

Improved Dijkstra Algorithm for Mobile Robot Path Planning and Obstacle Avoidance

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Abstract: Optimal path planning avoiding obstacles is among the most attractive applications of mobile robots (MRs) in both research and education. In this paper, an optimal collision-free algorithm is designed and implemented practically based on an improved Dijkstra algorithm. To achieve this research objectives, first, the MR obstacle-free environment is modeled as a diagraph including nodes, edges and weights. Second, Dijkstra algorithm is used offline to generate the shortest path driving the MR from a starting point to a target point. During its movement, the robot should follow the previously obtained path and stop at each node to test if there is an obstacle between the current node and the immediately following node. For this aim, the MR was equipped with an ultrasonic sensor used as obstacle detector. If an obstacle is found, the MR updates its diagraph by excluding the corresponding node. Then, Dijkstra algorithm runs on the modified diagraph. This procedure is repeated until reaching the target point. To verify the efficiency of the proposed approach, a simulation was carried out on a hand-made MR and an environment including 9 nodes, 19 edges and 2 obstacles. The obtained optimal path avoiding obstacles has been transferred into motion control and implemented practically using line tracking sensors. This study has shown that the improved Dijkstra algorithm can efficiently solve optimal path planning in environments including obstacles and that STEAM-based MRs are efficient cost-effective tools to practically implement the designed algorithm.

Keywords: Mobile robot (MR); STEAM; path planning; obstacle avoidance; improved dijkstra algorithm

1 Introduction

In the current 21st century, robots are occupying a key place in several domains including industry, agriculture, healthcare, construction, logistics and education... In education, extensive works



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have focused on implementing Science, Technology, Engineering, Art and Mathematics (STEAM) approaches around mobile robots (MRs). With the integration of modern technologies such as 3D printing, open-source programming environments, artificial intelligence (AI) and Internet of Things (IoT) in the fourth industrial revolution 4.0 context, MRs are playing constructive roles. Due to their capability of locomotion and ability to interact with their close environments, MRs (controlled or autonomous) can be used in education and research. MRs are inherently multidisciplinary systems involving several disciplines ranging from material engineering to cognitive science and computer vision. Concepts like kinematics, 3D printing, mechanical design and material science belong to the Mechanical Engineering (ME). In addition, an MR should include motors, sensors, controllers, power electronics and batteries which are concepts related to Electrical Engineering (EE). Computer Engineering (CE) and Computer Science (CS) may be present in a MR through programming and more attractively through modern applications such as path planning, computer vision, wireless sensor networks, communication, artificial intelligence. Purchasing “professional” mobile robotics platforms is unfortunately expensive for a relatively limited use (as low as 1%) as reported by [1]. To overcome this difficulty, many researchers resorted to design their own DIY (Do-It-Yourself) platforms under various technical (should be based on open resources), educational (can be used to implement various educational activities), cost (should be affordable) and efficiency (should perform several tasks) considerations.

In this paper, the progressive steps of a STEAM framework around an MR are detailed. A particular attention is allocated to the components in order to make it as easy as possible to be implemented by students at different levels. In addition, the problem of path planning and obstacle avoidance is addressed from its algorithmic and motion control perspectives. The flow-chart of the work conducted in this study is presented in [Fig. 1](#).

The main contributions of the current study can be highlighted as follows:

- Implement a STEAM framework around an MR in a simple and detailed manner in order to make it easy to be conducted from scratch by students, educators and researchers.
- Show the multi-disciplinarity of MRs and how they can be used to assess skills in different disciplines.
- Address the problem of path planning and obstacle avoidance of the DIY MR using an efficient and computationally low-cost algorithm (an improved Dijkstra algorithm).
- Show how the shortest path solution derived from the algorithm is transferred into control motion and implemented practically.

The remainder of this paper is described as follows. In Section 2, a relatively detailed state-of-the-art of previously developed mobile robotics platforms is presented as well as selected researches about path planning. Section 3 will include the robot design and the steps of the STEAM framework. In Section 4, a path planning and obstacle avoidance application of the designed mobile robot will be presented. Section 5 concludes the paper.

2 Related Works

During the last few years, many researches about mobile robotics platforms have been conducted from STEAM perspective. These researches were different according to several factors. In this section, selected platforms will be described according to many features including the year of implementation, the mechanical design tool, the cost, the used sensors, the programming tool, the application and the achieved skills. A special focus will be allocated to the path planning and its relevant components.

