assignment1

October 25, 2017

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In []: # -*- coding: utf-8 -*-
        import heapq
        import numpy as np
        import re
        from copy import deepcopy
        from collections import Hashable
        class PuzzleNode(object):
            def __init__(self, size, state=[], parent=None,
                         heuristicCost=0, moves=0):
                ,, ,, ,,
                PuzzleNode class representing the state of tiles as a 2D array.
                Oparam size: Size of puzzle board
                Okeyword state: Starting state of board (default: [])
                Okeyword parent: Pointer to parent PuzzleNode (default: None)
                Okeyword heuristicCost: Estimated cost of going from current node to
                                        goal node (default: 0)
                Okeyword moves: Cost of going from initial node to current node
                                 (default: 0)
                Oivar board: 2D array representing state of tiles
                @ivar pathCost: Cost of going from initial node to current node, plus
                                 estimated cost of going from current node to goal node
                ,,,,,,
                self.board = np.array(state or [[0]*size for count in range(size)]) \
                    if type(state) != np.ndarray else state
                self.parent = parent
                self.pathCost = moves + heuristicCost
                self.moves = moves
                if not len(state):
                    self.__populateSelf()
            def __populateSelf(self):
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boardSize = len(self.board)
    cellNumber = 0
    for i in range(boardSize):
        for j in range(boardSize):
            self.board[i][j] = cellNumber
            cellNumber += 1
def __hash__(self):
   return int(''.join(
        str(_) for _ in self.board.flatten()
    ))
def __eq__(self, other):
    if self.__class__ != other.__class__:
        return False
    else:
        return np.array_equal(self.board, other.board)
def __lt__(self, other):
    if self.__class__ != other.__class__:
        return False
    else:
        return self.pathCost < other.pathCost</pre>
def __str__(self):
    11 11 11
    Prints 8-puzzle in following format:
    +----+
    1 - 1 - 1 - 1
    1 - 1 - 1 - 1
    1 - 1 - 1 - 1
    +----+
    Square puzzles of other sizes are printed analogously.
    Oreturn: Puzzle board formatted as described above.
    HHHH
    boardSize = len(self.board)
    maxCellSize = len(str(boardSize**2 - 1))
    # Fun thing I learned: I've separated the bits which use re.sub and
    # those which use str.replace because re.sub tends to be slower.
    # Also, just because I don't need it if I'm not actually using a
    # non-static expression.
    # More fun discussion here:
    # https://stackoverflow.com/questions/452104/is-it-worth-using-pythons-re-comp
    currentCellSize = 1
    cellsWithSpaces = str(self.board.tolist())[1:-1]
    for i in range(maxCellSize, 0, -1):
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digitRegex = re.compile(r'(\b\d{{{}}}\b)'.format(currentCellSize))
            cellsWithSpaces = digitRegex.sub(r'\1 |'.rjust(i+4),
                                             cellsWithSpaces)
            currentCellSize += 1
        withDashes = re.sub(r'b0b', '-', cellsWithSpaces)
        return '+' + '-' * (boardSize * (maxCellSize + 3) - 1) + '+\n' + \
            withDashes.replace('[', '|') \
                      .replace(']', '\n') \
                      .replace(', ', '') + \
            '+' + '-' * (boardSize * (maxCellSize + 3) - 1) + '+'
class InvalidBoardFormatError(TypeError):
    def __init__(self, board):
        self.message = "Board is not an instance of {} or {}: {}" \
            .format(list, np.ndarray, board)
class IncorrectBoardSizeError(ValueError):
    def init (self, board, size):
        self.message = 'Size \{0\} does not match dimensions of board \{1\}x\{1\}' \
            .format(size, board)
class InvalidTilesError(ValueError):
    def __init__(self, actualOccurrences, tile):
        self.message = 'Board contains {} occurrences of tile <{}> when it' \
                       ' should contain 1'.format(actualOccurrences, tile)
class UnsolvableBoardError(ValueError):
    def __init__(self, board):
        self.message = 'Board {} is unsolvable'.format(board)
def memoize(func):
    class memoized(dict):
        def __init__(self, func):
            self.func = func
        def __call__(self, arg):
            # Can we actually memoize this?
            if not isinstance(arg, Hashable):
                return self.func(arg)
            return self[arg]
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def __missing__(self, key):
            value = self[key] = self.func(key)
            return value
        def repr (self):
            return repr(self.func)
    return memoized(func)
def validatePuzzle(n, state, makeSolvable=False):
    Validates given n-puzzle.
    Oparam n: Size of puzzle board
    Oparam state: State of puzzle
    @keyword makeSolvable: If state is unsolvable and makeSolvable is True,
                           swaps two adjacent tiles such that the parity of the
                           board is reversed. If False or state is solvable,
                           nothing is swapped. (default: False)
    Oreturn err: Error code 0 if n and state are valid; else
                 -1 if IncorrectBoardSizeError, InvalidBoardFormatError, or
                 InvalidTilesError are raised; else
                 -2 if UnsolvableBoardError is raised
