

Report on Final Project Quarto

(Laboratories Included)

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link to repository:

labs: https://github.com/milad818/computational_intelligence_2022

final project: <https://github.com/milad818/CI2022-23-Quarto-Final-Project>

Laboratories

Laboratory assignments are done in group with my dear friend Masoud Karimi.

Lab 1

Task

Given a number N and some lists of integers $P = (L_0, L_1, L_2, \dots, L_n)$, determine, if possible, $S = (L_{s_0}, L_{s_1}, L_{s_2}, \dots, L_{s_n})$ such that each number between 0 and $N - 1$ appears in at least one list

$$\forall n \in [0, N - 1] \exists i : n \in L_{s_i}$$

and that the total numbers of elements in all L_{s_i} is minimum.

Instructions

- Create the directory `lab1` inside the course repo (the one you registered with Andrea)
- Put a `README.md` and your solution (all the files, code and auxiliary data if needed)
- Use `problem` to generate the problems with different N
- In the `README.md`, report the the total numbers of elements in L_{s_i} for problem with $N \in [5, 10, 20, 100, 500, 1000]$ and the total number on `nodes` visited during the search. Use `seed=42`.
- Use `GitHub Issues` to peer review others' lab

Group Members: Milad Zakhireh, Masoud Karimi

`__init__.py`

Import

numpy

```
import problem
import state as S
import property as P
import search as SRCH
```

```
N = 5
seed = 42
goal = set(range(N))
initial = ([])
parentState = dict()
stateCost = dict()

props = P.Properties(N, seed, goal, initial)
```

```

problem = problem.makeProblem(props.N, props.seed)
# problem = sorted(problem, key=lambda l: len(l))

initialState = S.State(props.initial)
goalState = S.State(props.goal)
# initialState = S.State(numpy.array([1,2,34,3,5]))

def h(newState : S , currentState : S ):
    # return (numpy.sum((currentState != goalState) &
    (numpy.array(list(currentState.contain())))).size
    # return numpy.sum( ( currentState != goalState ) &
    (numpy.array(list(currentState.contain()))))
    # return len(currentState.contain())
    #
    _cS = currentState.copyData()
    _nS = newState.copyData()
    _cS = set(_cS)
    _nS = set(_nS)
    # _cS |= set(_nS)
    _cS &= set(_nS)
    return len (_cS)

path = SRCH.doesSearch(initialState, goalState, parentState, stateCost, problem,
                        unitCost= lambda s:1,
                        priorityFunction=lambda nS,cS : h(nS , cS) )

W = 0
for p in path:
    W += len(p)

print (f"path = {path}")
print(f"W = {W}")

```

priorityQueue.py

```

import
heapq

class PriorityQueue:
    """A basic Priority Queue with simple performance optimizations"""

```

```

def __init__(self):
    self._data_heap = list()
    self._data_set = set()

def __bool__(self):
    return bool(self._data_set)

def __contains__(self, item):
    return item in self._data_set

def push(self, item, p=None):
    assert item not in self, f"Duplicated element"
    if p is None:
        p = len(self._data_set)
    self._data_set.add(item)
    heapq.heappush(self._data_heap, (p, item))

def pop(self):
    p, item = heapq.heappop(self._data_heap)
    self._data_set.remove(item)
    return item

```

problem.py

```

import
random

import numpy

def makeProblem(N, seed=None):
    random.seed(seed)
    return [ numpy.array(list(set(random.randint(0, N - 1) for n in
range(random.randint(N // 5, N // 2)))) for n in range(random.randint(N, N * 5))
]

```

property.py

```

class
Properties:
    def __init__(self, numberOfElements, seed, finalGoal, initialList):
        self.N = numberOfElements
        self.seed = seed
        self.goal = list(finalGoal)
        self.initial = list(initialList)

```

search.py

```

import
logging

import priorityQueue as PQ
import state
import problem as PROB
import numpy

def goalStateCheck(stateToBeChecked : state, goal : state):
    return set(stateToBeChecked.contain()) == set(goal.contain())

def resultOfAction(currentState : state, action: numpy.ndarray):
    # temp = currentState.contain()
    # temp.append(action)
    # temp = tuple(temp)
    # return temp

    #
    cS = set(currentState.copyData())
    cS |= set(action)
    return state.State(cS)

def doesSearch ( initialStat: state,
                 goal : state,
                 parentState : dict,
                 stateCost: dict,
                 problem : PROB,
                 unitCost : callable,

```

```

        priorityFunction : callable
    ):
# print(len(problem))
# print(problem)
frontier = PQ.PriorityQueue()
parentState.clear()
stateCost.clear()

currentState = initialStat
parentState[currentState] = None
stateCost[currentState] = 0

print("spanning tree")
while currentState is not None and not goalStateCheck(currentState, goal):
    for p in problem:
        newState = resultOfAction(currentState, p)
        cost = unitCost(p)

        # if type(res) != int:
        #     newState = res
        #     cost = unitCost(p)
        # else:
        #     continue
        # newState = resultOfAction(currentState, p)
        # cost = unitCost(p)
        if newState not in stateCost and newState not in frontier:
            parentState[newState] = currentState
            stateCost[newState] = stateCost[currentState] + cost
            frontier.push(newState, p = priorityFunction(newState,
currentState))
            # logging.warning(f"Added new node to frontier (cost={
stateCost[newState]})")
            elif newState in frontier and stateCost[newState] >
stateCost[currentState] + cost:
                old_cost = stateCost[currentState]
                parentState[newState] = currentState
                stateCost[newState] = stateCost[currentState] + cost
                # logging.warning(f"Updated node cost in frontier: {old_cost} ->
{stateCost[newState]})")

        if frontier:
            currentState = frontier.pop()

```

```

        else:
            currentState = None

    path = list()
    s = currentState

    while s:
        path.append(s.copyData())
        s = parentState[s]
        # path.append(s.copyData())

    logging.warning(f"Found a solution in {len(path):,} steps; visited
{len(stateCost):,} states")
    return list(reversed(path))

```

state.py

```

import
numpy

class State:
    def __init__(self, data: numpy.ndarray):
        self._data = data

    def __hash__(self):
        return hash(bytes(self._data))
    #
    def __eq__(self, other):
        return bytes(self._data) == bytes(other._data)

    def __lt__(self, other):
        return bytes(self._data) < bytes(other._data)
    #
    def __str__(self):
        return str(self._data)

    def __repr__(self):
        return repr(self._data)

    def __len__(self):

```

```

        return len(self._data)

    def contain(self):
        return self._data

    def copyData(self):
        return self._data.copy()

    def createSet (self, x, y):
        return set

```

Results achieved through the solution discovered are as below:

for N = 5:

Found a solution in 3 steps; visited 32 states

path = [{0}, {0, 1, 3}, {0, 1, 2, 3, 4}]

W = 9

for N = 10:

Found a solution in 3 steps; visited 776 states

path = [{2, 5}, {1, 2, 5, 6, 8}, {0, 1, 2, 3, 4, 5, 6, 7, 8, 9}]

W = 17

for N = 20:

Found a solution in 4 steps; visited 15,887 states

path = [{0, 1, 2, 7}, {0, 1, 2, 6, 7, 9, 16, 19}, {0, 1, 2, 3, 5, 6, 7, 8, 9, 10, 13, 14, 16, 17, 19}, {0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19}]

W = 47

Review of Marco Masera on My Work for Lab1



Marco-Masera commented on Oct 19, 2022



Hi. I reviewed your code.

If I got it right you were trying to write an A* search algorithm. In this I found a few issues:

-The priority function computes the intersection between the elements covered by the old state and the ones in the new state. But since creating a state means adding new elements to the existing set, this intersection always returns the set of the previous state.

This cannot function as an heuristic function for A* as it does not return a lower bound for the best solution from the given Node. A possible and simple function could be computing the number of elements that still needs to be covered to get to a solution.

-UnitCost and stateCost: the unit_cost does not represent the cost of a Node, since the cost is given by the total number of elements in the lists chosen, not the number of lists itself.

Moreover, it seems like the stateCost dictionary is used in order to update the parentState dictionary. But this should not be necessary: if the A* algorithm works properly, given a certain state, you can be sure it will always be visited with the lowest cost path from the starting one, so there is no need to keep track of that.

-In the search function, when you find a node that you already visited but with a lower cost, you update the stateCost. Keep in mind that this does not update the priorityQueue.

My suggestion is to:

-Modify the States so you keep track not only of the elements covered by also the cost of this node, given by the total number of elements. Alternatively you could keep track of the lists used - which will cause more memory consumption but could allow for better performance improvements later.

-Modify the priority function so that it returns a lower bound for the best solution from the given node

-Order the priority queue with parameter priorityFunction() + real_cost_to_node (kept inside the state object as stated in point 1).

-Any node will be visited only once, and the first solution found will be (on of) the best solution(s).

Other possible performance improvements include:

-The search space you are generating right now include different permutations of the same lists set. Since order is not a matter here, you could tweak it so that for a given set of lists it only generates one node that contains this set - regardless of order.

-The State objects keeps the information about covered elements as a numpy array, but each time you do some operations you convert it to a set. You could decide to keep it as a set in the first place. Also sometimes you call copyData(), then you create a set from the copied data, and then you compute the intersection. This clones the data 3 times. You could simply call intersection() from the original set and Python will return a new set containing the result without changing the original.

My Peer Review of Marco D'Almo's Work for Lab1

It has to be mentioned that the code has recently been modified. It was not complete when I reviewed it.



milad818 commented on Oct 23, 2022



Review by Milad

Major Issues

- According to what I could understand from the code and additional comments, the code is incomplete and therefore, no conclusion is drawn of the achieved results.
- It seems like you are on the right track trying to move along the graph through the nearest vertex with the lowest cost without visiting the already visited vertices; however, graph creation and starting the problem could never take place due to some issues in implementation e.g. trying to loop through an integer or other data type that loops cannot work on.

Problem generator

Problem generator as given by professor, with default N values.

```
N_VALUES = [5, 10, 20, 100, 500, 1000]
SEED = 42
```

In [3]:

```

import logging
import random
from copy import copy
import random
import platform
from collections import Counter
from gx_utils import *

def problem(N, seed=None):
    '''Function to generate a set covering problem with number N, default
seed is None'''
    random.seed(seed)
    return [
        list(set(random.randint(0, N - 1) for n in range(random.randint(N //
5, N // 2))))
        for n in range(random.randint(N, N * 5))
    ]

```

Greedy solution

Greedy solution as given by professor.

In [4]:

```

import logging

def greedy(N):
    '''Just adds lists sorted from shortest to longest until the set is
covered'''
    goal = set(range(N))
    covered = set()
    solution = list()
    nodes = 0
    all_lists = sorted(problem(N, seed=SEED), key=lambda l: len(l))
    while goal != covered:
        x = all_lists.pop(0)
        nodes+=1
        if not set(x) < covered:
            solution.append(x)
            covered |= set(x)

    logging.info(
        f"Greedy solution for N={N}: w={sum(len(_) for _ in solution)}
(bloat={(sum(len(_) for _ in solution)-N)/N*100:.0f}%, Nodes visited:
{nodes})"
    )
    logging.debug(f"{solution}")

```

In [5]:

```

logging.getLogger().setLevel(logging.INFO)
for N in N_VALUES:
    greedy(N)

```

```

INFO:root:Greedy solution for N=5: w=5 (bloat=0%, Nodes visited: 13)
INFO:root:Greedy solution for N=10: w=13 (bloat=30%, Nodes visited: 14)
INFO:root:Greedy solution for N=20: w=46 (bloat=130%, Nodes visited: 14)
INFO:root:Greedy solution for N=100: w=332 (bloat=232%, Nodes visited: 23)
INFO:root:Greedy solution for N=500: w=2162 (bloat=332%, Nodes visited: 28)
INFO:root:Greedy solution for N=1000: w=4652 (bloat=365%, Nodes visited: 27)

```

"longest first" Greedy

Greedy solution as seen in the professor version, just starting from the longest random list first. Performs generally worse, fairly better in just one case, $N = 20$. It is probably a random consequence of the list generation.

