

Search Project

Artificial Intelligence

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1. Identify this problem's environment properties and define an appropriate performance measure.

Performance Measure, Environment, Actuators and Sensors (PEAS):

- **Performance Measure**, that describes the success criteria for the system's behavior: arrive to the parking (the F) following the possible path.
- **Environment**, context where the system perceives and acts: A grid of X rows and Y columns in which each of them can have an O, and X or an F (with walls around) and a car.
- **Actuators or actions**, that are the system's response to each percept set or sequence: Turn 90° to the left or to the right and go ahead if we are in an X or maintain the same way if we are in an O.
- **Sensors**, that are the system's perceptual inputs from the environment at a certain instant: Detect which tile we are in, its type and if it have walls.

This system is **fully observable** that means that the system's sensors give it access to the complete state of the environment at each point in time.

It is a **single system** because operates by itself in the environment.

It is **deterministic** because the next state of the environment is completely determined by the current state and the action executed by the system.

It is **sequential**, because the current decision can affect all future decisions.

It is **static**, because the environment is unchanged while the system is deliberating.

It is **discrete** because it has a limited number of distinct, clearly defined actions.

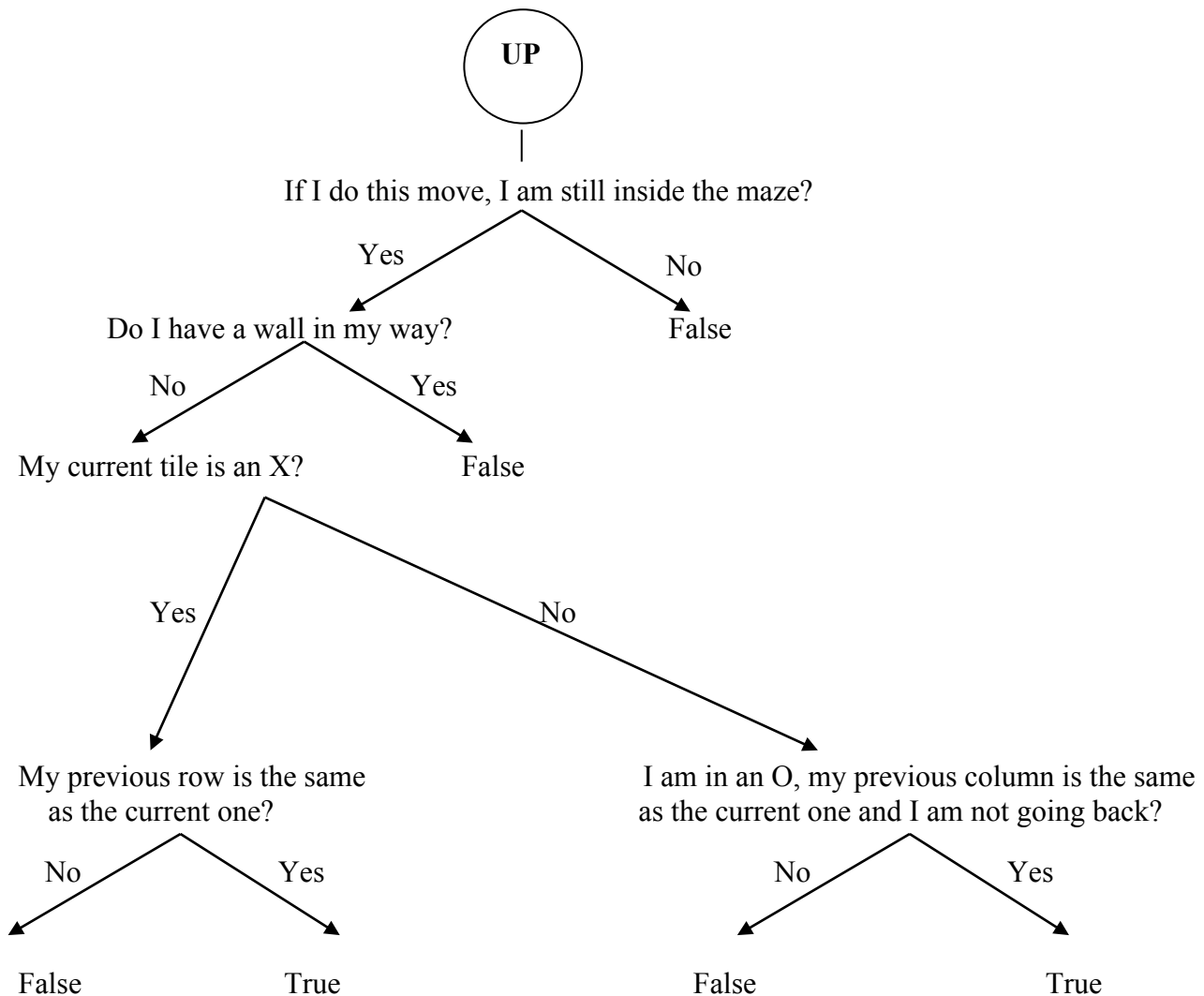
And finally, it is **known** because the system knows the outcomes of all actions.