Mona robotics platform has been developed in The University of Manchester, UK [1] in 2019. This platform is mainly based on open-source programmable devices, low-cost and including several sensors. Among the advantages of this platform is that it allows robots communication which strengthens the concept of swarm of robots largely used in research and education. The concept of swarm (group of interacting) mobile robots has been also addressed in [2] where the authors have developed an interface allowing full integration of robots in the ROS (Robotics Operating System). Through the novel driver interface, mobile robotics practitioners may focus more on their tasks which can include data manipulation, robots' communication and sensors deployment. For teaching some concepts of science (the pulsimeter), robotic and coding (programming) tools have been used in a middle school in Turkey (2018-2019) [3]. Via a structured approach, students have developed their own experiments using the open-access Arduino platform. Small printable robots (called "printbots") have been presented in [4] as strong educational tools for studying mobile robots and manipulators. The usefulness of these robots took benefit from the availability of 3D printers. Several courses have been designed (practice and simulation) over 8 years in a Spanish institute. The concept of deep learning (as extension of machine learning) is the trend of nowadays research in computer science and engineering. A real need of experimental platforms to easily conduct deep learning projects has motivated the authors in [5] to implement an affordable robotics platform around the well-known Arduino board. The main advantage of this platform is that it includes a mobile phone for capturing photos and videos which can be then easily handled in several integrated development environments (IDEs) such as Python, Matlab and C/C++.

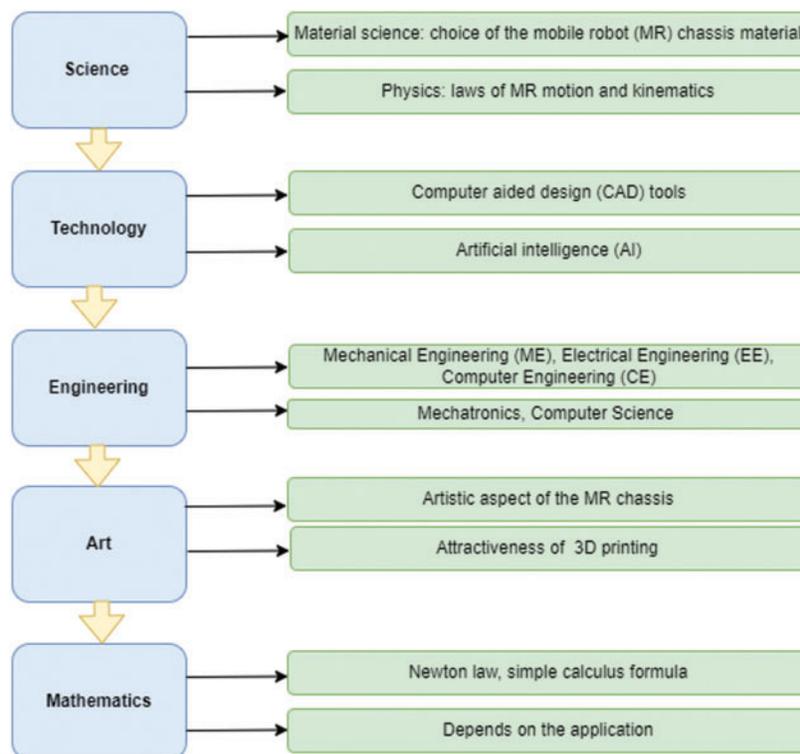


Figure 1: Flow-chart of a STEAM framework based on a mobile robot (MR)

When participating in a European robotics contest, the team of [6] has faced technical issues in their robots, 3 days before the beginning. Thus, they had resort to open hardware and software resources as well as to smartphones and cloud services to carry out “fast” prototypes which have later achieved 83% of required tasks of the competition. In 2015, a Spanish team has designed a low-cost (around 35 Euro) robotics platform using Arduino and Android mobile operating system to be used in various courses covering information technology, communication and engineering [7]. Similar to the work presented in [4], the authors in [8] have designed low-cost robots using the 3D printing technology and have conducted 3 educational experiments with students. Several concepts including design, sensing and programming have been successfully taught. During the European robotics week (2018), the work developed in a workshop for teaching STEM concepts has been described in detail in [9]. Although the participants had no previous experience in programming and mathematics, the main objectives of the workshop have been reported (by the authors) to be achieved. Developing countries such as Mexico are missing practical and hands-on tools to teach robotics in elementary and university schools which may result on decrease in motivation. To overcome this difficulty, the authors in [10] have used a hexapod robot to assess the programming skills. The achieved results were reported to be excellent. In the same direction, the authors in [11] have built a robotics platform only for 25 USD (2018) to implement a line-follower application where several skills related to sensors, Bluetooth communication and microcontrollers’ programming have been improved in a STEM framework. More successful STEM implementations using robotics platforms can be found in [12–27]. Although these studies are distributed over different periods and were designed using different tools and resources, they all led to results promoting the capability of thinking as well as improving 21st century skills.