In [6]:

```

def longest_greedy(N):
    '''As the greedy, but the first list to be added is the longest'''
    goal = set(range(N))
    covered = set()
    solution = list()
    nodes=0
    all_lists = sorted(problem(N, seed=SEED), key=lambda l: len(l))
    first_iter = True
    while goal != covered:
        if first_iter:
            x = all_lists.pop()
            nodes+=1
            first_iter = False
        else:
            x = all_lists.pop(0)
            nodes+=1
        if not set(x) < covered:
            solution.append(x)
            covered |= set(x)

    logging.info(
        f"Alternate greedy version for N={N}: w={sum(len(_) for _ in
solution)} (bloat={(sum(len(_) for _ in solution)-N)/N*100:.0f}%, Nodes
visited: {nodes})")

logging.getLogger().setLevel(logging.INFO)
for N in N_VALUES:
    longest_greedy(N)

INFO:root:Alternate greedy version for N=5: w=5 (bloat=0%, Nodes visited: 5)
INFO:root:Alternate greedy version for N=10: w=14 (bloat=40%, Nodes visited:
15)
INFO:root:Alternate greedy version for N=20: w=36 (bloat=80%, Nodes visited:
15)
INFO:root:Alternate greedy version for N=100: w=340 (bloat=240%, Nodes visite
d: 19)
INFO:root:Alternate greedy version for N=500: w=2187 (bloat=337%, Nodes visit
ed: 28)

```

INFO:root:Alternate greedy version for N=1000: w=4699 (bloat=370%, Nodes visited: 28)

Different solution

Another possible solution with the implementation of Dijkstra search. Note - this is incomplete.

In [7]:

```
from gx_utils import *

class Graph:
    '''graph initialization, compute'''
    def __init__(self, num_vert):
        self.v = num_vert
        self.edges = [[-1 for i in range(num_vert)] for j in range(num_vert)]
        self.visited = []

    def add_edge(self, u, v, weight):
        self.edges[u][v] = weight
        #self.edges[v][u] = weight

def create_graph(graph, problem_list):
    for i in range(graph.v):
        for j in range(graph.v):
            graph.add_edge(i, j, len(problem_list[j]))
            graph.add_edge(j, i, len(problem_list[i]))
```

In [8]:

```
def dijkstra_sol(N):

    GOAL = set(range(N))

    def problem_start(N):
        'starts the problem, adds the first vertex, so the shortest list, to
the visited'
        p = sorted(problem(N, seed=SEED), key=lambda l: len(l))
        g = Graph(len(p))
        create_graph(g, p)

        return g, p

    def visited_to_set(state, problem_list):
        sol = [tuple(problem_list[e]) for e in state]
        return set(sum((e for e in sol), start=()))

    def goal_test(state, problem_list):
        return visited_to_set(state, problem_list) == GOAL

    _, p = problem_start(N)

    D = {v:float('inf') for v in range(_v)}
    solution = p
```

```

for start in range(_v):
    g, p = problem_start(N)
    pq = PriorityQueue()
    pq.push((0, start), p=0)
    #pq.put((0, start))
    D = {v:float('inf') for v in range(_v)}
    D[start]=0

    while pq:
        (dist, current_vertex) = pq.pop()
        g.visited.append(current_vertex)

        for neighbor in range(g.v):

            if g.edges[current_vertex][neighbor] != -1:
                new_elements = visited_to_set(g.visited, p) -
set(p[neighbor])
                distance = g.edges[current_vertex][neighbor] -
len(new_elements)

                if neighbor not in g.visited:
                    old_cost = D[neighbor]
                    cost = D[current_vertex] + distance

                    if cost < old_cost:
                        pq.push((cost, neighbor), p=cost)
                        D[neighbor] = cost

                if goal_test(g.visited, p):

                    new_solution = [p[elem] for elem in g.visited]
                    weight_new = sum(len(_) for _ in new_solution)
                    weight_old = sum(len(_) for _ in solution)
                    solution = new_solution if weight_new<weight_old else
solution

                    if weight_new==N:
                        return solution

                    pq = PriorityQueue()

    return solution

# incomplete and possibly non functioning approach, left it here to leave
trace of what was delivered
# will probably complete/review it for learning purposes

logging.getLogger().setLevel(logging.DEBUG)

for N in [5, 10, 20]:
    solution = dijkstra_sol(N)

```

In [15]:

```

logging.info(
    f" Solution for N={N:,}: "
    + f"w={sum(len(_) for _ in solution):,} "
    + f"(bloat={(sum(len(_) for _ in solution)-N)/N*100:.0f}%), solution
is {solution}"
)

```

```

INFO:root: Solution for N=5: w=5 (bloat=0%), solution is [[0], [1], [4], [2],
[3]]

```

```

INFO:root: Solution for N=10: w=10 (bloat=0%), solution is [[8, 9, 3], [6], [
0, 4], [2, 5], [1, 7]]

```

```

INFO:root: Solution for N=20: w=33 (bloat=65%), solution is [[6, 9, 11, 12, 1
7], [8, 4, 7], [18, 2, 15], [8, 16, 5], [1, 3, 13, 14], [17, 18, 7], [17, 10,
1, 7], [16, 9, 19, 6], [0, 1, 2, 7]]

```

NOTE: everything here is copied from professor Squillero's solution and is here for the sake of completion of the exercise and to keep track of the different algorithms. It should not be considered as mine.

Dijkstra's

Professor Squillero's solution, with added comments for learning purposes.

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<https://github.com/squillero/computational-intelligence>

Free for personal or classroom use; see 'LICENSE.md' for details.

In [10]:

```

from gx_utils import *

def dijkstra(N, all_lists):
    """Vanilla Dijkstra's algorithm"""

    # Goal and utility variables initialization
    # all_lists is the problem generated sorted list
    GOAL = set(range(N))

    # generates a tuple consisting of the lists of the problem
    # ordered and unchangeable list
    all_lists = tuple(set(tuple(_) for _ in all_lists))
    frontier = PriorityQueue()
    nodes = 0

    # converts a state (list of visited nodes) into a set
    # to be able to compare with goal
    def state_to_set(state):
        return set(sum((e for e in state), start=()))

    # comparison with goal, returns bool
    def goal_test(state):

```

```

    return state_to_set(state) == GOAL

# returns a list of lists, which
# could be the possible step -> all except
# subsets or equivalent sets
def possible_steps(state):
    current = state_to_set(state)
    return [l for l in all_lists if not set(l) <= current]

# computation of the internal weight
# per state
def w(state):
    cnt = Counter()
    cnt.update(sum((e for e in state), start=()))
    return sum(cnt[c] - 1 for c in cnt if cnt[c] > 1), -sum(cnt[c] == 1
for c in cnt)

# keeps on updating the frontier based on
# internal weight computed by w() function
# and then choosing the next state as the one with
# the lowest internal weight
state = tuple()
while state is not None and not goal_test(state):
    nodes += 1
    for s in possible_steps(state):
        frontier.push((*state, s), p=w((*state, s)))
    state = frontier.pop()

logging.debug(f"dijkstra: SOLVED! nodes={nodes:,}; w={sum(len(_) for _ in
state):,}; iw={w(state)}")
return state

```

In [14]:

```

logging.getLogger().setLevel(logging.INFO)

for N in [5, 10, 20]:
    solution = dijkstra(N, problem(N, seed=42))
    logging.info(
        f" Solution for N={N:,}: "
        + f"w={sum(len(_) for _ in solution):,} "
        + f"(bloat={(sum(len(_) for _ in solution)-N)/N*100:.0f}%) "
    )

INFO:root: Solution for N=5: w=5 (bloat=0%)
INFO:root: Solution for N=10: w=10 (bloat=0%)
INFO:root: Solution for N=20: w=23 (bloat=15%)

```

Hill Climbing

Professor Squillero Hill Climbing solution, with added comments for learning purposes.

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<https://github.com/squillero/computational-intelligence>

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In [20]:

```
def hc(N, all_lists):
    """Vanilla Hill Climber"""

    # same operation as for Dijkstra
    all_lists = set(tuple(_) for _ in all_lists)

    # state evaluation by number of lists in the counter
    # and total sum of the elements in them
    def evaluate(state):
        cnt = Counter()
        cnt.update(sum((e for e in state), start=()))
        return len(cnt), -cnt.total()

    # tweak function on the current solution
    # randomly eliminates a list from the solution
    # or adds a list non present in the solution
    # from the possible lists
    def tweak(solution):
        new_solution = set(solution)
        while new_solution and random.random() < 0.7:
            r = random.choice(list(new_solution))
            new_solution.remove(r)
        while random.random() < 0.7:
            a = random.choice(list(all_lists - solution))
            new_solution.add(a)
        return new_solution

    # sets a max number of consecutive useless steps
    # to stop tweaking
    current_solution = set()
    useless_steps = 0
    while useless_steps < 10_000:
        useless_steps += 1
        candidate_solution = tweak(current_solution)
        if evaluate(candidate_solution) > evaluate(current_solution):
            useless_steps = 0
            current_solution = copy(candidate_solution)
            logging.debug(f"New solution: {evaluate(current_solution)}")
    return current_solution
```

In [30]:

```
logging.getLogger().setLevel(logging.INFO)

for N in [5, 10, 20, 100, 1000]:
    solution = hc(N, problem(N, seed=42))
    logging.info(
        f" Solution for N={N:,}: "
        + f"w={sum(len(_) for _ in solution):,} "
```



```

    + f"(bloat={(sum(len(_) for _ in solution)-N)/N*100:.0f}%)"
)
INFO:root: Solution for N=5: w=5 (bloat=0%)
INFO:root: Solution for N=10: w=11 (bloat=10%)
INFO:root: Solution for N=20: w=24 (bloat=20%)
INFO:root: Solution for N=100: w=214 (bloat=114%)
INFO:root: Solution for N=1,000: w=3,383 (bloat=238%)

def hc_adapt(N, all_lists):
    """Vanilla Hill Climber"""

    # same operation as for Dijkstra
    all_lists = set(tuple(_) for _ in all_lists)

    # state evaluation by number of lists in the counter
    # and total sum of the elements in them
    def evaluate(state):
        cnt = Counter()
        cnt.update(sum((e for e in state), start=()))
        return len(cnt), -cnt.total()

    # tweak function on the current solution
    # randomly eliminates a list from the solution
    # or adds a list non present in the solution
    # from the possible lists
    def tweak0(solution):
        new_solution = set(solution)
        while new_solution and random.random() < 0.7:
            r = random.choice(list(new_solution))
            new_solution.remove(r)
        while random.random() < 0.7:
            a = random.choice(list(all_lists - solution))
            new_solution.add(a)
        return new_solution

    def tweak1(solution):
        new_solution = set(solution)
        while new_solution and random.random() < 0.8:
            r = random.choice(sorted(list(new_solution), key=lambda x:
len(x))[0:5])
            new_solution.remove(r)
        while random.random() < 0.8:
            a = random.choice(list(all_lists - solution))
            new_solution.add(a)
        return new_solution

    def tweak2(solution):
        new_solution = set(solution)
        while new_solution and random.random() < 0.8:
            r = sorted(list(new_solution), key=lambda x: len(x))[0]
            new_solution.remove(r)
            a = random.choice(list(all_lists - solution))

