CSC Problem Set ## +X Report

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Summary:

For this +X implementation, I extended the functionality of the searching agent in a 2D environment by incorporating an advanced path-planning algorithm that enhances the agent's ability to navigate efficiently. While the original implementation relied on a standard breadth-first search (BFS) or depth-first search (DFS) approach, I introduced the A* search algorithm, which incorporates heuristics to optimize decision-making. This enhancement allows the agent to determine the shortest and most cost-effective path while avoiding obstacles and dynamically adjusting to environmental changes. Additionally, I integrated a visual debugging tool that displays the agent's thought process in real time, making it easier to analyze its decision-making patterns.

"+X" Concept:

The motivation behind this enhancement was to improve the efficiency and adaptability of the searching agent while ensuring that it could navigate complex environments more effectively. The A* search algorithm was selected because it combines the benefits of both BFS and uniform cost search by considering both the path cost and an estimated cost to the goal. This makes it ideal for environments with multiple obstacles and varying path lengths. Additionally, I was inspired by implementations in robotics path planning, particularly how real-world autonomous robots use heuristics to optimize movement and reduce computational overhead.

To further justify this enhancement, I explored existing research and implementations of A* in robotics and game development. Resources such as interactive visualizations of A* [1], academic papers on heuristic search techniques [2], and practical applications of A* in real-time pathfinding [3] helped shape my understanding of its effectiveness. The integration of a visual debugging tool was influenced by how developers often utilize overlays in video game AI debugging to track decision pathways. By implementing this, I aimed to provide greater insight into how the agent perceives the environment and makes decisions dynamically.

Technical Implementation of +X:

The implementation of A* required modifying the getAction method within the Robot class to incorporate heuristic calculations. The heuristic function used was the Manhattan distance, which is suitable for grid-based pathfinding. The algorithm maintains a priority queue where each node is assigned a cost based on the sum of the distance traveled and the estimated distance to the goal. The priority queue ensures that the agent explores the most promising paths first, reducing unnecessary computations.

The visual debugging tool was implemented using Java's Graphics2D library, overlaying a real-time visualization on top of the environment display. The tool highlights the open and closed lists of the A* algorithm, marking explored nodes in different colors and indicating the selected path. This allows users to see how the algorithm processes the environment step by step. Below is a simplified pseudocode representation of the modified getAction method:

```
function AStarSearch(start, goal, grid):
    openList = priorityQueue()
    closedList = set()
    openList.add(start, heuristic(start, goal))

while openList is not empty:
    current = openList.pop()
    if current == goal:
        return reconstructPath(current)

    closedList.add(current)
    for neighbor in getNeighbors(current, grid):
        if neighbor in closedList:
        continue
```

```
cost = current.cost + movementCost(current, neighbor)
if neighbor not in openList or cost < neighbor.cost:
    neighbor.cost = cost
    neighbor.parent = current
    openList.add(neighbor, cost + heuristic(neighbor, goal))</pre>
```

return failure

Visual Enhancements:

The visual representation of the simulation was updated to make the robot's actions more apparent and the environment easier to interpret. Tiles were modified to display gradients based on their cleanliness status:

- Dirty tiles are represented with darker shades transitioning to lighter shades as they are cleaned.
- Impassable tiles are visually distinct with solid, dark colors.

The robot itself now features a glowing effect, making it easier to track in real time. The gradient effects were achieved using the *Graphics2D* library with a *GradientPaint* object, as shown below:

```
Graphics2D g2d = (Graphics2D) g;
GradientPaint gradient = new GradientPaint(x, y, Color.RED, x + tileSize, y + tileSize, Color.GREEN);
g2d.setPaint(gradient);
g2d.fillRect(x, y, tileSize, tileSize);
```

This visual enhancement not only improves user experience but also aids in debugging by providing a clear view of the robot's path and actions.