Path planning of mobile robots includes the algorithms and the resulting control motion strategies that allows the MR to move in its close environment following the shortest path and avoiding collision [28]. In their survey paper [29], the authors have provided a summary of the algorithms used for path planning. They reported that Dijkstra’s algorithm remains among the best ones. The concept of neural dynamics has been used in [30] for solving the path planning problem of an autonomous robot moving in an unknown dynamic environment. From a control perspective, [31] has developed a switching control law for obstacle avoidance of a MR. The practical implementation has been shown to be without heavy computational effort nor a high memory requirement. A hardware platform and a software algorithm have been developed in [32] for a car system. Dijkstra’s algorithm has been successfully applied for path planning and obstacle avoidance in a generated unknown indoor environment. Offline path planning algorithms have been used to generate data for training a neural network algorithm used online for path planning [33]. Although the results were reported to be promising, the time-consuming character of neural networks were considered as a limitation. Other researches have tackled the path planning of MRs using Q-learning algorithm [34] and a combination of Dijkstra and A* algorithms [35]. As compared to the state-of-the-art algorithms for path planning, the improved Dijkstra algorithm has shown its superiority at least in three levels. Firstly, Dijkstra algorithm is simple in its structure because it works on any environment as modeled by the designer from the beginning without adding any nodes like in [28] where extra nodes are included. In fact, adding nodes may induce extra computational burden. Secondly, the Dijkstra algorithm has the ability to find the optimal path from a starting node to any other node in the graph including the target node defined by the user. Its computational complexity is known to be $O(n^2)$ where n is the number of nodes in the diagraph. This complexity is stated in [29] to be better than the complexity of A* algorithm for example. Finally, our proposed improved Dijkstra algorithm is shown through this study to operate in a dynamic environment in which even moving obstacles can be detected and the MR can react to update its path accordingly without any additional computation time.

From the previous literature review, it can be noticed that:

- Mobile robots (MRs) are efficient systems around which STEAM framework can be implemented for achieving several skills in education and conducting successful researches.
- Path planning from both algorithmic and hardware perspectives can be practically implemented using the STEAM designed MR.
- Dijkstra's algorithm remains among the best algorithms for MR path planning.

3 STEAM Methodology Implementation

In this work, the progressive steps of designing and manufacturing a low-cost robotics platform to be used in education and research are detailed. The process follows a science, technology, engineering, art and mathematics (STEAM) approach. At each step, the component related to STEAM is clarified. The designed platform use free access resources in order to reduce the cost and make the design affordable by students and researchers particularly in developing countries where resources are relatively limited.

3.1 Mobile Robot Chassis Design

As an immediate consequence of the emergence of 3D printers, students and researchers can take benefit from this new technology to develop their own robotics platforms. However, the software side remains challenging because commercial (professional) software is still relatively expensive. As opposite to several platforms which require paid licenses, we chose a relatively new platform, Tinkercad (a product from Autodesk Company), known to be online and free¹. This platform includes several tools that can help in 3D design, electric/electronic circuits, code-block programming and many free courses and tutorials. Moreover, this platform allows the user to create his personal account and keep all his previous designs and circuits in cloud storage. In this work, Tinkercad is used for two tasks: the mechanical design and the programming of the microcontroller during the simulation step.

In Tinkercad, the mechanical design is based on the principle of constructive solid geometry (CSG) where basic shapes and on adding/removing other shapes to construct complex forms are used. A screenshot of a simple mobile robotics platform customized for conducting this work is shown in Fig. 2.

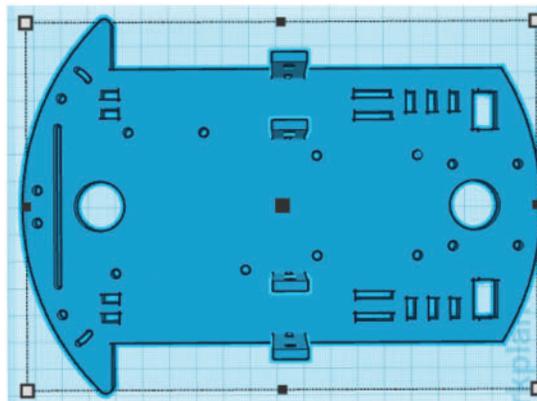


Figure 2: Screenshot of Tinkercad-based design of the mobile robot

¹www.tinkercad.com

3.2 Hardware Specification

In this study, the MR platform has been designed to be functional, low-cost, and extendable to other applications [27]. The following parts have been assembled to construct the MR:

- Wheels: two standard wheels in addition to a castor wheel in order to provide to the MR stability and sufficient traction [24].
- Motors: two direct current (DC) motors to operate the MR.
- Motors' driver: to control the MR motion.
- Microcontroller: to generate the control signals applied to the driver as well as to receive the sensors' signals.
- Battery: to provide the energy to the MR.
- Sensors: to serve as perception tools for the MR close environment.