2. Give the precise problem formulation needed for the problem to be solved by a search method.

- *States*: they are the position of the car in every moment.
- *Initial state*: the car is in some place out of the maze.
- *Goal test*: is the car in the flag?
- *Actions*: move if the type of the square allows me to move and there is not a wall blocking the movement.
- *Transition model*: the car has moved so we are in a different square.
- *Path cost*: the number of movements.

The flow diagram for the operator to show how the actions are done is the following:

It is done for the Up action because the others are the same.

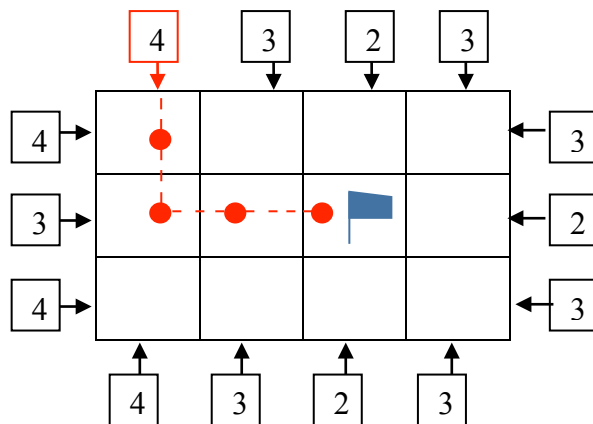


- Reason whether it is possible to apply the Relaxed Problem Technique to define an admissible and consistent heuristic. If so, apply this technique and give the resulting heuristic. Otherwise, explain whether you can define any other admissible and consistent heuristic.

The relaxed problem technique consists in removing some restrictions to the actions of the problem, we think that it is possible to apply this technique to our problem, and we are going to explain it.

The restrictions of our problem will be, that the tile of the parking has walls around, so there is only one possible entrance. On the other hand, every tile has its own movement, so in some tiles we can only go ahead and in others we can only turn. We have decided to remove those restrictions to the actions so we can allow the car to move without taking into account the type of the tile and the walls. However, we have to think that the diagonal movements are not allowed.

For example, in the following picture we can see the path without restrictions, realizing that the number of actions of the path coincides with a Manhattan Distance. So we have decided to apply that distance as our heuristic. Justifying that we can use the relaxed problem technique.



4. Explain which search method is best to solve this problem. In your explanation you should include a comparison of the method of your choice against others you discard.