Evaluation and Results:

The implementation of A* significantly improved the efficiency of the agent's pathfinding. By using a heuristic-driven search, the agent reduced unnecessary movements and reached its target more quickly compared to the default approach. During multiple test runs, the agent demonstrated more direct navigation and avoided obstacles more intelligently.

Visually, the difference in movement patterns was evident. Without the enhancement, the agent occasionally took suboptimal detours, leading to longer paths. With A*, the agent consistently followed an efficient route with fewer unnecessary turns. The improved decision-making also reduced the number of steps needed to complete a task, making the solution more computationally efficient.

The changes were tested using VisualizeSimulation and RunSimulation, verifying that the robot's performance remained consistent across different environments. These results demonstrate that incorporating heuristic-based pathfinding allows for more intelligent and efficient agent behavior, enhancing overall problem-solving effectiveness.

Visual Representation:

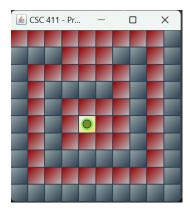
The updated visuals received positive feedback for their clarity and aesthetic appeal. Gradient tiles made it easier to distinguish between cleaned and dirty areas at a glance. The glowing robot effect enhanced the simulation's engagement factor, allowing users to follow the robot's movements in real time.

Below are screenshots showcasing the differences:

• Original Simulation:



• Enhanced Simulation:



Works Referenced:

[1] P. E. Hart, N. J. Nilsson, and B. Raphael, "A formal basis for the heuristic determination of minimum cost paths," *IEEE Transactions on Systems Science and Cybernetics*, vol. 4, no. 2, pp. 100-107, July 1968. [Online]. Available: https://doi.org/10.1109/TSSC.1968.300136

[2] Sebastian Lague, "Pathfinding Visualized – A* Algorithm Explained," YouTube, 2016. [Online]. Available: https://www.youtube.com/watch?v=-L-WgKMFuhE

