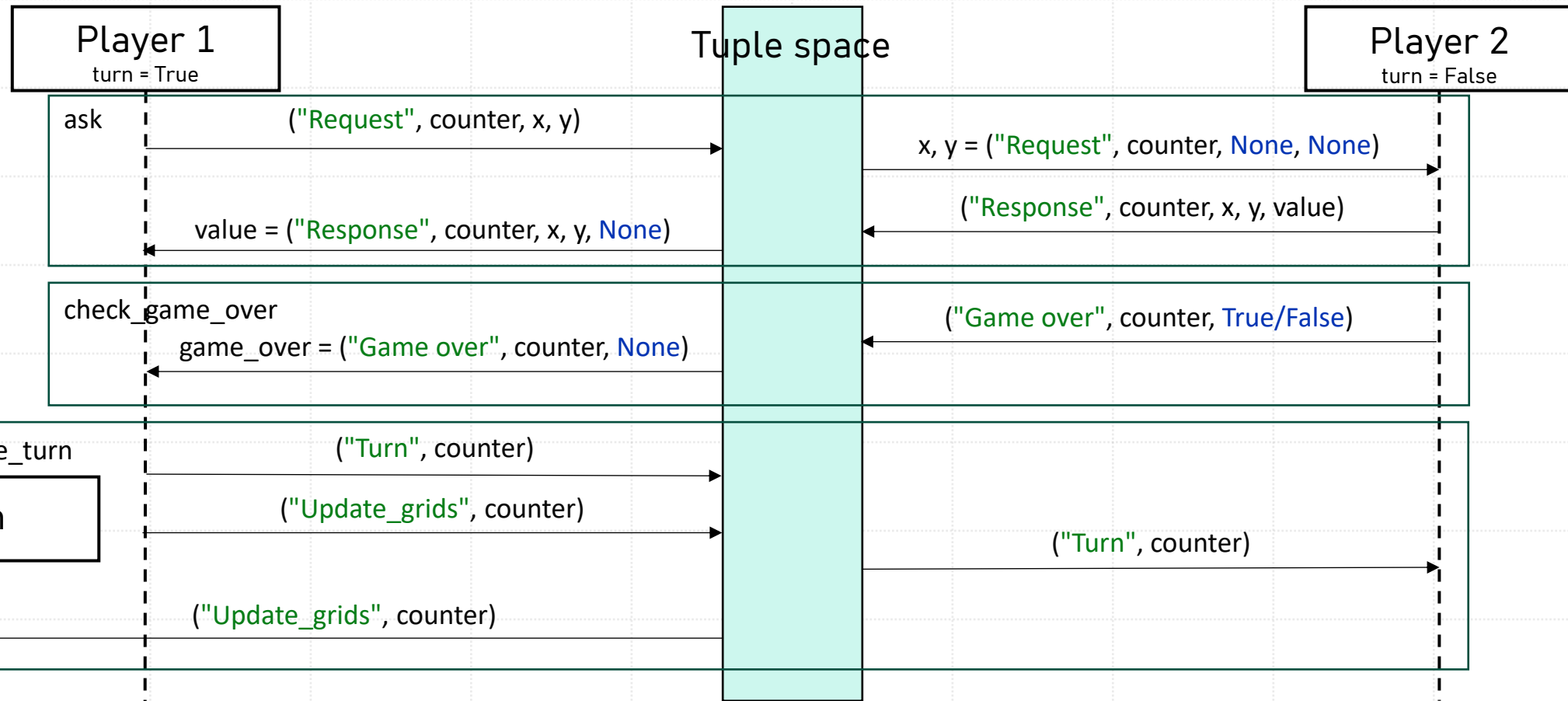
- A class to represent an agent in the Battleship game
- **Attributes:**
 - *id : str* → Name of the agent
 - *ts : BlockingTupleSpace* → Tuple space for the communication between the agents
 - *turn : bool* → A flag indicating if it is the agent's turn
 - *counter : int* → Turn counter
 - *q_table : dict* → The Q-table for storing the Q-values
 - *grid_size : int* → Size of the gaming grid
 - *ships : list* → A list of Ship objects representing the ships in the game
 - *sunken_ships : list* → A list to keep track of the indices of sunken ships
 - *done : bool* → A flag indicating if the game is over
 - *player_grid : numpy.ndarray* → A 2D array representing the player's grid with ship positions
 - *opponent_grid : numpy.ndarray* → A 2D array representing the opponent's grid with hits and misses
 - *human : bool* → A flag indicating if the player is human or not, if it is human, do not show its opponent grid and take the action from input