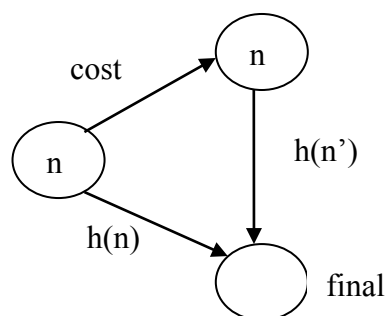
To solve this problem we have decided that the best method is to use an Informed search algorithm, in this case A* based on graph search.

We have chosen this method attending to these criteria:

- Informed search uses a heuristic, to help us getting the solution in a more optimal way. So we have discarded Blind search algorithms.
- Greedy best first is complete with graph search, so we discard this algorithm based on basic search. We should use this method but it is not optimal too so we discard it too.
- Finally, we have A* based on graph search, that is optimal if its heuristic is admissible and it takes into account the repeated states to find the best path.

An heuristic of a given node (n) is admissible when it is not bigger than the sum of the cost and the other heuristic of n', that is:

$$h(n) < \text{cost} + h(n')$$



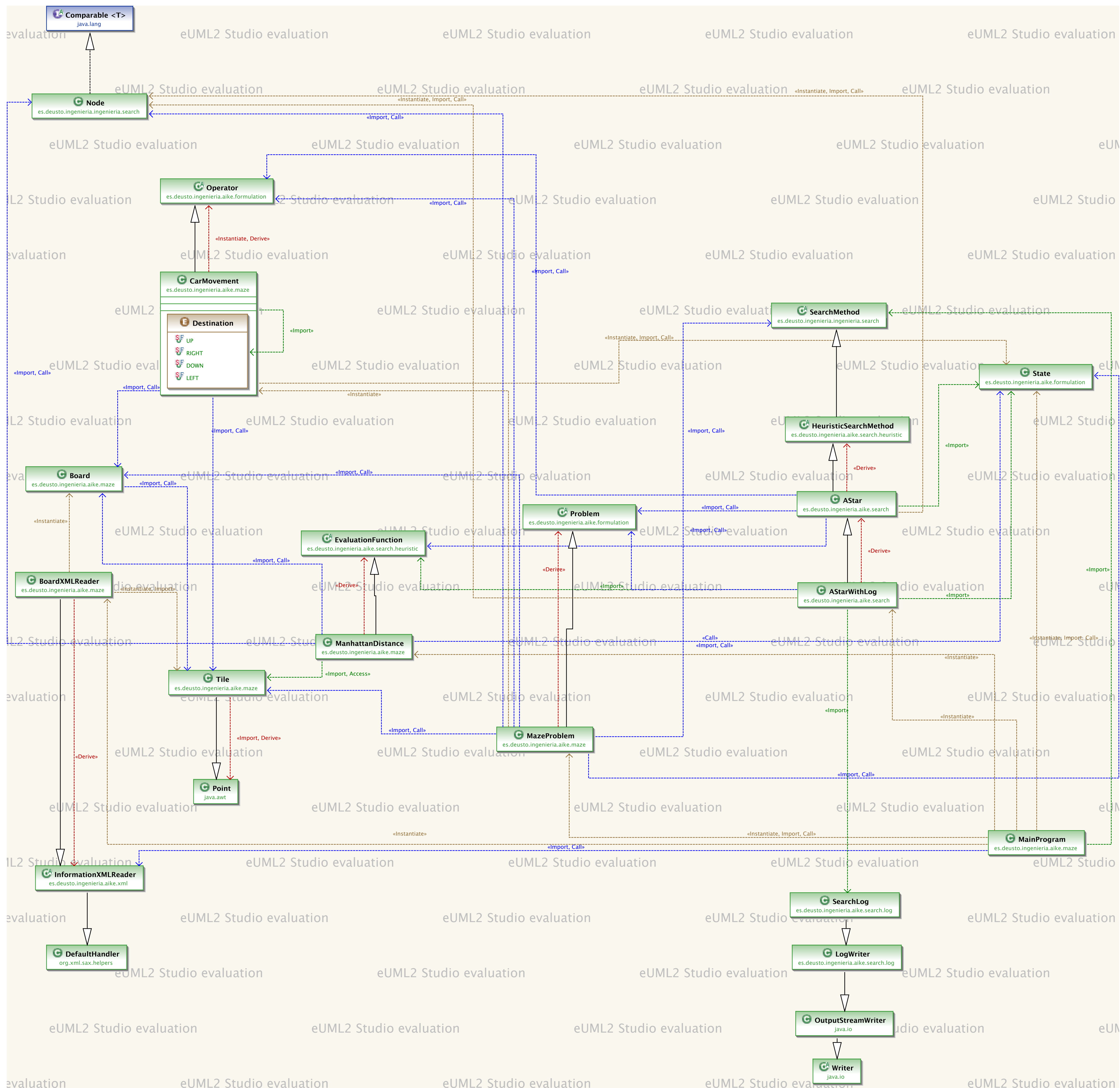
This algorithm is not provided with the templates so it has been decided to implement it in another new classes, using some classes of the templates as Node or State.