    Oreturn solvableBoard: If state is unsolvable and makeSolvable is True,
                           a solvable board. If either state is solvable or
                           makeSolvable is False, None.
    @raise IncorrectBoardSizeError: Raised if size of both state and all
                                    sublists of state doesn't equal n
    Oraise InvalidBoardFormatError: Raised if state is formatted incorrectly
    Oraise InvalidTilesError: Raised if state doesn't contain every number
                              from 0 to n^2-1 exactly once
    @raise UnsolvableBoardError: Raised if state cannot be solved to goal state
                                 (i.e. board parity is not even for odd n;
                                 board parity is not even for even n with 0 on
                                 odd row; or board parity is not odd for even
                                 n with O on even row, counting from top
                                 starting from 1 modified from Johnson &
                                 Story's (1879) use of the Manhattan distance
                                 to fit our different goal state).
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    err = 0
    solvableBoard = None
   try:
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if type(state) not in (np.ndarray, list):
    raise InvalidBoardFormatError(state)
if len(state) != n:
    raise IncorrectBoardSizeError(n, state)
for i in range(n):
    if len(state[i]) != n:
        raise IncorrectBoardSizeError(n, state)
expectedTile = 0
for j in range(n**2):
    actualOccurrences = sum(_.count(expectedTile) for _ in state)
    if actualOccurrences != 1:
        raise InvalidTilesError(actualOccurrences, expectedTile)
    expectedTile += 1
solvable = True
# Parity of board permutations
numTiles = 0
for k in range(n):
    for l in range(n):
        currentTile = state[k][1]
        if currentTile not in (0, 1):
            flattenedBoard = np.array(state).flatten().tolist()
            currentTileIdx = flattenedBoard.index(currentTile)
            numTiles += len(list(filter(
                lambda tile: tile < currentTile and tile != 0,</pre>
                flattenedBoard[currentTileIdx+1:]
            )))
permutationsParity = numTiles % 2
if n % 2 != 0:
    if permutationsParity != 0:
        solvable = False
else:
    # Parity of row with empty square
    emptyI, emptyJ = list(zip(*np.where(np.array(state) == 0)))[0]
    emptySquareRowParity = emptyI % 2
    if permutationsParity != emptySquareRowParity:
        solvable = False
if not solvable:
    if makeSolvable:
        # Unsolvable puzzles can be made solvable by swapping the first
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# two non-zero tiles.
                for x in range(1, n):
                    for y in range(1, n):
                        if state[x-1][y-1] != 0:
                            state[x][y], state[x-1][y-1] = \
                                 state[x-1][y-1], state[x][y]
                            return err, solvableBoard
            else:
                raise UnsolvableBoardError(state)
    except (
        InvalidBoardFormatError,
        IncorrectBoardSizeError,
        InvalidTilesError
    ):
        err = -1
    except UnsolvableBoardError:
        err = -2
    return err, solvableBoard
@memoize
def misplacedTiles(state):
    Heuristic function for number of misplaced tiles in given state.
    Oparam state: State of n-puzzle board
    Oreturn misplacedTiles: Number of misplaced tiles in state
    boardSize = len(state)
    goal = PuzzleNode(boardSize)
    misplacedTiles = 0
    for i in range(boardSize):
        for j in range(boardSize):
            if (state[i][j] != goal.board[i][j]):
                misplacedTiles += 1
    return misplacedTiles
@memoize
def manhattanDistance(state):
    Heuristic function for how far away each tile is from its goal position,
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Oparam state: State of puzzle board
    Creturn distance: Sum of how far away each tile is from its goal position
    boardSize = len(state)
    goalCoords = [(i, j) for i in range(boardSize) for j in range(boardSize)]
    goalCoordsDict = {tile: coords for tile, coords in enumerate(goalCoords)}
    distance = 0
    for x in range(boardSize):
        for y in range(boardSize):
            tile = state[x][y]
            if tile != 0:
                goalX, goalY = goalCoordsDict[tile]
                distance += abs(x - goalX) + abs(y - goalY)
    return distance
def getChildStates(state):
    Generates child states by looking at state to determine which tiles
    can move into the empty space, then returning all swaps of the empty
    space with those tiles.
    Oparam state: Current PuzzleNode
    Oreturn possibleChildren: Possible child states
    11 11 11
    # First find which tiles can move into empty space
    emptyI, emptyJ = list(zip(*np.where(state.board == 0)))[0]
    canMoveIntoEmpty = set()
    for i in (1, -1):
        if (emptyI + i < len(state.board)) and (emptyI + i >= 0):
            canMoveIntoEmpty.add((emptyI + i, emptyJ))
        if (emptyJ + i < len(state.board)) and (emptyJ + i >= 0):
            canMoveIntoEmpty.add((emptyI, emptyJ + i))
    # Swap tiles to generate each possible child state
    possibleChildren = []
    for possibleMove in canMoveIntoEmpty:
        i, j = possibleMove
        possibleChild = deepcopy(state.board)
        possibleChild[i][j], possibleChild[emptyI][emptyJ] = \
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using sum of individual Manhattan distances as the distance metric.