- **Important methods**

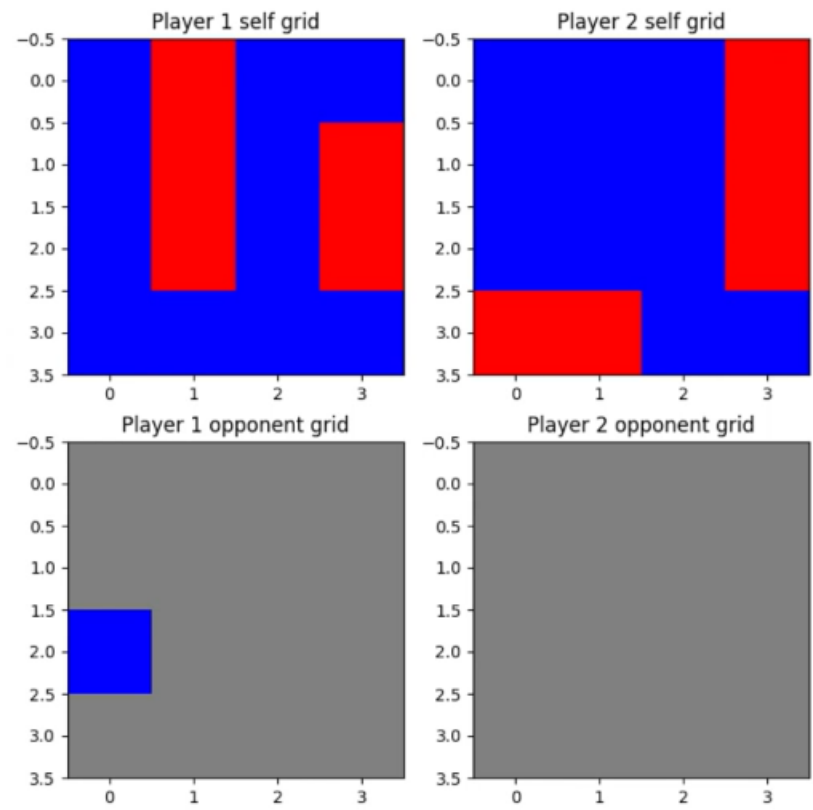
- `build_ships(self)` → Randomly places ships on the grid. Ships cannot overlap or touch each other
- `step(self)` → Performs a step in the game, either making a move or responding to a request (depending on whether it is its turn or not)
- `get_q_values(state)` → Retrieves the Q-values for a given state from the Q-table. If the state is not in the Q-table, initializes it with random values. The Q-table is implemented as a dictionary
- `choose_action(self)` → Chooses an action based on the current state and Q-values. It follows an epsilon-greedy policy
- `ask(self, x, y)` → Sends a request to the opponent and returns the response
- `change_turn(self)` → Changes the turn to the other player. It is used to synchronize the agents
- `loop(self, delay=2)` → Runs the game loop until the game is over

The game is played by 2 agents with the same policy (but it can be different). The agents run in two separate threads and they communicate with each other using a simple tuple space.

Communication



Turn 0





Open issues and future deployment

- Efficiency with bigger grids (and so bigger Q-tables)
- Better Q-table representation and implementation
- Improve the policy (number of turns, avoid hitting the same cell twice...)
- Different policies for different agents
- Better interface for playing human vs agent

Thank you for your attention