```

In [16]:

```

        new_solution.add(a)
    return new_solution

# sets a max number of consecutive useless steps
# to stop tweaking
current_solution = set()
useless_steps = 0
while useless_steps < 10_000:
    useless_steps += 1

    if useless_steps < 5000:
        candidate_solution = tweak0(current_solution)
    elif 5000 < useless_steps < 8000:
        candidate_solution = tweak1(current_solution)
    else:
        candidate_solution = tweak2(current_solution)

    if evaluate(candidate_solution) > evaluate(current_solution):
        useless_steps = 0
        current_solution = copy(candidate_solution)
        logging.debug(f"New solution: {evaluate(current_solution)}")
return current_solution

```

In [29]:

```

logging.getLogger().setLevel(logging.INFO)

for N in [5, 10, 20, 100, 500, 1000]:
    solution = hc_adapt(N, problem(N, seed=42))
    logging.info(
        f" Solution for N={N:,}: "
        + f"w={sum(len(_) for _ in solution):,} "
        + f"(bloat={(sum(len(_) for _ in solution)-N)/N*100:.0f}%"
    )

INFO:root: Solution for N=5: w=5 (bloat=0%)
INFO:root: Solution for N=10: w=10 (bloat=0%)
INFO:root: Solution for N=20: w=24 (bloat=20%)
INFO:root: Solution for N=100: w=214 (bloat=114%)
INFO:root: Solution for N=500: w=1,509 (bloat=202%)
INFO:root: Solution for N=1,000: w=3,469 (bloat=247%)

```

Lab 2

Task

Given a number N and some lists of integers $P = (L_0, L_1, L_2, \dots, L_n)$, determine, if possible, $S = (L_{s_0}, L_{s_1}, L_{s_2}, \dots, L_{s_n})$ such that each number between 0 and $N - 1$ appears in at least one list

$$\forall n \in [0, N - 1] \exists i : n \in L_{s_i}$$

and that the total numbers of elements in all L_{s_i} is minimum.

Group Members: Milad Zakhireh (300708), Masoud Karimi (300283)

main.py

```
import
problem

import myEA
seed = 42
Ns = [5, 10, 20, 100, 500, 1000]
populationSizes = [10, 10, 10, 20, 30, 200]
problemSizes = [4, 4, 4, 7, 10, 13]
offspringSizes = [5, 5, 5, 10, 15, 50]
generatorsSize = 500
for index in range(len(Ns)):
    myProblem = problem.Problem(Ns[index], seed)
    ea = myEA.EA(problemSizes[index], populationSizes[index],
offspringSizes[index], generatorsSize, myProblem.setOfProblem)
    print(ea.resultFitness)
    print(ea.coveredLists)
```

myEA.py

```
import
copy

from collections import namedtuple, Counter
import random
from functools import cmp_to_key
import numpy

Individual = namedtuple("Individual", ["genome", "fitness"])
class EA:
```

```

def __init__(self, problemSize, populationSize, offspringSize, generatorsSize,
setOfProblem):
    self.problemSize = problemSize
    self.populationSize = populationSize
    self.offspringSize = offspringSize
    self.generatorsSize = generatorsSize
    self.listOfIndevidualsAndTheirFitness = list()
    self.rankedListOfIndevidualsAndTheirFitness = list()
    self.roulletWheelOfIndevidualsAndTheirFitness = list()
    self.setOfProblem = setOfProblem
    self.creatingPopulation(self.setOfProblem)
    self.breeding()
    self.resultIndex = self.rankedListOfIndevidualsAndTheirFitness[0].genome
    self.resultFitness = self.rankedListOfIndevidualsAndTheirFitness[0].fitness
    self.coveredLists = numpy.array(self.setOfProblem,
dtype=set)[list(self.resultIndex)]

```

```

def creatingPopulation(self, setOfProblem ):
    # Individual = namedtuple("Individual", ["genome", "fitness"])
    for genome in [tuple([ random.choice(range(len(setOfProblem))) for _ in
range(self.problemSize)]) for __ in range(self.populationSize) ]:
        self.listOfIndevidualsAndTheirFitness.append(Individual(genome,
self.forEvaluation(genome, setOfProblem)))
        self.rankOfEachIndevidualByMeansOfRouletteWheelApproach()
        self.rankedAsRoulletWheel()

```

```

def compare(self,pair1, pair2):
    _, fitness1 = pair1
    _, fitness2 = pair2
    digits1, total1 = fitness1
    digits2, total2 = fitness2

    if digits1 == digits2:
        if total1 < total2:
            return -1
        else:
            return 1
    if digits1 < digits2:
        return -1

```

```

        else:
            return 1

    def rankOfEachIndeividualByMeansOfRouletteWheelApproach(self):
        self.rankedListOfIndeividualsAndTheirFitness =
sorted(self.listOfIndeividualsAndTheirFitness, key=cmp_to_key(self.compare),reverse
= True)

    def rankedAsRouletteWheel(self):
        length = len(self.rankedListOfIndeividualsAndTheirFitness)
        for eachIndeividual in
range(len(self.rankedListOfIndeividualsAndTheirFitness)):
            genome, _ = self.rankedListOfIndeividualsAndTheirFitness[eachIndeividual]
            length-=1

self.rouletteWheelOfIndeividualsAndTheirFitness.append(Individual(genome,length))


    def forEvaluation(self, genome, setOfProblem=None):
        localList = list()
        if setOfProblem is not None:
            for index in genome:
                localList.append(setOfProblem[index])
        else:
            for index in genome:
                localList.append(self.setOfProblem[index])
        cnt = Counter()
        cnt.update(sum((e for e in localList), start=()))
        return len(cnt), -cnt.total()

    def breeding(self):
        for g in range(self.generatorsSize):
            offspring = list()
            for i in range(self.offspringSize):
                if random.random() < 0.3:
                    p = self.tournament()
                    o = self.mutation(p.genome)
                    # print(o)
                else:
                    p1 = self.tournament()
                    p2 = self.tournament()
                    o = self.cross_over(p1.genome, p2.genome)

```

```

        # print(o)
        f = self.forEvaluation(o)
        offspring.append(Individual(o, f))
    self.listOfIndividualsAndTheirFitness += offspring
    self.rankOfEachIndividualByMeansOfRouletteWheelApproach()
    self.rankedAsRouletteWheel()

    self.rankedListOfIndividualsAndTheirFitness =
self.rankedListOfIndividualsAndTheirFitness[:self.populationSize]
    self.rouletteWheelOfIndividualsAndTheirFitness =
self.rouletteWheelOfIndividualsAndTheirFitness[:self.populationSize]

    def tournament(self, tournament_size=2):
        return max(random.choices(self.rouletteWheelOfIndividualsAndTheirFitness,
k=tournament_size), key=lambda i: i.fitness)

    def cross_over(self, g1, g2):
        cut = random.randint(0, self.problemsize)
        return g1[:cut] + g2[cut:]

    def mutation(self, g):
        g= list(g)
        point = random.randint(0, self.problemsize - 1)
        indexForMutation = random.randint(0,len(self.setOfProblem)-1)
        return tuple(g[:point] + [indexForMutation] + g[point + 1:])

```

problem.py

```

import
random

```

```

class Problem:

    def __init__(self, n, seed):
        self.problemSize = n
        self.listOfProblem = self.myProblem(n, seed)
        self.setOfProblem = self.createSetFromListOfProblem()

    def myProblem(self, N, seed=None):
        """Creates an instance of the problem"""

```

```

        random.seed(seed)
        return [
            list(set(random.randint(0, N - 1) for n in range(random.randint(N //
5, N // 2))))
            for n in range(random.randint(N, N * 5))
        ]

    def creatSetFromListOfProblem(self):
        sortedList = set()
        for _ in self.listOfProblem:
            sortedelement = sorted(_)
            sortedList.add(tuple(sortedelement))
        return sorted(sortedList)

```

Results

N	weight	nodes
5	5	[(3,) (1,) (4,) (0, 2)]
10	10	[(2, 5) (1, 7) (0, 4) (3, 6, 8, 9)]
20	29	[(2, 6, 8, 10, 12, 15, 18) (0, 1, 3, 7, 9, 10, 11, 15)(1, 3, 8, 11, 14, 19) (4, 5, 8, 13, 15, 16, 17, 19)]
100	244	many nodes
500	1739	many nodes
1000	3980	many nodes

Review of Marco Masera on My Work for Lab 2



Marco-Masera commented on Nov 13, 2022 • edited ▾



Hi. I leave the review for Lab2

General considerations

It's a good implementation of GA for this problem. The code is very clean and readable, I like this stile of writing.

Major iusses

I did not find any major iusses in the code or in the solution.

Suggestions

These are not iusses but suggestions. You can try to play around a little bit more with different phases during the execution of the algorithm.

I did too write my algorithm with no phases - meaning it follows the same rules for every generation - and I found that after some iterations the individuals become too much similar one with another, and the solution stops improving. After the presentation of last monday I realised how much the result can improve with a finer control of the balance between exploration and exploitation.

Conclusion

I don't have much to say except for: good job :)

My Peer Review of Shayan Taghinezhad Roudbaraki's Work for Lab2



milad818 commented on Nov 12, 2022



Review by Milad

In general, the implementation is good and there seem to be some innovative ideas behind it. However, it is worth mentioning some issues which should have been given more attention.