The material used to construct the MR can be plastic, MDF (medium density fiber), acrylic or metal frame. In order to reduce the cost, in this study, the choice was a chassis fabricated using plastic material.

To select the driven motors and the battery pack capacity, an online motor sizing tool was used². For a mobile robot, direct current (DC) motors are commonly used. However, those motors have the drawback of turning very fast without generating sufficient torque. The solution usually adopted consists of reducing the speed (Ω) by using a gear-down in order to produce enough torque (T) and keep the power (P) constant as in Eq. (1).

$$P = T \times \Omega \quad (1)$$

Since the MR will move in a flat surface, the torque will be calculated following Eq. (2) below.

$$T = \frac{\text{Mass} \times \text{WheelRadius}}{\text{Number of wheels}} \times \text{efficiency} \times \text{safety factor} \quad (2)$$

The safety factor is used here to allow the battery pack to power the MR embedded electronics such as the microcontroller and the sensors. Moreover, some imperfections in the surface will be overcome by the MR. The results provided by the sizing motor tool are summarized in Tab. 1. Based on the results of Tab. 1, the corresponding selected MR system parts as well as their technical specifications are provided in Tab. 2. In order to implement the path planning applications, an ultrasonic sensor and two infrared line following sensors have been also included to the MR platform.

Table 1: DC motors and battery pack capacity sizing

Inputs	Values
Total mass (Kg)	2
Number of drive motors (#)	2
Radius of drive wheel (m)	0.03
Robot velocity (m/s)	2
Maximum incline angle (deg.)	10
Supply voltage (V)	12

(Continued)

²<https://www.robotshop.com/community/blog/show/drive-motor-sizing-tool>

Table 1: Continued

Inputs	Values
Desired acceleration (m/s ²)	1
Desired operating time (min)	5
Total efficiency (%)	65
Outputs	Values
Angular velocity (rad/s)	66.667
Torque (g.cm)	1271.9
Total power (W)	8.3184
Maximum current (A)	0.6932
Battery pack capacity (mAh)	1155

Table 2: MR parts specifications

Part	Specifications	Cost (USD)
DC motor with wheel	Number of wheeled motors: 2 Operating voltage: 3-6 V Torque: 800 g.cm Load current: 170 mA A castor wheel for MR equilibrium	10
L298 DC motor driver	Operating voltage: up to 46 V Total DC current: up to 4 A Number of driven motors: 4	10
Battery pack	Number of batteries: 2 Capacity: 2600 mAh/battery Operating voltage: 3.7 V	10
ARDUINO UNO R3 Microcontroller	Series: AVR-Atmega Type: MCU 8-bit Number of digital pins: 14 Number of analog inputs: 6 Clock frequency: 16 Mhz Number of analog inputs: 6	15
Distance Ultrasonic sensor HC-SR04	Number of sensors: 1 Accuracy: 2 cm Detection range: 2 cm – 450 cm	15
Line follower sensor	Number: 2 Ref: KY-033 Principle: infrared transceiver Power: 5 V	15

(Continued)

Table 2: Continued

Part	Specifications	Cost (USD)
Other additional parts	Jumper wires ON/OFF switch Small breadboard	5
Total cost		80

DC Motor Driver

The L298N module has the following pinouts:

- Two terminal screwed power pins for each one of the two motors. These pins provide the voltage for driving the motors. The motors used in the MR require a voltage between 7 and 12 V.
- One terminal screwed pin for the Ground.
- A 5 V pin used for control signals' generation.
- Two Enable pins (ENA and ENB) used in general for controlling the motors speeds. This will be done through a jumper. If the jumper exists, the motors work at their maximum speeds and if it is removed, the Pulse Width Modulation (PWM) input is activated and thus the motors speeds can be controlled.
- The pins IN1, IN2, IN3 and IN4 are used for controlling the motors directions and for stopping them. Note here that (IN1, IN2) are used for one motor and (IN3, IN4) are used for the other motor.

Since the robot can move Reverse, Forward, Left, Right and Stop, the corresponding combinations of the IN1, IN2, IN3 and IN4 states are summarized in [Tab. 3](#).

Table 3: The L298 driver logic controller

IN1	IN2	IN3	IN4	Direction
0	0	0	0	Stop
1	0	1	0	Forward
0	1	0	1	Reverse
1	0	0	1	Left
0	1	1	0	Right

Ultrasonic Sensor

The ultrasonic sensor (HC-SR04) is an electronic device able to measure the distance to the first object existing in front of it. It is based on the same principle of radar and sonar. In fact, the sensor source generates an ultra-high frequency signal which is reflected after hitting the obstacle. The reflected echo is sensed by the sensor receiver. The microcontroller measures the duration between sending the signal and receiving the echo and then calculates the distance using the formula:

$$\text{distance} = \frac{\text{speed} \times \text{duration}}{2} \quad (3)$$

The speed of the ultrasonic wave is known to be 343 m/s.