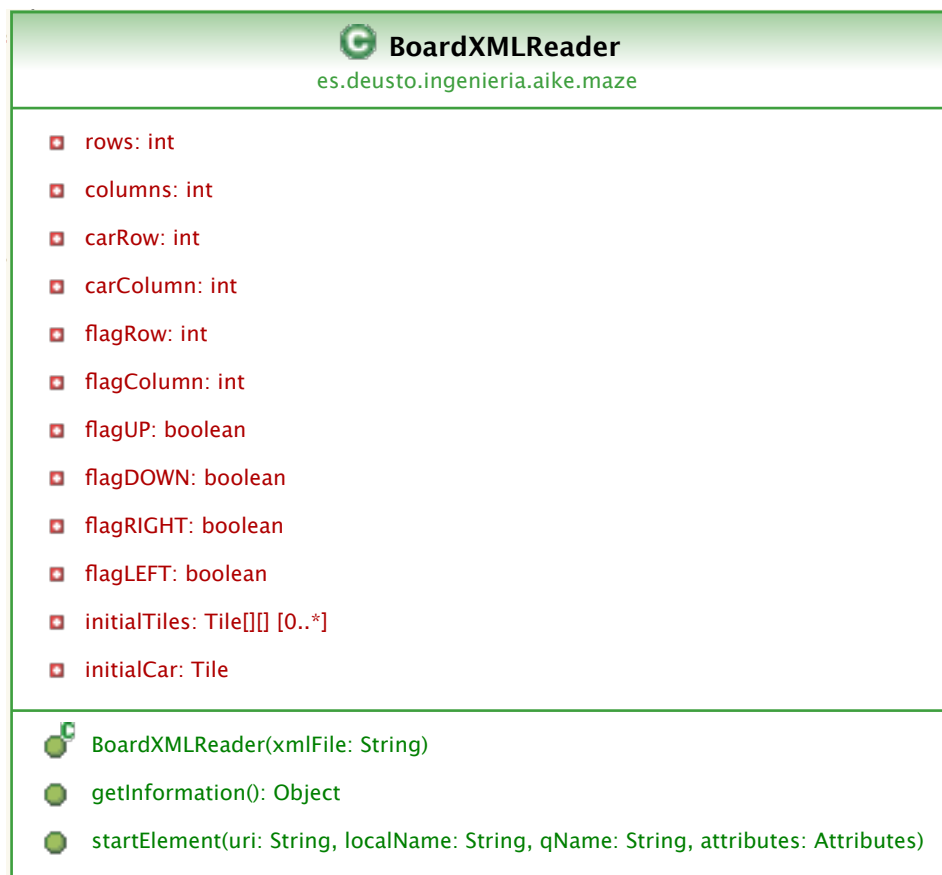
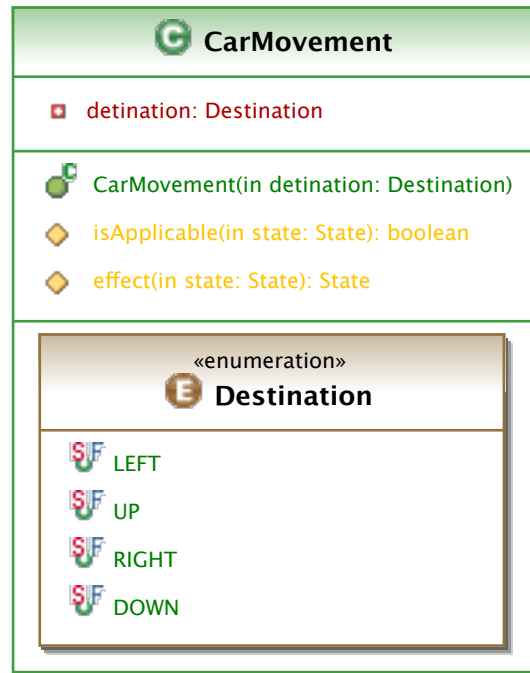
5. UML class diagram that include the objects from the templates FORMULATION and SEARCH discussed during the practical sessions.