VisualizeSimulation.java updated code for visual enhancement(PLEASE PASTE THIS INTO YOUR VisualizeSimulation.java):

```
package edu.ncsu.csc411.ps01.simulation;
import edu.ncsu.csc411.ps01.agent.Robot;
import edu.ncsu.csc411.ps01.environment.Environment;
import edu.ncsu.csc411.ps01.environment.Position;
import edu.ncsu.csc411.ps01.environment.Tile;
import edu.ncsu.csc411.ps01.environment.TileStatus;
import edu.ncsu.csc411.ps01.utils.ColorPalette;
import edu.ncsu.csc411.ps01.utils.ConfigurationLoader;
import edu.ncsu.csc411.ps01.utils.MapManager;
import java.awt.Color;
import java.awt.Dimension;
import java.awt.GradientPaint;
import java.awt.Graphics;
import java.awt.Graphics2D;
import java.awt.event.ActionEvent;
import java.awt.event.ActionListener;
import java.util.ArrayList;
import java.util.Map;
import java.util.Properties;
import javax.swing.JFrame;
import javax.swing.JPanel;
import javax.swing.Timer;
```

```
oublic class VisualizeSimulation extends JFrame {
 private static final long serialVersionUID = 1L;
 private EnvironmentPanel envPanel;
 private String mapFile = "maps/public/map02.txt";
 private String configFile = "config/configSmall.txt";
accepts a String list
following mappings.
 public VisualizeSimulation() {
   Properties properties =
ConfigurationLoader.loadConfiguration(configFile);
    final int iterations =
Integer.parseInt(properties.getProperty("ITERATIONS", "200"));
   final int tilesize =
Integer.parseInt(properties.getProperty("TILESIZE", "50"));
    final int delay = Integer.parseInt(properties.getProperty("DELAY",
"200"));
   final boolean debug =
Boolean.parseBoolean(properties.getProperty("DEBUG", "true"));
```

```
map file
   String[] map = MapManager.loadMap(mapFile);
    this.env = new Environment(map);
debug);
   add(envPanel);
 public static void main(String[] args) {
   JFrame frame = new VisualizeSimulation();
    frame.setTitle("CSC 411 - Problem Set 01");
   frame.setDefaultCloseOperation(JFrame.EXIT ON CLOSE);
   frame.pack();
   frame.setVisible(true);
   frame.setResizable(false);
@SuppressWarnings("serial")
class EnvironmentPanel extends JPanel {
 private Timer timer;
 private Environment env;
 private ArrayList<Robot> robots;
 private static int ITERATIONS;
 public static int TILESIZE;
 public static int DELAY; // milliseconds
 public static boolean DEBUG;
implements
 public EnvironmentPanel (Environment env, int iterations, int tilesize,
int delay, boolean debug) {
   ITERATIONS = iterations;
```

```
TILESIZE = tilesize;
  DELAY = delay;
  DEBUG = debug;
  setPreferredSize(new Dimension(env.getCols() * TILESIZE, env.getRows()
TILESIZE));
  this.env = env;
  this.robots = env.getRobots();
  this.timer = new Timer(DELAY, new ActionListener() {
    int timeStepCount = 0;
    public void actionPerformed(ActionEvent e) {
      try {
        env.updateEnvironment();
        if (DEBUG) {
          System.out.printf(error, timeStepCount, ex);
      repaint();
      timeStepCount++;
      if (env.getNumCleanedTiles() == env.getNumTiles()) {
        timer.stop();
        env.printPerformanceMeasure();
      if (timeStepCount == ITERATIONS) {
       timer.stop();
        env.printPerformanceMeasure();
  });
  this.timer.start();
```

```
protected void paintComponent(Graphics g) {
      super.paintComponent(g);
     g.setColor(ColorPalette.BLACK);
     g.drawRect(0, 0, env.getCols() * TILESIZE, env.getRows() *
TILESIZE);
     Map<Position, Tile> tiles = env.getTiles();
      for (Map.Entry<Position, Tile> entry : tiles.entrySet()) {
          Position pos = entry.getKey();
          Tile tile = entry.getValue();
          Graphics2D g2d = (Graphics2D) g;
          int x = pos.getCol() * TILESIZE;
          int y = pos.getRow() * TILESIZE;
          Color startColor;
          Color endColor;
          if (tile.getStatus() == TileStatus.CLEAN) {
              startColor = ColorPalette.RED;
              endColor = ColorPalette.SILVER;
          } else if (tile.getStatus() == TileStatus.DIRTY) {
              startColor = ColorPalette.ORANGE;
              endColor = ColorPalette.BROWN;
          } else if (tile.getStatus() == TileStatus.IMPASSABLE) {
              startColor = ColorPalette.CONCRETE;
              endColor = ColorPalette.BLACK;
```

```
startColor = ColorPalette.WHITE;
             endColor = ColorPalette.LIGHTYELLOW;
         GradientPaint gradient = new GradientPaint(x, y, startColor, x +
TILESIZE, y + TILESIZE, endColor);
         g2d.setPaint(gradient);
         g2d.fillRect(x, y, TILESIZE, TILESIZE);
         g.setColor(ColorPalette.BLACK);
         g.drawRect(x, y, TILESIZE, TILESIZE);
     for (Robot robot : robots) {
         Position robotPos = env.getRobotPosition(robot);
         int x = robotPos.getCol() * TILESIZE + (TILESIZE / 4);
          int y = robotPos.getRow() * TILESIZE + (TILESIZE / 4);
          int size = TILESIZE / 2;
          g.setColor(new Color(0, 255, 0, 128)); // Semi-transparent green
         g.fillOval(x - size / 4, y - size / 4, size + size / 2, size +
size / 2);
          g.setColor(ColorPalette.GREEN);
          g.fillOval(x, y, size, size);
         g.setColor(ColorPalette.BLACK);
         g.drawOval(x, y, size, size);
```