Line Follower Sensor

Line follower sensor (KY-033) is a sensor that gives the mobile robot the ability to detect colored (white/light-black/dark) lines or surfaces. The sensor working operation is based on detecting reflected light being emitted by its infrared transmitter [20]. By measuring the amount of reflected infrared light, it can detect transitions from light to dark (lines) or even objects directly in front of it. If a HIGH value is received, the surface/line is dark/black and if the received value is LOW, the surface/line is light/white. This principle will be used later for following the optimal path derived from the path planning and obstacle avoidance algorithm using two line-follower sensors.

4 Path Planning and Obstacle Avoidance

4.1 Obstacle-free Path Planning Based on Dijkstra Algorithm

In this section, a path planning and obstacle avoidance algorithm is designed and its output transferred to a global motion planning for the MR [29]. Path planning is known to be among hard-to-solve problems since its main purpose is to generate an optimal path driving the MR from a starting point to a target point in an environment known a priori [28]. The path planning problem solution can be either global (when the problem is solved offline before starting the MR mission) or local (when the environment is sensed while the robot is moving alongside its trajectory). The arrival of new information may contribute to update the optimal path. Dijkstra's algorithm has shown high competitiveness when compared to similar algorithms such as A* and its variants. Due to its usefulness, it has been adopted in many applications such as Google Maps [29]. Dijkstra's algorithm (as a global algorithm) has been combined to a dynamic window used as a local path planning algorithm in an unknown indoor environment [30] such as for logistics applications. In this paper, an approach using Dijkstra algorithm, first, in a global level and then, iteratively in a local level is designed, evaluated and tested on a real DIY MR. The key idea is to generate a global optimal path. During the MR motion, information received from an ultrasonic sensor (used as an obstacle detector) will serve to update the robot trajectory in order to avoid collision.

In this work, a MR is assumed to move inside its environment including obstacles. A path planning solution is the optimal sequence of movements that drive the MR from a starting point to a target point while following the shortest path and avoiding obstacles. The MR close environment is modeled using graph theory as a diagraph $G = \langle V, E, W \rangle$ where V is the set of vertices (including the source vertex s), E is the set of edges and W is the set of edges' weights [36]. Some of the vertices are connected through edges having non-zero positive weights. The idea behind Dijkstra algorithm is to find the sub-graph having the minimum distance between the source vertex and all other vertices including the target vertex. Mathematically, path planning is defined as an optimization problem as in Eq. (4).

Among the efficient algorithms previously used for solving **problem 1**, Dijkstra's algorithm is adopted in this paper. The main concept of this algorithm is to construct a set of vertices starting at the source vertex and ending at the target vertex such that the sum of edges' weights (called distances) is minimal. **Problem 1** is posed as follows.

Problem 1: Shortest path in an obstacle-free digraph

Find the sub-graph $Q \in V$ including the source and target vertices which minimizes the distance from s to t . without loss of generality, let's consider the source vertex s as #1 and the target vertex t as # p .

$$d_{st} = \sum_{j=1}^p d_{j,j+1}, j \in Q \quad (4)$$

Q is assumed to be non-empty and includes p vertices, i.e., $1 \leq p \leq N$. The vertex $j+1$ is adjacent to the preceding vertex j .

The steps of this algorithm are presented as follows.

Algorithm 1: Dijkstra_Shortest_Path (G, s, t)

// Inputs: An oriented diagraph G , a source and target vertices (s, t).

// Outputs: The shortest path from s to t and the length of such path.

1. Initialize the current vertex c to s and the distance between the current vertex and every vertex v in the graph to infinity: $d(v) \leftarrow \infty$
2. Initialize the optimal path set of vertices to the source vertex s : $Q = \{s\}$.
3. Update the distance of the vertices adjacent to the current vertex as:

$$d(v) \leftarrow \min(d(v), d(c) + W(c, v)) \quad (5)$$

where $W(c, v)$ is the weight of the edge (c, v) .

4. Compare the distances of those vertices to the current vertex.
 5. Move to the vertex with the smallest distance and mark it as the current vertex.
 6. Include this vertex in the optimal set of vertices, Q .
 7. Repeat steps 3-6 until reaching the target vertex.
 8. Output the optimal set of vertices and the sum of the consecutive weights from the source to the target.
-

4.2 Path Planning with Obstacle Avoidance

If the environment of the MR modeled as a diagraph includes obstacles, the path planning avoiding obstacles problem is formulated as in **Problem 2** below.