Major Issues

- I could understand from your code that you have only employed the crossover operator and no mutation has been carried out which could possibly lead to more efficient results.
- It was kind of challenging to understand how the variable N has been used and I failed to discover the logic behind the corresponding lines.
- Results are not included in the readme file

Minor Issues

- Despite the nice implementation of the idea in mind, the code is not so much readable. Using a reasonable number of comments could be helpful.

```
import random, logging
seed = 42
N = 10

logging.getLogger().setLevel(logging.DEBUG)

def problem(N, seed=None):
    """Creates an instance of the problem"""

    random.seed(seed)
    return [
        list(set(random.randint(0, N - 1) for n in range(random.randint(N //
5, N // 2))))
        for n in range(random.randint(N, N * 5))
    ]

all_lists = problem(10, 42)

logging.debug(f"all_lists is {all_lists}")
DEBUG:root:all_lists is [[0, 4], [1, 2, 3], [9, 6], [0, 1], [8, 9, 3], [8, 3],
, [0, 3, 4, 7, 9], [4, 5, 6], [1, 3, 5], [1, 6], [0, 9, 4, 5], [8, 1, 6], [9,
3, 5], [0, 3], [1, 3, 6], [2, 5, 7], [1, 3, 4, 9], [8, 2, 3], [3, 4, 5, 6, 8]
, [0, 3], [1, 3, 4, 6], [3, 6, 7], [2, 3, 4], [9, 6], [8, 2, 3, 7], [0, 1], [
9, 2, 6], [6], [8, 0, 4, 1], [1, 4, 5, 6], [0, 4, 7], [8, 1, 4], [2, 5], [9,
5], [0, 1, 3, 4, 5], [9, 3], [1, 7], [8, 2], [8, 2, 7], [8, 9, 3, 6], [4, 5,
```



```
6], [8, 1, 3, 7], [0, 5], [0, 9, 3], [0, 3], [0, 5], [8, 3], [8, 2, 3, 7], [1, 3, 6, 7], [5, 6]]
```

In [51]:

```
def fitness(genome, n=N):
    correct_n = len(set(genome))
    repeated_n = len(genome) - correct_n
    return correct_n - repeated_n

def cross_over(g1, g2):
    return g1 + g2

def tournament(population, tournament_size=2):
    return max(random.choices(population, k=tournament_size), key=lambda i:
i.fitness)
```

In [52]:

```
from collections import namedtuple
from collections import Counter

Individual = namedtuple("Individual", ["genome", "fitness"])
population = list()

for i in range(10):
    genome = random.choice(all_lists)
    indiv = Individual(genome, fitness(genome, 10))
    population.append(indiv)

logging.debug(f"poulation is {population}")

DEBUG:root:poulation is [Individual(genome=[6], fitness=1), Individual(genome=[9, 2, 6], fitness=3), Individual(genome=[1, 4, 5, 6], fitness=4), Individual(genome=[8, 3], fitness=2), Individual(genome=[0, 1], fitness=2), Individual(genome=[0, 9, 3], fitness=3), Individual(genome=[8, 1, 3, 7], fitness=4), Individual(genome=[8, 1, 3, 7], fitness=4), Individual(genome=[0, 3, 4, 7, 9], fitness=5), Individual(genome=[0, 1], fitness=2)]
```

In [54]:

```
logging.getLogger().setLevel(logging.INFO)

NUM_GENERATIONS = 100
OFFSPRING_SIZE = 3
POPULATION_SIZE = 10

for g in range(NUM_GENERATIONS):
    offspring = list()
    for i in range(OFFSPRING_SIZE):
        p1 = tournament(population)
        p2 = tournament(population)
        o = cross_over(p1.genome, p2.genome)
        f = fitness(o)
        offspring.append(Individual(o, f))
    population += offspring
    population = sorted(population, key=lambda i: i.fitness,
reverse=True)[:POPULATION_SIZE]
```

```
logging.info(f"population is {population}")
INFO:root:population is [Individual(genome=[1, 4, 5, 6, 0, 3, 4, 7, 9], fitness=7), Individual(genome=[1, 4, 5, 6, 0, 3, 4, 7, 9], fitness=7), Individual(genome=[8, 3, 0, 3, 4, 7, 9, 9, 2, 6], fitness=6), Individual(genome=[1, 4, 5, 6, 8, 3, 0, 3, 4, 7, 9, 9, 2, 6], fitness=6), Individual(genome=[0, 3, 4, 7, 9, 9, 2, 6, 1, 4, 5, 6], fitness=6), Individual(genome=[8, 3, 0, 3, 4, 7, 9, 9, 2, 6, 1, 4, 5, 6], fitness=6), Individual(genome=[9, 2, 6, 1, 4, 5, 6, 0, 3, 4, 7, 9], fitness=6), Individual(genome=[0, 3, 4, 7, 9], fitness=5), Individual(genome=[0, 1, 0, 3, 4, 7, 9], fitness=5), Individual(genome=[9, 2, 6, 1, 4, 5, 6], fitness=5)]
```

Lab 3

Task

Write agents able to play *Nim*, with an arbitrary number of rows and an upper bound k on the number of objects that can be removed in a turn (a.k.a., *subtraction game*).

The player taking the last object wins.

- Task3.1: An agent using fixed rules based on *nim-sum* (i.e., an *expert system*)
- Task3.2: An agent using evolved rules
- Task3.3: An agent using minmax
- Task3.4: An agent using reinforcement learning

Instructions

- Create the directory `lab3` inside the course repo
- Put a `README.md` and your solution (all the files, code and auxiliary data if needed)

Group Members: Milad Zakhireh (300708), Masoud Karimi (300283)

main.py

```
from
Nim
import
MyNim

from RowObjectsPair import RowObjectsPair
from copy import deepcopy
x = MyNim(3, 1)
s0 = (0,1)
r0, o0 = s0
srtgy0 = RowObjectsPair(r0, o0)
# 3.1 & 3.2
# MyNim.doNimming(x,{"p":0.61})
```

```

# -----
print(x.inferedStatus["possible_moves"])
# 3.3
state0 = deepcopy( x._rows )
# x.inferedStatus["possible_moves"] -= {s0}
state1 = x.staticNimming(x._rows, srtgy0)
print(x.minMax(state0, state1))

```

Nim.py

```

import
itertools

from operator import xor
import random
from RowObjectsPair import RowObjectsPair
from copy import deepcopy
from typing import Callable

class MyNim:

    def __init__(self, numRows : int, k : int = None )-> None:
        self.countOfRows = numRows
        self._rows = [(i*2)+1 for i in range(numRows)]
        self.copyOfRows = list()
        self._k = k
        self.inferedStatus = dict()
        self.inferedStatusForCopiedRows = dict()
        self.inferableInformation()

    def __bool__(self):
        return sum(self._rows) > 0

    def __str__(self):
        return "<" + " ".join(str(_) for _ in self._rows) + ">"

    @property
    def rows(self) -> tuple:
        return tuple(self._rows)

```

```

@property
def k(self) -> int:
    return self._k

def deepCopyOf_rows(self):
    self.copyOfRows = deepcopy(self._rows)
    return self.copyOfRows

# 3.1, 3,2
def nimming(self, strategy :RowObjectsPair, ifFindingNimSum = None) ->
None:
    if not ifFindingNimSum:
        row = strategy.row
        numObjects = strategy.numberOfObject
        # assert self._rows[row] >= numObjects
        # assert self._k is None or numObjects <= self._k
        self._rows[row] -= numObjects
    else:
        row = strategy.row
        numObjects = strategy.numberOfObject
        self.copyOfRows[row] -= numObjects

def staticNimming(self, rows, strategy :RowObjectsPair):
    row = strategy.row
    numObjects = strategy.numberOfObject
    rows[row] -= numObjects
    return deepcopy(rows)

def doNimming(self, gemone:dict = None):
    if gemone is not None:
        strategies = [self.properActionToPerformNimSum,
self.evaluate(gemone)]
    else:
        strategies = [self.properActionToPerformNimSum,
self.pureRandomStrategy]
    print(f"initial board : {self._rows}")
    player = 0
    while self:
        # if not player:
        #     strategy = strategies[player]()
        # else:
        strategy = strategies[player]()
        self.nimming(strategy)

```

```

        print(f" Player: {player}, Board: {self._rows}")
        player = 1 - player
        print(f"winner is player {player}")

def inferableInformation(self, evaluation = False) -> None:
    if not evaluation:
        self.inferedStatus["oddRows"] = [count for count,rowValue in
enumerate(self._rows) if rowValue%2]
        self.inferedStatus["evenRows"] = [count for count, rowValue in
enumerate(self._rows) if not rowValue % 2]
        self.inferedStatus["nimSum"] = self.nimSum()
        self.inferedStatus["longets"] = self._rows.index(max(self._rows,
key= lambda i:i))
        # self.inferedStatus["shortest"] = self._rows.index(min(i for i in
self._rows if i > 0))
        self.inferedStatus["ifRowsAreEven"] = len([count for
count,rowValue in enumerate(self._rows) if rowValue%2])%2 == 0
        self.inferedStatus["possible_moves"] = set([(r, a1) for r,i in
enumerate(self._rows) for a1 in range(i + 1) if a1 >= self.k and a1 <= a1 + 1])
    else:
        self.inferedStatusForCopiedRows["oddRows"] = [count for count,
rowValue in enumerate(self._rows) if rowValue % 2]
        self.inferedStatusForCopiedRows["evenRows"] = [count for count,
rowValue in enumerate(self._rows) if not rowValue % 2]
        self.inferedStatusForCopiedRows["nimSum"] = self.nimSum()
        self.inferedStatusForCopiedRows["longets"] =
self._rows.index(max(self._rows, key=lambda i: i))
        # self.inferedStatusForCopiedRows["shortest"] =
self._rows.index(min(i for i in self._rows if i > 0))
        self.inferedStatusForCopiedRows["ifRowsAreEven"] = len([count for
count, rowValue in enumerate(self._rows) if rowValue % 2]) % 2 == 0
        self.inferedStatusForCopiedRows["possible_moves"] = set([(r, a1)
for r, i in enumerate(self._rows) for a1 in range(i + 1) if a1 >= self.k and a1
<= a1 + 1])

def pureRandomStrategy(self)-> RowObjectsPair:
    selectedRow = random.choice([row for row, count in
enumerate(self._rows) if count > 0 ])
    numberOfObjects = random.randint(1,self._rows[selectedRow])
    return RowObjectsPair(selectedRow, numberOfObjects)

```

```

def pureRandomStrategyPluseWhole(self)-> RowObjectsPair:
    selectedRow = random.choice([row for row, count in
    enumerate(self._rows) if count > 0 ])
    numberOfObjects = self._rows[selectedRow]
    return RowObjectsPair(selectedRow, numberOfObjects)

def nimSum(self, ifFindingNimSum = None) -> list:
    if not ifFindingNimSum:
        *_ , result = itertools.accumulate(self._rows, xor)
        return result
    else:
        *_ , result = itertools.accumulate(self.copyOfRows, xor)
        return result

def properActionToPerformNimSum(self) -> RowObjectsPair:
    self.inferableInformation()
    x = self.inferedStatus["nimSum"]
    if x:
        possibleMoves = self.inferedStatus["possible_moves"]

        for rowActionPair in possibleMoves:
            _ = self.deepCopyOf_rows()
            indexOfPossibleObjects, possibleObjects = rowActionPair
            self.nimming(RowObjectsPair(indexOfPossibleObjects,
possibleObjects), self.copyOfRows)
            if not self.nimSum(self.copyOfRows):
                self.copyOfRows = list()
                return RowObjectsPair(indexOfPossibleObjects,
possibleObjects)

            self.copyOfRows = list()
            return self.pureRandomStrategy()
    self.copyOfRows = list()
    return self.pureRandomStrategy()

def evaluate(self, selectionThreshold: dict) -> Callable:
    def evolvable() -> RowObjectsPair:
        self.inferableInformation()
        if random.random() < selectionThreshold["p"]:
            ply = RowObjectsPair(self.inferedStatus["shortest"],
random.randint(1, self._rows[self.inferedStatus["shortest"]]))

```

```

        else:
            ply = RowObjectsPair(self.inferedStatus["longets"],
random.randint(1, self._rows[self.inferedStatus["longets"]]))
            return ply
        return evolvable

```

#3.3 MIN MAX

```

def ifWinner(self, playerState):
    if playerState.count(0) == (self.countOfRows-1):
        return True

def evaluateMinMax(self, p0,p1):

    if self.ifWinner(p0):
        return 1
    elif self.ifWinner(p1):
        return -1
    else:
        return 0

def minMax(self, s0, s1):
    self.inferableInformation()
    possibleActions = self.inferedStatus["possible_moves"]
    eval = self.evaluateMinMax(s0, s1)
    if eval != 0:
        return eval

    evaluation = list()
    for possibleAction in possibleActions :
        s0 = deepcopy(s1)
        if s0[possibleAction[0]]>= possibleAction[1]:
            s0 = self.staticNimming(s0, RowObjectsPair(possibleAction[0],
possibleAction[1]))
        else:
            continue
        eval = self.minMax(s1, s0)
        if type(eval) == tuple:
            _, eval = eval
        evaluation.append((possibleAction,-eval))

```

```
return max(evaluation, key=lambda k: k[1])
```

RowObjectsPair.py

```
class
RowObjectsPair:

    def __init__(self, row, numberOfObject):
        self.row = row
        self.numberOfObject = numberOfObject

    def __str__(self):
        return "<STRATEGY= row: " + str(self.row) + ", objects: " +
str(self.numberOfObject)+ ">"
```

Review of Marco Masera on My Work for Lab 3



Marco-Masera commented on Dec 18, 2022



Hi. I leave the review for lab3.