The UML Diagram of the project has been divided into two sections just because the huge amount of relations between classes. This way, it will be easier for the reader of the document to understand the whole diagram.

The first one shows the relation between all the classes in the project (and templates) whilst the second one shows the structure of each new class generated for the implementation of the activity.

Both diagrams can be found in the following pages:





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G

MazeProblem

G

MazeProblem()

G

createOperators()

G

solve(searchMethod: SearchMethod)

G

isFinalState(state: State): boolean

G

MainProgram

S

main(args: String[])

G

Tile

SF

serialVersionUID: long

▣

type: String

▣

up_wall: boolean

▣

down_wall: boolean

▣

right_wall: boolean

▣

left_wall: boolean

G

Tile(type, row, column)

G

Tile(type, row, column, up_wall, down_wall, right_wall, left_wall)

G

Tile()

G

Tile(row, column)

G

setRow(row)

G

getRow(): int

G

setColumn(column)

G

getColumn(): int

G

getPosition(): Point

G

getType(): String

G

setType(type)

G

isUp_wall(): boolean

G

setUp_wall(up_wall)

G

isDown_wall(): boolean

G

setDown_wall(down_wall)

G

isRight_wall(): boolean

G

setRight_wall(right_wall)

G

isLeft_wall(): boolean

G

setLeft_wall(left_wall)

G

equals(obj): boolean

G

toString(): String

G

clone(): Object

Board

■

 tiles: Tile[][] [0..*]

■

 car: Tile

■

 previousTile: Tile

■

 flag: Tile

Board(tiles: Tile[][], car: Tile, flag: Tile)

getCar(): Tile

setCar(car: Tile)

getFlag(): Tile

setFlag(flag: Tile)

getPreviousTile(): Tile

setPreviousTile(previousTile: Tile)

setTiles(tiles: Tile[][])

getTiles(): Tile[][]

getTile(row: int, column: int): Tile

moveCar(tile: Tile)

equals(obj: Object): boolean

toString(): String

clone(): Object

ManhattanDistance

calculateG(in nodo: Node): double

calculateH(in nodo: Node): double

■

 distanciaManhattan(in ficha1: Tile, in ficha2: Tile): int

AStarWithLog

es.deusto.ingenieria.aike.search

AStarWithLog(function: EvaluationFunction)

search(problem: Problem, initialState: State): Node

AStar

es.deusto.ingenieria.aike.search

AStar(in function: EvaluationFunction)

search(in problem: Problem, in initialState: State): Node

◆

 expand(in node: Node, in problem: Problem, in generatedStates: List<State>, in expandedStates: List<State>): List<Node>

■

 updateChildren(in successorNode: Node, in frontier: List<Node>, in expandedNodes: List<Node>)