possibleChild[emptyI][emptyJ], possibleChild[i][j] possibleChildren.append(possibleChild)

return possibleChildren

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def solvePuzzle(n, state, heuristic, printed=True):
    n-puzzle solver using A*, modified from R. Shekhar's A* code from
    lesson 3.1.
    Oparam n: Size of puzzle board
    Oparam state: Starting state of puzzle
    Oparam heuristic: Heuristic function handler
    @keyword printed: If True, solvePuzzle prints all states of the board (as
                      lists of lists) from initial state to goal, as well as
                      the number of moves to reach the goal and the maximum
                      frontier size. If False, nothing is printed. (default:
                      True)
    Ovar frontier: A priority queue ordered by path cost
    Ovar explored: A dictionary of previously seen states mapped to the
                   PuzzleNode (and thus pathCost, depth) at which we saw them.
                   This prevents repeated states, including those which arise
                   from simply undoing the previous move.
    Oreturn total Moves: Number of moves required to reach goal state from
                        initial state
    Oreturn maxFrontierSize: Maximum size of frontier during the search
    @return err: -1 if n or state is invalid, -2 if board can't be solved,
                 else O. If err is non-zero, moves and frontierSize are O.
    11 11 11
   moves = 0
    totalMoves = 0
   maxFrontierSize = 0
    err, solvableBoard = validatePuzzle(n, state)
    if not err:
        state = solvableBoard or state
        initialNode = PuzzleNode(
            size=n,
            state=state,
            heuristicCost=heuristic(state)
        goalNode = PuzzleNode(size=n)
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explored = {}
frontier = [initialNode]
maxFrontierSize += 1
while frontier:
    currentNode = heapq.heappop(frontier)
    if currentNode == goalNode:
        totalMoves = currentNode.moves
        if printed:
            solutionPath = [str(currentNode.board.tolist())]
            while currentNode.parent:
                solutionPath.append(
                    str(currentNode.parent.board.tolist())
                currentNode = currentNode.parent
            print('\n'.join(solutionPath[::-1]))
            print(totalMoves)
            print(maxFrontierSize)
        return total Moves, maxFrontierSize, err
    explored[currentNode] = currentNode
    childStates = getChildStates(currentNode)
    for childState in childStates:
        childNode = PuzzleNode(
            size=n,
            state=childState,
            parent=currentNode,
            heuristicCost=heuristic(childState),
            moves=currentNode.moves + 1
        )
        exploredNode = explored.get(childNode)
        # If we: a) haven't yet explored this node, or
        # b) explored this node when it had a higher pathCost,
        # then we should put it into the frontier so that
        # we can explore it.
        if (
            exploredNode is None or
            childNode.pathCost < exploredNode.pathCost</pre>
        ):
            heapq.heappush(frontier, childNode)
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# If we're already planning to explore this node later on
                # (i.e. it's in the frontier), AND the version of the node
                # we're currently looking at has a lower pathCost than the
                # one in our frontier, then we should replace the node in
                # our frontier with the one we're currently looking at.
                elif (
                    exploredNode is not None and
                    exploredNode in frontier and
                    childNode.pathCost < exploredNode.pathCost</pre>
                ):
                    frontier.remove(exploredNode)
                    heapq.heappush(frontier, childNode)
            moves += 1
            if len(frontier) > maxFrontierSize:
                maxFrontierSize = len(frontier)
    return total Moves, maxFrontierSize, err
def testSolvePuzzle(boards, heuristics):
    Tests solvePuzzle, printing out number of moves and maximum frontier size
    for each given initial condition.
    Oparam boards: Scrambled boards to use as initial states in tests
    Oparam heuristics: Available heuristics for A* algorithm
    for board in boards:
        for heuristic in heuristics:
            n = len(board)
            print('Current board: {}'.format(board))
            print('Current heuristic: {}'.format(heuristic))
            moves, frontierSize, err = solvePuzzle(n=n,
                                                    state=board,
                                                    heuristic=heuristic,
                                                    printed=False)
            print('Moves to solve current board: {}'.format(moves))
            print('Current board\'s max frontier size: {}\n'.format(
                  frontierSize))
heuristics = [misplacedTiles, manhattanDistance]
testBoards = [
    [[5, 7, 6], [2, 4, 3], [8, 1, 0]],
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[[7, 0, 8], [4, 6, 1], [5, 3, 2]],

[[2, 3, 7], [1, 8, 0], [6, 5, 4]]

]

testSolvePuzzle(testBoards, heuristics)
```

initial state	heuristic	number of moves	max frontier size
[[5, 7, 6], [2, 4, 3], [8, 1, 0]]	Misplaced tiles	28	28547
	Manhattan distance	28	937
[[7, 0, 8], [4, 6, 1], [5, 3, 2]]	Misplaced tiles	25	15444
	Manhattan distance	25	983
[[2, 3, 7], [1, 8, 0], [6, 5, 4]]	Misplaced tiles	17	452
	Manhattan distance	17	68

From the above table, it's clear that the Manhattan distance heuristic far outperforms the misplaced tiles heuristic in terms of the number of possible positions it generates (i.e., Manhattan distance usually has a frontier size smaller than misplaced tiles, by 1 order of magnitude).