General considerations

The lack of documentation and comments in the code makes it hard to understand how the code works; the choice of having a single class implement almost all of the logic, while being an interesting design choice, makes it even harder.

It would also be useful to have some tests to run.

Perhaps you were planning to add these later, in case I'll update the review.

That said I liked the idea of having a more "generic" agent and having evolving rules and nimsum as simple strategies nested inside it. It's probably not the cleanest or more readable way to do it, but it's elegant in a way.

Nim Sum:

It's a working implementation of the NimSum strategy for games with $K = \text{None}$, while it does not extend to games with K defined. You could implement NimSum for games with K defined by computing the xor-sum on a transformed list where each value is the module of the number of elements in the row and $k+1$: $[n \% (k+1) \text{ for } n \text{ in rows}]$

Evolving rules:

The implementation of the evolving rules part seems incomplete. The class implements it with a genome dict that contains a single value p , specifying the probability of choosing between two different rules. But the code that manages the creation and evolution of individuals isn't there.

Min Max:

I find this implementation confusing and I'm not 100% sure it's correct.

I don't really understand why it takes two states as input, and when recursively calling itself it pass its second argument as first argument to the next call, while the new status is passed as the second argument. The most straightforward implementation would be to pass one state of the game to the function. Again I think it would be really useful to provide some comments or documentation to explain these choices.

Reinforcement learning agent:

Missing.

Conclusion:

This review isn't the most useful; unfortunately I had a hard time navigating inside the code without documentation, comments or tests to run.

The project of course is not complete so this makes the review even less useful.

What I can say is that NimSum seems correct, the evolving rules part seems *on the good rail* but needs to have the part of the code that actually does the evolution; you can start from the code you wrote for Lab2, which was good.

You could also experiment with more rules to choose from or a more complex combination of different rules.

The MinMax part is a little bit confusing and I'd suggest writing a more straightforward minmax recursive function.

Final Project (Quarto)

The project is implemented with the guidance of my dear friend Masoud Karimi who helped me so much to improve my background knowledge alongside with the lectures all throughout the course.

main.py

```
# Free for
personal or
classroom
use; see
'LICENSE.md'
for details.

# https://github.com/squillero/computational-intelligence

import logging
import argparse
```

```

import random
import quarto
import numpy
from numpy import save
import matplotlib.pyplot as plot
import copy

class RandomPlayer(quarto.Player):
    """Random player"""

    def __init__(self, quarto: quarto.Quarto, learningPhase) -> None:
        super().__init__(quarto, learningPhase)

    def choose_piece(self) -> int:
        return random.randint(0, 15)

    def place_piece(self) -> tuple[int, int]:
        return random.randint(0, 3), random.randint(0, 3)

def main():
    # game = quarto.Quarto()
    # playerRL = quarto.Player(game, True)
    # playerR = RandomPlayer(game, True)
    # game.set_players((playerRL, playerR))
    # game.getAvailablePieces()
    # game.getFreePlaces()
    # placeReward, pieceReward =
game.learnModelParams(copy.deepcopy(game.availablePieces))
    # save("finalPlaceWeight2.npy", placeReward)
    # save("finalPieceWeight2.npy", pieceReward)

    pieceReward = numpy.load('finalPieceWeight2.npy', allow_pickle=True)
    pieceReward = dict(enumerate(pieceReward.flatten(), 1))
    pieceReward = pieceReward[1]
    placeReward = numpy.load('finalPlaceWeight2.npy', allow_pickle=True)
    placeReward = dict(enumerate(placeReward.flatten(), 1))
    placeReward = placeReward[1]
    print(pieceReward)
    print(placeReward)

    # -----/\ Battle Field
    /\-----

```

```

rounds = 800
runIndex = 10
RLProportion = []
Randomproportion = []
valueX = []

for j in range(runIndex):
    RL = 0
    rand = 0
    draw = 0
    for i in range(rounds):
        battleField = quarto.Quarto()
        agentRL = quarto.Player(battleField, False)
        agentRandom = RandomPlayer(battleField, False)
        agentRL.pieceWeightDict = pieceReward
        agentRL.placeWeightDict = placeReward
        battleField.set_players((agentRL, agentRandom))
        winner = battleField.run()
        if winner == 0:
            RL += 1
        elif winner == 1:
            rand += 1
        else:
            draw += 1
    valueX.append(j)
    Proportion = RL / rounds
    RLProportion.append(Proportion)
    Randomproportion.append(1 - Proportion)
    print(f"RL rate ={RL / rounds} and random rate = {rand / rounds}")

plot.semilogy(valueX, RLProportion, "b")
plot.axhline(y=0.5, color='r', linestyle='--')
plot.xlim([-1.0, runIndex])
plot.ylim([0, 1])
plot.title("Precentage Of RL Agent Wins")
plot.legend(["RL agent", "Mean"])
plot.xlabel("Runs")
plot.ylabel(f"{rounds}-Rounds")

plot.show()

if __name__ == '__main__':

```

```

parser = argparse.ArgumentParser()
parser.add_argument('-v', '--verbose', action='count', default=0,
help='increase log verbosity')
parser.add_argument('-d',
                    '--debug',
                    action='store_const',
                    dest='verbose',
                    const=2,
                    help='log debug messages (same as -vv)')
args = parser.parse_args()

if args.verbose == 0:
    logging.getLogger().setLevel(level=logging.WARNING)
elif args.verbose == 1:
    logging.getLogger().setLevel(level=logging.INFO)
elif args.verbose == 2:
    logging.getLogger().setLevel(level=logging.DEBUG)

main()

```

quarto/objects.py

```

# Free for
personal or
classroom
use; see
'LICENSE.md'
for details.

# https://github.com/squillero/computational-intelligence
import numpy
import numpy as np
from abc import abstractmethod
import copy
import random
from functools import cmp_to_key

class Player(object):

```

```

def __init__(self, quarto, learningPhase, rounds=500, alpha=0.2,
randomFactor=0.1) -> None:
    self.__quarto = quarto
    # self.__board = quarto.get_board_status
    self.historyOfMoves = [] # place, reward
    self.historyOfPiece = [] # piece, reward
    self.alpha = alpha
    self.randomFactor = randomFactor
    self.placeWeightDict = {}
    self.pieceWeightDict = {}
    self.learningPhase = learningPhase
    self.setReward(self.__quarto)
    self.rounds = rounds
    self.currentRound = 0
    self.tempSelected = set()

def setReward(self, quarto):
    if self.learningPhase:
        for i, row in enumerate(quarto.get_board_status()):
            for j, col in enumerate(row):
                self.placeWeightDict[(i, j)] =
np.random.uniform(low=1.0, high=0.1)

        for pI,_ in enumerate(quarto.get_all_pieces()):
            self.pieceWeightDict[pI] = np.random.uniform(low=1.0,
high=0.1)

@abstractmethod
def choose_piece(self) -> int:
    # self.__quarto.boa
    self.__quarto.getAvailablePieces()
    maxG = -10e15
    maxGrule = -10e9

    next_piece_index = None
    randomN = np.random.random()
    if self.learningPhase:

        if randomN < self.randomFactor:
            next_piece_index = random.randint(0, 15)

    else:

```

```

        pieceWeight = []
        for pieceIndex in self.__quarto.availablePieces:
            selectedPieceWeight =
self.__quarto.calcPieceWeight(pieceIndex)
            pieceWeight.append((pieceIndex, selectedPieceWeight))

        sortedList = sorted(pieceWeight,
key=cmp_to_key(self.compare), reverse=True)
        # the worst
        next_piece_index_help, _ = sortedList[0]
        # next_piece_index = next_piece_index_help

        for pieceIndex, weight in pieceWeight:
            new_piece_index = pieceIndex
            tempW = weight
            if self.pieceWeightDict[new_piece_index] >= maxG or
tempW >= maxGrule:
                next_piece_index = new_piece_index
                maxG = self.pieceWeightDict[pieceIndex]
                maxGrule = tempW

        if next_piece_index is None:
            print("help piece")
            next_piece_index = next_piece_index_help

    else:
        for pieceIndex in self.__quarto.availablePieces:
            new_piece_index = pieceIndex
            if self.pieceWeightDict[new_piece_index] >= maxG:
                next_piece_index = new_piece_index
                maxG = self.pieceWeightDict[pieceIndex]

    return next_piece_index

def update_player_board(self):
    self.__board = self.__quarto.get_board_status

def compare(self, pair1, pair2):
    _, fitness1 = pair1
    _, fitness2 = pair2

```

```

        if fitness2 > fitness1:
            return -1
        else:
            return 1

@abstractmethod
def place_piece(self) -> tuple[int, int]:
    # self.__board= self.__quarto.get_board_status()
    pI = self.__quarto.get_selected_piece()
    self.__quarto.updatePlayedPiecePlace()
    self.__quarto.getFreePlaces()
    maxG = -10e15
    maxGrun = -10e15
    next_move = None
    randomN = np.random.random()

    if self.learningPhase:

        if randomN < self.randomFactor:
            freePLaces = [tuple(i) for i in self.__quarto.allFreePlaces]
            next_move = random.choice(freePLaces)
        else:
            pairXYfeasibility = []
            for placeXY in self.__quarto.allFreePlaces:
                weight = self.__quarto.calcPlaceWeight(pI, placeXY)
                pairXYfeasibility.append((placeXY, weight))
            sortedList = sorted(pairXYfeasibility,
key=cmp_to_key(self.compare), reverse=True)
            next_move_help, _ = sortedList[0]
            next_move = next_move_help

            for action, palceWeight in pairXYfeasibility:
                new_state = action
                # and palceWeight >= maxGrun
                if self.placeWeightDict[new_state] >= maxG and
palceWeight >= maxGrun:
                    next_move = new_state
                    maxG = self.placeWeightDict[new_state]
                    maxGrun = palceWeight

            if next_move is None:
                print("place help")
                next_move = next_move_help

```

```

        else:
            for action in self.__quarto.allFreePlaces:
                new_state = action
                if self.placeWeightDict[new_state] >= maxG:
                    next_move = new_state
                    maxG = self.placeWeightDict[new_state]
            return next_move

    def get_game(self):
        return self.__quarto

    # def updateMovesHistory(self, place):
    #     if self.__quarto.assignReward() != 0:
    #         reward = 1 / self.__quarto.assignReward()
    #     else:
    #         reward = self.__quarto.assignReward()
    #     self.historyOfMoves.append((place, reward))

    def updateHistoryOfMoves(self, place):
        reward = self.__quarto.assignReward()
        self.historyOfMoves.append((place, reward))

    def updateHistoryOfPiece(self, piece):
        reward = self.__quarto.assignReward()
        self.historyOfPiece.append((piece, reward))

class CharacCounter(object):
    def __init__(self, selectedPieceCharacteristic = None):
        if selectedPieceCharacteristic == None:
            self.high = 0
            self.notHigh = 0
            self.colored = 0
            self.notColored = 0
            self.solid = 0
            self.notSolid = 0
            self.square = 0
            self.notSquare = 0
        else:
            if selectedPieceCharacteristic.HIGH:
                self.high = 1