Problem 2: Shortest path in a digraph including obstacles

Find the sub-graph $Q' \in Q$ including the source and target vertices which minimizes the distance from s to t while avoiding obstacles.

Based on the obstacle-free Dijkstra algorithm (**Algorithm 1**), a solution for the path planning with obstacles problem (**Problem 2**) is designed and implemented in this paper.

Algorithm 2: Proposed path planning algorithm avoiding obstacles (Q)

// Inputs: The optimal path derived from the obstacle-free path planning algorithm Q solution of **Problem 1**.

(Continued)

Algorithm 2: Continued

// Outputs: The shortest path $Q' \in Q$ from the source s to the target vertex t avoiding obstacles.

1. Follow the obstacle-free path derived from Dijkstra algorithm (**Algorithm 1**) starting at the source vertex.
 2. Stop at each vertex and test whether the following vertex includes an obstacle.
 3. If an obstacle is found, the vertex including the obstacle is algorithmically excluded from the path by assigning infinity to its distance (An extremely high value is assigned in practice).
 4. Run **Algorithm 1** using the new graph from which the vertex including the obstacle is excluded.
 5. Repeat the process until reaching the target vertex.
-

4.3 Results

In this sub-section, the proposed algorithm for path planning and obstacle avoidance is tested on a case study to verify its performance. The derived solution is recorded in a look-up table and then transferred into a control motion law applied practically to the mobile robot designed in Section 3 above. The environment is modeled as a diagraph composed of 9 nodes (vertices) numbered from 1 to 9 and shown in Fig. 4a. The edges as well as the diagraph weights are provided in Tab. 4.

Table 4: The diagraph nodes, edges and weights

Node i	1	1	2	2	2	3	3	4	4	5	5	6	9	7	5	9	9	9	3
Node j	2	3	3	4	6	5	9	5	6	6	7	7	7	8	8	7	5	3	6
W(i,j) (10 cm)	5	19	13	8	28	5	8	35	27	12	9	10	7	18	23	22	11	8	15

Dijkstra shortest path algorithm (**Algorithm 1**) was run on the diagraph of Fig. 3a. The starting vertex was chosen as #1 and the target vertex is #8. The obtained optimal path is found to be $Q' = \{1, 2, 3, 5, 8\}$ (Fig. 3b) and the optimal length is found to be 460 cm. Two obstacles were arbitrarily placed in nodes 2 and 5. As per the designed algorithm, the MR should stop at every node it reaches in order to test the existence of eventual obstacles at its adjacent node. From the beginning, since an obstacle exists in node 2, the MR should update its path by applying **algorithm 1** to the diagraph after excluding node 2. Node 2 is here excluded by assigning a high value (5000 in our case) to all edges connecting it to all its adjacent nodes. The optimal path is then updated to become $Q' = \{1, 3, 5, 8\}$ (Fig. 3c) having 470 cm as length. The MR should move now to node 3. While stopping at node 3, the MR tests if there is any obstacle in the next optimal path node. Since node 5 includes an obstacle, the optimal path is updated by excluding node 5 using the same way as in node 2. **Algorithm 1** is run on the new diagraph. The final optimal path is $Q' = \{1, 3, 9, 7, 8\}$ (Fig. 3d) having 520 cm as an optimal length. Since the improved Dijkstra algorithm developed in this paper is big- $O(n^2)$, its convergence time is proportional to the square of the number of nodes in the diagraph modeling the environment. In practice, the MR is expected to operate indoor where the number of nodes is relatively small and consequently the convergence time will be relatively small. The convergence time depends also on the microcontroller on which the algorithm is implemented. In our case, the used Arduino board has acceptable computational efficiency and memory capacity which achieved in this case study good performance.