```



```

        self.notHigh = 0
    else:
        self.notHigh = 1
        self.high = 0

    if selectedPieceCharacteristic.COLOURED:
        self.colored = 1
        self.notColored = 0
    else:
        self.notColored = 1
        self.colored = 0

    if selectedPieceCharacteristic.SOLID:
        self.solid = 1
        self.notSolid = 0
    else:
        self.notSolid = 1
        self.solid = 0

    if selectedPieceCharacteristic.SQUARE:
        self.square = 1
        self.notSquare = 0
    else:
        self.notSquare = 1
        self.square = 0

class Piece(object):

    def __init__(self, high: bool, coloured: bool, solid: bool, square:
bool) -> None:
        self.HIGH = high
        self.COLOURED = coloured
        self.SOLID = solid
        self.SQUARE = square

class Quarto(object):

    MAX_PLAYERS = 2
    BOARD_SIDE = 4

    def __init__(self) -> None:
        self.__players = ()

```

```

        self.reset()

    def reset(self):
        self.__board = np.ones(shape=(self.BOARD_SIDE, self.BOARD_SIDE),
dtype=int) * -1
        self.__pieces = []
        self.__pieces.append(Piece(False, False, False, False)) # 0
        self.__pieces.append(Piece(False, False, False, True)) # 1
        self.__pieces.append(Piece(False, False, True, False)) # 2
        self.__pieces.append(Piece(False, False, True, True)) # 3
        self.__pieces.append(Piece(False, True, False, False)) # 4
        self.__pieces.append(Piece(False, True, False, True)) # 5
        self.__pieces.append(Piece(False, True, True, False)) # 6
        self.__pieces.append(Piece(False, True, True, True)) # 7
        self.__pieces.append(Piece(True, False, False, False)) # 8
        self.__pieces.append(Piece(True, False, False, True)) # 9
        self.__pieces.append(Piece(True, False, True, False)) # 10
        self.__pieces.append(Piece(True, False, True, True)) # 11
        self.__pieces.append(Piece(True, True, False, False)) # 12
        self.__pieces.append(Piece(True, True, False, True)) # 13
        self.__pieces.append(Piece(True, True, True, False)) # 14
        self.__pieces.append(Piece(True, True, True, True)) # 15
        self.__current_player = 0
        self.__selected_piece_index = -1
        self.getFreePlaces()
        self.learningPhase = False
        self.allFreePlaces = None
        self.availablePieces = None

    def set_players(self, players: tuple[Player, Player]):
        self.__players = players

    def updatePlayedPiecePlace(self):
        self.getAvailablePieces()
        self.getFreePlaces()

    def getAvailablePieces(self):
        allIndices = set([0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14,
15])
        board = self.get_board_status().ravel()
        allSelectedPieces = set(board[board > -1])
        self.availablePieces = list(allIndices - allSelectedPieces)

```

```

def getFreePlaces(self):
    self.freePlaces = np.where(self.__board == -1)
    self.allFreePlaces = zip(self.freePlaces[0], self.freePlaces[1])


def compare(self, pair1, pair2):
    _, fitness1 = pair1
    _, fitness2 = pair2

    if fitness2 > fitness1:
        return -1
    else:
        return 1


def select(self, pieceIndex: int) -> bool:
    """
    select a piece. Returns True on success
    """
    if pieceIndex not in self.__board:
        self.__selected_piece_index = pieceIndex
        return True
    return False


def place(self, x: int, y: int) -> bool:
    """
    Place piece in coordinates (x, y). Returns true on success
    """
    if self.__placeable(x, y):
        self.__board[x, y] = self.__selected_piece_index
        return True
    return False


def __placeable(self, x: int, y: int) -> bool:
    return not (y < 0 or x < 0 or x > 3 or y > 3 or self.__board[x, y]
>= 0)


def print(self):
    """
    Print the __board
    """
    for row in self.__board:
        print("\n -----")

```

```

        print("|", end="")
        for element in row:
            print(f" {element: >2}", end=" |")
        print("\n ----- \n")
        print(f"Selected piece: {self.__selected_piece_index}\n")

def get_piece_characteristics(self, index: int) -> Piece:
    """
    Gets characteristics of a piece (index-based)
    """
    return copy.deepcopy(self.__pieces[index])

def get_board_status(self) -> np.ndarray:
    """
    Get the current __board status (__pieces are represented by index)
    """
    return copy.deepcopy(self.__board)

def get_all_pieces(self):
    return copy.deepcopy(self.__pieces)

def get_selected_piece(self) -> int:
    """
    Get index of selected piece
    """
    return copy.deepcopy(self.__selected_piece_index)

def __check_horizontal(self) -> int:
    hDict = dict()
    hDict[0] = None
    hDict[1] = None
    hDict[2] = None
    hDict[3] = None

    for i in range(self.BOARD_SIDE):
        initList = CharacCounter()

        high_values = [
            elem for elem in self.__board[i] if elem >= 0 and
self.__pieces[elem].HIGH
        ]
        initList.high = len(high_values)

```

```

        coloured_values = [
            elem for elem in self.__board[i] if elem >= 0 and
self.__pieces[elem].COLOURED
        ]
        initList.colored = len(coloured_values)

        solid_values = [
            elem for elem in self.__board[i] if elem >= 0 and
self.__pieces[elem].SOLID
        ]
        initList.solid = len(solid_values)

        square_values = [
            elem for elem in self.__board[i] if elem >= 0 and
self.__pieces[elem].SQUARE
        ]
        initList.square = len(square_values)

        low_values = [
            elem for elem in self.__board[i] if elem >= 0 and not
self.__pieces[elem].HIGH
        ]
        initList.notHigh = len(low_values)

        noncolor_values = [
            elem for elem in self.__board[i] if elem >= 0 and not
self.__pieces[elem].COLOURED
        ]
        initList.notColored = len(noncolor_values)

        hollow_values = [
            elem for elem in self.__board[i] if elem >= 0 and not
self.__pieces[elem].SOLID
        ]
        initList.notSolid = len(hollow_values)

        circle_values = [
            elem for elem in self.__board[i] if elem >= 0 and not
self.__pieces[elem].SQUARE
        ]
        initList.notSquare = len(circle_values)
        hDict[i] = initList

```

```

        if len(high_values) == self.BOARD_SIDE or len(
            coloured_values
        ) == self.BOARD_SIDE or len(solid_values) == self.BOARD_SIDE or
len(
            square_values) == self.BOARD_SIDE or len(low_values) ==
self.BOARD_SIDE or len(
            noncolor_values) == self.BOARD_SIDE or len(
            hollow_values) == self.BOARD_SIDE or len(
            circle_values) == self.BOARD_SIDE:
            return self.__current_player, None

    return -1, hDict

def __check_vertical(self):
    vDict = dict()
    vDict[0] = None
    vDict[1] = None
    vDict[2] = None
    vDict[3] = None

    for i in range(self.BOARD_SIDE):
        # counts the total value of hight are selected
        initList = CharacCounter()
        high_values = [
            elem for elem in self.__board[:, i] if elem >= 0 and
self.__pieces[elem].HIGH
        ]
        initList.high = len(high_values)

        coloured_values = [
            elem for elem in self.__board[:, i] if elem >= 0 and
self.__pieces[elem].COLOURED
        ]
        initList.colored = len(coloured_values)

        solid_values = [
            elem for elem in self.__board[:, i] if elem >= 0 and
self.__pieces[elem].SOLID
        ]
        initList.solid = len(solid_values)

        square_values = [

```

```

        elem for elem in self.__board[:, i] if elem >= 0 and
self.__pieces[elem].SQUARE
    ]
    initList.square = len(square_values)

    low_values = [
        elem for elem in self.__board[:, i] if elem >= 0 and not
self.__pieces[elem].HIGH
    ]
    initList.notHigh = len(low_values)

    noncolor_values = [
        elem for elem in self.__board[:, i] if elem >= 0 and not
self.__pieces[elem].COLOURED
    ]
    initList.notColored = len(noncolor_values)

    hollow_values = [
        elem for elem in self.__board[:, i] if elem >= 0 and not
self.__pieces[elem].SOLID
    ]
    initList.notSolid = len(hollow_values)

    circle_values = [
        elem for elem in self.__board[:, i] if elem >= 0 and not
self.__pieces[elem].SQUARE
    ]
    initList.notSquare = len(circle_values)

    vDict[i] = initList

    if len(high_values) == self.BOARD_SIDE or len(
        coloured_values
    ) == self.BOARD_SIDE or len(solid_values) == self.BOARD_SIDE or
len(
    square_values) == self.BOARD_SIDE or len(low_values) ==
self.BOARD_SIDE or len(
    noncolor_values) == self.BOARD_SIDE or len(
    hollow_values) == self.BOARD_SIDE or len(
    circle_values) == self.BOARD_SIDE:
        return self.__current_player, None
    return -1, vDict

```

```

def new_check_diagonal(self):
    LToRdiagDict = dict()
    LToRdiagDict[0] = None
    LToRdiagDict[1] = None
    LToRdiagDict[2] = None
    LToRdiagDict[3] = None

    for i in range(self.BOARD_SIDE):
        # if self.__board[i, i] < 0:
        #     break
        high_values = []
        coloured_values = []
        solid_values = []
        square_values = []
        low_values = []
        noncolor_values = []
        hollow_values = []
        circle_values = []
        LTRinitiallist = CharacCounter()

        if self.__pieces[self.__board[i, i]].HIGH:
            if self.__board[i, i] != -1:
                high_values.append(self.__board[i, i])
                LTRinitiallist.high = len(high_values)
            else:
                if self.__board[i, i] != -1:
                    low_values.append(self.__board[i, i])
                    LTRinitiallist.notHigh = len(low_values)

        if self.__pieces[self.__board[i, i]].COLOURED:
            if self.__board[i, i] != -1:
                coloured_values.append(self.__board[i, i])
                LTRinitiallist.colored = len(coloured_values)
            else:
                if self.__board[i, i] != -1:
                    noncolor_values.append(self.__board[i, i])
                    LTRinitiallist.notColored = len(noncolor_values)

        if self.__pieces[self.__board[i, i]].SOLID:
            if self.__board[i, i] != -1:
                solid_values.append(self.__board[i, i])
                LTRinitiallist.solid = len(solid_values)
            else:

```



```

        if self.__board[i, i] != -1:
            hollow_values.append(self.__board[i, i])
            LTRinitiallist.notSolid = len(hollow_values)

    if self.__pieces[self.__board[i, i]].SQUARE:
        if self.__board[i, i] != -1:
            square_values.append(self.__board[i, i])
            LTRinitiallist.square = len(square_values)
    else:
        if self.__board[i, i] != -1:
            circle_values.append(self.__board[i, i])
            LTRinitiallist.notSquare = len(circle_values)

    LToRdiagDict[i] = LTRinitiallist

    if len(high_values) == self.BOARD_SIDE or len(coloured_values) ==
self.BOARD_SIDE or len(
        solid_values) == self.BOARD_SIDE or len(square_values) ==
self.BOARD_SIDE or len(
        low_values
    ) == self.BOARD_SIDE or len(noncolor_values) == self.BOARD_SIDE or
len(
        hollow_values) == self.BOARD_SIDE or len(circle_values) ==
self.BOARD_SIDE:
        return self.__current_player, None

RToLdiagDict = dict()
RToLdiagDict[0] = None
RToLdiagDict[1] = None
RToLdiagDict[2] = None
RToLdiagDict[3] = None

for i in range(self.BOARD_SIDE):
    # if self.__board[i, self.BOARD_SIDE - 1 - i] < 0:
    #     break
    high_values = []
    coloured_values = []
    solid_values = []
    square_values = []
    low_values = []
    noncolor_values = []
    hollow_values = []
    circle_values = []

```

```

RTLinitiallist = CharacCounter()
if self.__pieces[self.__board[i, self.BOARD_SIDE - 1 - i]].HIGH:
    if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
        high_values.append(self.__board[i, self.BOARD_SIDE - 1 -
i])

        RTLinitiallist.high = len(high_values)
    else:
        if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
            low_values.append(self.__board[i, self.BOARD_SIDE - 1 -
i])

            RTLinitiallist.notHigh = len(low_values)

        if self.__pieces[self.__board[i, self.BOARD_SIDE - 1 -
i]].COLOURED:
            if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                coloured_values.append(
                    self.__board[i, self.BOARD_SIDE - 1 - i])
                RTLinitiallist.colored = len(coloured_values)
            else:
                if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                    noncolor_values.append(
                        self.__board[i, self.BOARD_SIDE - 1 - i])
                    RTLinitiallist.notColored = len(noncolor_values)

        if self.__pieces[self.__board[i, self.BOARD_SIDE - 1 -
i]].SOLID:
            if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                solid_values.append(self.__board[i, self.BOARD_SIDE - 1
- i])

                RTLinitiallist.solid = len(solid_values)
            else:
                if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                    hollow_values.append(self.__board[i, self.BOARD_SIDE - 1
- i])

                    RTLinitiallist.notSolid = len(hollow_values)

        if self.__pieces[self.__board[i, self.BOARD_SIDE - 1 -
i]].SQUARE:
            if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                square_values.append(self.__board[i, self.BOARD_SIDE - 1
- i])

                RTLinitiallist.square = len(square_values)
        else:

```

```

        if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
            circle_values.append(self.__board[i, self.BOARD_SIDE - 1
- i])

            RTLinitiallist.notSquare = len(circle_values)

        RToldiagDict[i] = RTLinitiallist

        if len(high_values) == self.BOARD_SIDE or len(coloured_values) ==
self.BOARD_SIDE or len(
            solid_values) == self.BOARD_SIDE or len(square_values) ==
self.BOARD_SIDE or len(
            low_values
        ) == self.BOARD_SIDE or len(noncolor_values) == self.BOARD_SIDE or
len(
            hollow_values) == self.BOARD_SIDE or len(circle_values) ==
self.BOARD_SIDE:
            return self.__current_player, None

        retunDict = {
            "LTR": LToRdiagDict,
            "RTL": RToldiagDict
        }
        return -1, retunDict

def calcPieceWeight(self, pieceIndex):
    badPrise = -10e10
    copyOfBoard = copy.deepcopy(self.__board)
    self.updatePlayedPiecePlace()
    thisFreePlacesRewardForGivenPiece = []
    for possition in self.allFreePlaces:
        x, y = possition
        self.__board = copy.deepcopy(copyOfBoard)
        self.__board[x, y] = pieceIndex
        _, dictH = self.__check_horizontal()
        _, dictV = self.__check_vertical()
        _, dictD = self.new_check_diagonal()
        dictLTR, dictRTL = dictD["LTR"], dictD["RTL"]
        if dictH is not None and dictV is not None and dictLTR is not
None and dictRTL is not None:
            propertyH = dictH[x]
            propertyV = dictV[y]
            thisStepH = numpy.array(

```

```

        [propertyH.high, propertyH.notHigh, propertyH.colored,
propertyH.notColored, propertyH.solid,
        propertyH.notSolid, propertyH.square,
        propertyH.notSquare])
    thisStepV = numpy.array(
        [propertyV.high, propertyV.notHigh, propertyV.colored,
propertyV.notColored, propertyV.solid,
        propertyV.notSolid, propertyV.square,
        propertyV.notSquare])
    l1 = dictLTR[0]
    l2 = dictLTR[1]
    l3 = dictLTR[2]
    l4 = dictLTR[3]
    thisStepL = numpy.array(
        [l1.high + l2.high + l3.high + l4.high, l1.notHigh +
12.notHigh + l3.notHigh + l4.notHigh,
        l1.colored + l2.colored + l3.colored + l4.colored,
        l1.notColored + l2.notColored + l3.notColored +
14.notColored,
        l1.solid + l2.solid + l3.solid + l4.solid, l1.notSolid
+ l2.notSolid + l3.notSolid + l4.notSolid,
        l1.square + l2.square + l3.square + l4.square,
        l1.notSquare + l2.notSquare + l3.notSquare +
14.notSquare])

    r1 = dictRTL[0]
    r2 = dictRTL[1]
    r3 = dictRTL[2]
    r4 = dictRTL[3]
    thisStepR = numpy.array(
        [r1.high + r2.high + r3.high + r4.high, r1.notHigh +
r2.notHigh + r3.notHigh + r4.notHigh,
        r1.colored + r2.colored + r3.colored + r4.colored,
        r1.notColored + r2.notColored + r3.notColored +
r4.notColored,
        r1.solid + r2.solid + r3.solid + r4.solid, r1.notSolid
+ r2.notSolid + r3.notSolid + r4.notSolid,
        r1.square + r2.square + r3.square + r4.square,
        r1.notSquare + r2.notSquare + r3.notSquare +
r4.notSquare])

    h4 = len(thisStepH[thisStepH == 4])
    h3 = len(thisStepH[thisStepH == 3])

```

```

        h2 = len(thisStepH[thisStepH == 2])
        h1 = len(thisStepH[thisStepH == 1])
        h0 = len(thisStepH[thisStepH == 0])

        v4 = len(thisStepV[thisStepV == 4])
        v3 = len(thisStepV[thisStepV == 3])
        v2 = len(thisStepV[thisStepV == 2])
        v1 = len(thisStepV[thisStepV == 1])
        v0 = len(thisStepV[thisStepV == 0])

        l4 = len(thisStepL[thisStepL == 4])
        l3 = len(thisStepL[thisStepL == 3])
        l2 = len(thisStepL[thisStepL == 2])
        l1 = len(thisStepL[thisStepL == 1])
        l0 = len(thisStepL[thisStepL == 0])

        r4 = len(thisStepR[thisStepR == 4])
        r3 = len(thisStepR[thisStepR == 3])
        r2 = len(thisStepR[thisStepR == 2])
        r1 = len(thisStepR[thisStepR == 1])
        r0 = len(thisStepR[thisStepR == 0])

        temp = [h4, h3, h2, h1, h0, v4, v3, v2, v1, v0, l4, l3, l2,
11, l0, r4, r3, r2, r1, r0]
        forNormalize = numpy.array(temp)
        forNormalize = forNormalize[forNormalize != 0]
        totalValued = sum(forNormalize)

        if r4 != 0 or l4 != 0 or v4 != 0 or h4 != 0:
            thisFreePlacesRewardForGivenPiece.append((x, y),
badPrise))
        else:
            newPrise = 5 * (h3 + v3 + l3 + r3) + 2 * (h2 + v2 + l2 +
r2) + 1 * (h1 + v1 + l1 + r1) + 1 * (
                h0 + v0 + l0 + r0)

            newPrise = newPrise / totalValued
            thisFreePlacesRewardForGivenPiece.append((x, y),
newPrise))
    else:
        self.__board = copy.deepcopy(copyOfBoard)
        return badPrise

```

```

        sortedList = sorted(thisFreePlacesRewardForGivenPiece,
key=cmp_to_key(self.compare), reverse=True)
        worstPlaceForGivenPiece, worstPriseForGivenPiece = sortedList[0]
        self.__board = copy.deepcopy(copyOfBoard)
        return worstPriseForGivenPiece

def __check_diagonal(self):
    LTRdiagDict = dict()
    LTRdiagDict[0] = None
    LTRdiagDict[1] = None
    LTRdiagDict[2] = None
    LTRdiagDict[3] = None
    high_values = []
    coloured_values = []
    solid_values = []
    square_values = []
    low_values = []
    noncolor_values = []
    hollow_values = []
    circle_values = []

    for i in range(self.BOARD_SIDE):
        # if self.__board[i, i] < 0:
        #     break
        LTRinitiallist = CharacCounter()

        if self.__pieces[self.__board[i, i]].HIGH:
            if self.__board[i, i] != -1:
                high_values.append(self.__board[i, i])
                LTRinitiallist.high = len(high_values)
            else:
                if self.__board[i, i] != -1:
                    low_values.append(self.__board[i, i])
                    LTRinitiallist.notHigh = len(low_values)

        if self.__pieces[self.__board[i, i]].COLOURED:
            if self.__board[i, i] != -1:
                coloured_values.append(self.__board[i, i])
                LTRinitiallist.colored = len(coloured_values)
            else:
                if self.__board[i, i] != -1:
                    noncolor_values.append(self.__board[i, i])
                    LTRinitiallist.notColored = len(noncolor_values)

```

```

        if self.__pieces[self.__board[i, i]].SOLID:
            if self.__board[i, i] != -1:
                solid_values.append(self.__board[i, i])
                LTRinitiallist.solid = len(solid_values)
            else:
                if self.__board[i, i] != -1:
                    hollow_values.append(self.__board[i, i])
                    LTRinitiallist.notSolid = len(hollow_values)

        if self.__pieces[self.__board[i, i]].SQUARE:
            if self.__board[i, i] != -1:
                square_values.append(self.__board[i, i])
                LTRinitiallist.square = len(square_values)
            else:
                if self.__board[i, i] != -1:
                    circle_values.append(self.__board[i, i])
                    LTRinitiallist.notSquare = len(circle_values)

        LTRdiagDict[i] = LTRinitiallist

    if len(high_values) == self.BOARD_SIDE or len(coloured_values) ==
self.BOARD_SIDE or len(
        solid_values) == self.BOARD_SIDE or len(square_values) ==
self.BOARD_SIDE or len(
        low_values
    ) == self.BOARD_SIDE or len(noncolor_values) == self.BOARD_SIDE or
len(
        hollow_values) == self.BOARD_SIDE or len(circle_values) ==
self.BOARD_SIDE:
        return self.__current_player, None

RTLdiagDict = dict()
RTLdiagDict[0] = None
RTLdiagDict[1] = None
RTLdiagDict[2] = None
RTLdiagDict[3] = None
high_values = []
coloured_values = []
solid_values = []
square_values = []
low_values = []
noncolor_values = []