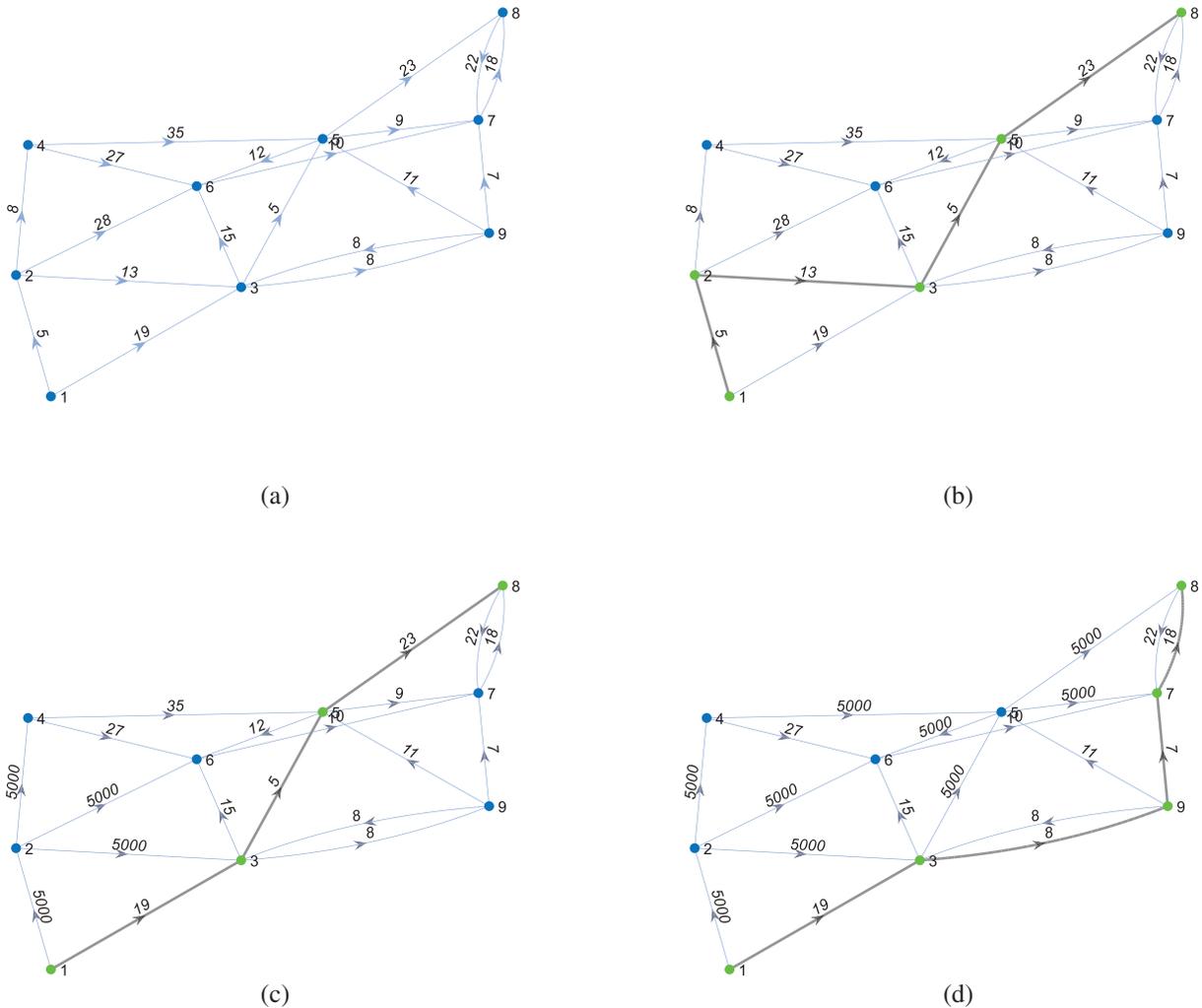


Figure 3: Illustration of the diagraphs of the path planning and obstacle avoidance implementation

4.4 Experimental Setup and Implementation

In order to practically implement the optimal path avoiding obstacles, an experimental setup around the DIY MR designed and assembled in the previous sections (Fig. 4) will be used.

In order to make the MR follow the optimal path marked using a black line, two KY-033 line-follower sensors have been used [27]. The line following logic is provided in Tab. 5. Theoretically, the MR can track the trajectory at any velocity less than its maximum velocity. However, during the experimentations, we were limited to the half of the maximum velocity for educational purpose in order to prevent losing trajectory tracking characteristic.