```

```

hollow_values = []
circle_values = []

for i in range(self.BOARD_SIDE):
    # if self.__board[i, self.BOARD_SIDE - 1 - i] < 0:
    #     break
    RTLinitiallist = CharacCounter()
    if self.__pieces[self.__board[i, self.BOARD_SIDE - 1 - i]].HIGH:
        if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
            high_values.append(self.__board[i, self.BOARD_SIDE - 1 -
i])

            RTLinitiallist.high = len(high_values)
        else:
            if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                low_values.append(self.__board[i, self.BOARD_SIDE - 1 -
i])

                RTLinitiallist.notHigh = len(low_values)

            if self.__pieces[self.__board[i, self.BOARD_SIDE - 1 -
i]].COLOURED:
                if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                    coloured_values.append(
                        self.__board[i, self.BOARD_SIDE - 1 - i])
                    RTLinitiallist.colored = len(coloured_values)
                else:
                    if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                        noncolor_values.append(
                            self.__board[i, self.BOARD_SIDE - 1 - i])
                        RTLinitiallist.notColored = len(noncolor_values)

            if self.__pieces[self.__board[i, self.BOARD_SIDE - 1 -
i]].SOLID:
                if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                    solid_values.append(self.__board[i, self.BOARD_SIDE - 1
- i])

                    RTLinitiallist.solid = len(solid_values)
                else:
                    if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                        hollow_values.append(self.__board[i, self.BOARD_SIDE - 1
- i])

                        RTLinitiallist.notSolid = len(hollow_values)

```



```

        if self.__pieces[self.__board[i, self.BOARD_SIDE - 1 -
i]].SQUARE:
            if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                square_values.append(self.__board[i, self.BOARD_SIDE - 1
- i])

                RTLinitiallist.square = len(square_values)
            else:
                if self.__board[i, self.BOARD_SIDE - 1 - i] != -1:
                    circle_values.append(self.__board[i, self.BOARD_SIDE - 1
- i])

                    RTLinitiallist.notSquare = len(circle_values)

        RTLdiagDict[i] = RTLinitiallist

        if len(high_values) == self.BOARD_SIDE or len(coloured_values) ==
self.BOARD_SIDE or len(
            solid_values) == self.BOARD_SIDE or len(square_values) ==
self.BOARD_SIDE or len(
            low_values
        ) == self.BOARD_SIDE or len(noncolor_values) == self.BOARD_SIDE or
len(
            hollow_values) == self.BOARD_SIDE or len(circle_values) ==
self.BOARD_SIDE:
            return self.__current_player, None

        retunDict = {
            "LTR": LTRdiagDict,
            "RTL": RTLdiagDict
        }
        return -1, retunDict

def assignReward(self):
    if not self.check_finished():
        return -100 * int(not self.check_finished())
    else:
        return 10e2

def check_winner(self) -> int:
    ...

    Check who is the winner
    ...

```

```

        checkV, _ = self.__check_vertical()
        checkH, _ = self.__check_horizontal()
        checkD, _ = self.__check_diagonal()

        l = [checkH, checkV, checkD]
        # l = [self.__check_horizontal(), self.__check_vertical(),
self.__check_diagonal()]
        for elem in l:
            if elem >= 0:
                return elem
        return -1

def check_finished(self) -> bool:
    ...

    Check who is the loser
    ...

    for row in self.__board:
        for elem in row:
            if elem == -1:
                return False
    return True

def learnModelParams(self, pieces):
    agentRL = 0
    agentRandom = 0
    draw = 0
    if self.__players[0].learningPhase:
        print("inlearning ")

        for epoch in range(self.__players[0].rounds):
            self.__board = np.ones(shape=(self.BOARD_SIDE,
self.BOARD_SIDE), dtype=int) * -1
            self.availablePieces = pieces
            winner = -1

            ++self.__players[0].currentRound
            while winner < 0 and not self.check_finished():

                self.updatePlayedPiecePlace()
                piece_ok = False
                while not piece_ok:
                    self.updatePlayedPiecePlace()

```

```

        selectedPiece =
self.__players[self.__current_player].choose_piece()
        piece_ok = self.select(selectedPiece)
        if self.__players[0].learningPhase:
            if piece_ok and not bool(self.__current_player):

self.__players[0].updateHistoryOfPiece(self.__selected_piece_index)
        piece_ok = False
        self.__current_player = (self.__current_player + 1) %
self.MAX_PLAYERS
        while not piece_ok:
            self.updatePlayedPiecePlace()
            place =
self.__players[self.__current_player].place_piece()
            x, y = place
            piece_ok = self.place(x, y)
            if self.__players[0].learningPhase:
                if piece_ok and not bool(self.__current_player):

self.__players[0].updateHistoryOfMoves(place)
            winner = self.check_winner()
            print(f"the winner is ={winner}")
            if winner == 0:
                agentRL += 1
            elif winner == 1:
                agentRandom += 1
            else:
                draw += 1
            # or winner == -1
            if winner == 0 :
                self.learn(self.__players[0])
            else:
                self.__players[0].historyOfMoves = []
                self.__players[0].historyOfPiece = []

        print(f"RL wins: {agentRL} and Random wins: {agentRandom} and
Draw is: {draw}, ")
        return self.__players[0].placeWeightDict,
self.__players[0].pieceWeightDict

def learn(self, player):
    target = 0

```

```

        for prev, reward in reversed(player.historyOfMoves):
            player.placeWeightDict[prev] = player.placeWeightDict[prev] +
player.alpha * (target - player.placeWeightDict[prev])
            target += reward

        target = 0
        for prev, reward in reversed(player.historyOfPiece):
            player.pieceWeightDict[prev] = player.pieceWeightDict[prev] +
player.alpha * (target - player.pieceWeightDict[prev])
            target += reward

        player.historyOfMoves= []
        player.historyOfPiece = []

        player.randomFactor -= 10e-5 # decrease random factor each episode
of play

```

```

def calcPlaceWeight(self, pieceIndex, placeXY):
    grandPrise = 10e10
    worstPrise = -grandPrise
    # 2 file 5
    thresholdFor3 = 2
    copyOfBoard = self.get_board_status()
    self.updatePlayedPiecePlace()
    thisFreePlacesRewardForGivenPiece = []

    x, y = placeXY
    self.__board = copy.deepcopy(copyOfBoard)
    self.__board[x, y] = pieceIndex
    _, dictH = self.__check_horizontal()
    _, dictV = self.__check_vertical()
    _, dictD = self.new_check_diagonal()
    dictLTR, dictRTL = dictD["LTR"], dictD["RTL"]
    if dictH is not None and dictV is not None and dictLTR is not None
and dictRTL is not None:
        propertyH = dictH[x]
        propertyV = dictV[y]
        thisStepH = numpy.array(
            [propertyH.high, propertyH.notHigh, propertyH.colored,
propertyH.notColored, propertyH.solid,
            propertyH.notSolid, propertyH.square,
            propertyH.notSquare])
        thisStepV = numpy.array(

```

```

        [propertyV.high, propertyV.notHigh, propertyV.colored,
propertyV.notColored, propertyV.solid,
        propertyV.notSolid, propertyV.square,
        propertyV.notSquare]))
l1 = dictLTR[0]
l2 = dictLTR[1]
l3 = dictLTR[2]
l4 = dictLTR[3]
thisStepL = numpy.array(
    [l1.high + l2.high + l3.high + l4.high, l1.notHigh +
l2.notHigh + l3.notHigh + l4.notHigh,
    l1.colored + l2.colored + l3.colored + l4.colored,
    l1.notColored + l2.notColored + l3.notColored +
l4.notColored,
    l1.solid + l2.solid + l3.solid + l4.solid, l1.notSolid +
l2.notSolid + l3.notSolid + l4.notSolid,
    l1.square + l2.square + l3.square + l4.square,
    l1.notSquare + l2.notSquare + l3.notSquare + l4.notSquare]))

r1 = dictRTL[0]
r2 = dictRTL[1]
r3 = dictRTL[2]
r4 = dictRTL[3]
thisStepR = numpy.array(
    [r1.high + r2.high + r3.high + r4.high, r1.notHigh +
r2.notHigh + r3.notHigh + r4.notHigh,
    r1.colored + r2.colored + r3.colored + r4.colored,
    r1.notColored + r2.notColored + r3.notColored +
r4.notColored,
    r1.solid + r2.solid + r3.solid + r4.solid, r1.notSolid +
r2.notSolid + r3.notSolid + r4.notSolid,
    r1.square + r2.square + r3.square + r4.square,
    r1.notSquare + r2.notSquare + r3.notSquare + r4.notSquare]))

h4 = len(thisStepH[thisStepH == 4])
h3 = len(thisStepH[thisStepH == 3])
h2 = len(thisStepH[thisStepH == 2])
h1 = len(thisStepH[thisStepH == 1])
h0 = len(thisStepH[thisStepH == 0])

v4 = len(thisStepV[thisStepV == 4])
v3 = len(thisStepV[thisStepV == 3])
v2 = len(thisStepV[thisStepV == 2])

```

```

v1 = len(thisStepV[thisStepV == 1])
v0 = len(thisStepV[thisStepV == 0])

l4 = len(thisStepL[thisStepL == 4])
l3 = len(thisStepL[thisStepL == 3])
l2 = len(thisStepL[thisStepL == 2])
l1 = len(thisStepL[thisStepL == 1])
l0 = len(thisStepL[thisStepL == 0])

r4 = len(thisStepR[thisStepR == 4])
r3 = len(thisStepR[thisStepR == 3])
r2 = len(thisStepR[thisStepR == 2])
r1 = len(thisStepR[thisStepR == 1])
r0 = len(thisStepR[thisStepR == 0])

temp = [h4, h3, h2, h1, h0, v4, v3, v2, v1, v0, l4, l3, l2, l1,
l0, r4, r3, r2, r1, r0]
forNormalize = numpy.array(temp)
forNormalize = forNormalize[forNormalize != 0]
totalValued = sum(forNormalize)

if r4 != 0 or l4 != 0 or v4 != 0 or h4 != 0:
    thisFreePlacesRewardForGivenPiece.append(((x, y),
grandPrise))
else:
    if h3 + v3 + l3 + r3 > thresholdFor3:
        thisFreePlacesRewardForGivenPiece.append(((x, y),
worstPrise))
    else:
        newPrise = -1 * (h3 + v3 + l3 + r3) + 2 * (h2 + v2 + l2
+ r2) + 3 * (h1 + v1 + l1 + r1) + 5 * (
            h0 + v0 + l0 + r0)
        newPrise = newPrise / totalValued
        thisFreePlacesRewardForGivenPiece.append(((x, y),
newPrise))

else:
    self.__board = copyOfBoard
    return grandPrise

# sortedList = sorted(thisFreePlacesRewardForGivenPiece,
key=cmp_to_key(self.compare), reverse=True)

```

```

        bestPlaceForGivenPiece, bestPriseForGivenPiece =
thisFreePlacesRewardForGivenPiece[0]
        self.__board = copyOfBoard
        return bestPriseForGivenPiece

def run(self) -> int:
    ...

    Run the game (with output for every move)
    ...

    winner = -1
    while winner < 0 and not self.check_finished():
        # self.print()
        piece_ok = False
        while not piece_ok:
            piece_ok =
self.select(self.__players[self.__current_player].choose_piece())
            piece_ok = False
            self.__current_player = (self.__current_player + 1) %
self.MAX_PLAYERS
        # self.print()
        while not piece_ok:
            x, y = self.__players[self.__current_player].place_piece()
            piece_ok = self.place(x, y)
        winner = self.check_winner()
    # self.print()
    return winner

```