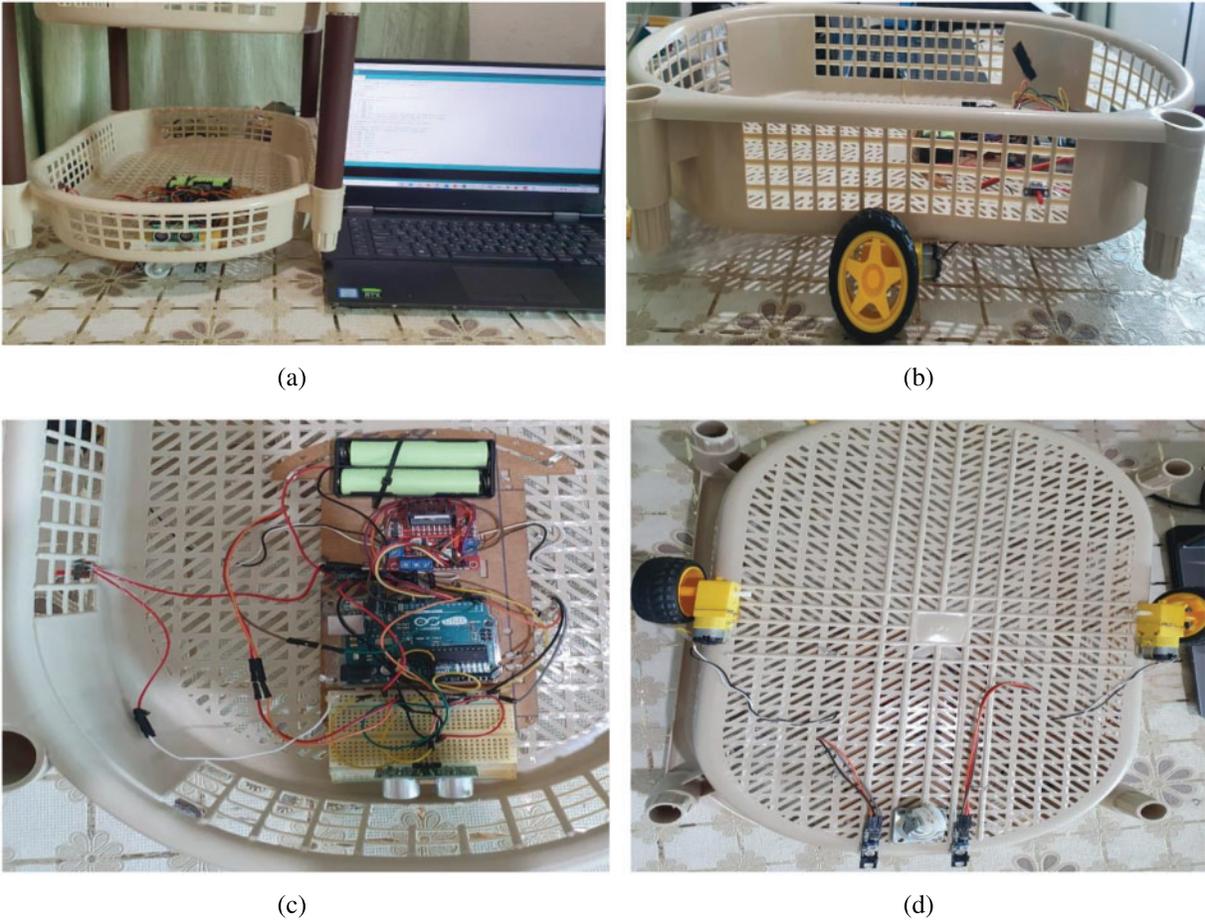


Figure 4: Experimental setup views: (a) Front (b) Left side (c) Top (d) Bottom

Table 5: Line following logic

Left sensor	Right sensor	Movement
HIGH	HIGH	Stop
LOW	HIGH	Right
HIGH	LOW	Left
LOW	LOW	Forward

Based on [Tab. 5](#), the optimal path is transferred into motion control movement as depicted in the movement sequence illustrated in [Fig. 5](#).

Algorithm 3: *Control motion (Q)*

// Inputs: The optimal path avoiding obstacles Q solution of **Problem 2**.

// Outputs: MR movement sequence.

1. Stop at starting node **1** for 1 s.
 2. Move Forward from node **1** to node **3**.
 3. Stop at node **3** for 1 s.
 4. Turn Right and move Forward from node **3** to node **9**.
 5. Stop at node **9** for 1 s.
 6. Turn Left and move Forward from node **9** to node **7**.
 7. Stop at node **7** for 1 s.
 8. Turn Right and move Forward from node **7** to node **8**.
 9. Stop at the target node **8**.
-



Figure 5: Illustration of the path planning avoiding obstacles

5 Conclusion

In this paper, a low-cost robotics platform for education and research purposes has been designed and implemented in the first part. The educational aim of this platform is to implement Science-Technology-Engineering-Art-Mathematics (STEAM) framework using an affordable system. The progressive steps of implementation included the mechanical 3D design, the hardware specification and the programming. In the second part of the paper, an optimal path planning avoiding obstacles algorithm has been developed based on an improved Dijkstra algorithm usually operating in environments without obstacles. The main contributions of this paper are the systematic procedure used in sizing and implementation of the mobile robot and the modification reported to the diagraph using a technique excluding the nodes including obstacles. The designed robot can support two to three trays which makes it useful for indoor logistics operations as well as in nonholonomic MR

identification and control. In addition, the MR is extendable to other applications in computer vision and artificial intelligence. Implementing other path planning algorithms and comparing them to the improved Dijkstra algorithm are expected as perspectives of the current study.

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Conflicts of Interest: The authors declare that they have no conflicts of interest to report regarding the present study.

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