



# North South University

## CSE 495: INTRODUCTION TO ROBOTICS

**SUMMER 2023**

Section: 1

### Home Work 2

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## Answer to question NO-01

a

Differential Flatness Method: Usually utilizes less computing power as it concentrates on determining trajectories that provide smoother and more effective pathways by satisfying system dynamics. It doesn't require a protracted search procedure.

A\* method: Generally require more computational resources, especially in complex environments, as it searches a discrete grid or graph representation of the environment for an optimal path.

b

For obstacle avoidance, we would use the A\* method. It is specially designed to find collision-free paths in environment with obstacles. It considers discrete obstacle representations and efficiently avoids collisions.



For movement in an obstacle-free 3D space, the differential flatness method is more appropriate. It generates smooth and feasible trajectories that satisfy system constraints without the need for complex pathfinding algorithms.

❑ Drawbacks of using  $A^*$  in obstacle-free space is it may introduce ~~unnecessary~~ unnecessary complexity and computational overhead when obstacles are absent.

❑ Drawbacks of using differential flatness in obstacle-filled space is Differential Flatness doesn't handle obstacle directly, so it may lead to collisions or require additional collision avoidance strategies.

## Answer to question No-2

### Pose stabilization :

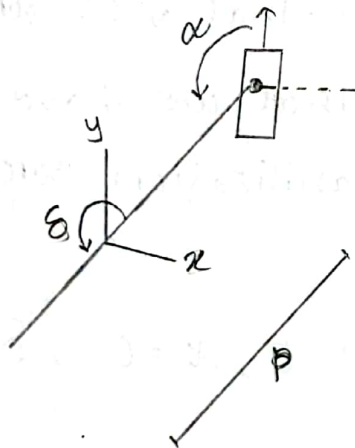
Given that,  $\dot{x}(t) = v(t) \cos \theta(t)$

$$\dot{y}(t) = v(t) \sin \theta(t)$$

$$\dot{\theta}(t) = \omega(t)$$

where control input are  $v$  and  $\omega$ .

In polar co-ordinates, pose stabilization of a unicycle robot.



where dynamic equations are,

$$\dot{p}(t) = -v(t) \cos \alpha(t)$$

$$\dot{\alpha}(t) = \frac{v(t) \sin \alpha(t)}{P(t)} - \omega(t)$$

$$\dot{\theta}(t) = \frac{v(t) \sin \alpha(t)}{P(t)}$$



By expressing the dynamics in polar form, a Lyapunov stability now easily can performed Lyapunov function,

$$V(p, \alpha, \theta) = \frac{1}{2} p^2 + \frac{1}{2} (\alpha^2 + k_3 \delta^2)$$

for closed loop control law,

$$\dot{V} = k_1 p \cos \alpha$$

$$\omega = k_2 \alpha + k_1 \frac{\sin \alpha \cos \alpha}{\alpha} (\alpha + k_3 \delta)$$

where,  $k_1, k_2, k_3 > 0$

Now, we have to show that  $\dot{V} < 0$  for the chosen control inputs. For that we have to show that system is stabilization for Lyapunov function,

$$1) V(0, 0, 0) = 0 \text{ for } p = 0, \alpha = 0, \delta = 0$$

$$2) V(p, \alpha, \delta) > 0 \text{ for } p \neq 0, \alpha \neq 0, \delta \neq 0 \text{ if } k_3 > 0$$

$$3) \dot{V}(p, \alpha, \theta) = p\dot{p} + \alpha\dot{\alpha} + k_3 \delta\dot{\delta}$$

$$= -k_1 p^2 \cos^2 \alpha - k_2 \alpha^2$$

### Answer to question No-3

Given nonlinear system,

$$\dot{x}_1 = -x_1 + x_2^3$$

$$\dot{x}_2 = -x_2 + u$$

using Lyapunov function  $v = \frac{1}{2} x_1^2 + \frac{1}{4} x_2^4$

Now, we have to find control  $u$  for stabilize the system,

Choose,  $v(x, u) = \frac{1}{2} x_1^2 + \frac{1}{4} x_2^4$

Now, if a system is given  $\dot{x} = f(x, u)$ , there exists a function  $v(x, u)$  such that,

1)  $v(x, u) = v(0, 0)$  for  $x=0$  and  $u=0$

2)  $v(x, u) > 0$  for  $x \neq 0, u \neq 0$

3)  $\dot{v}(x, u) < 0$  for  $x \neq 0, u \neq 0$

then,  $\dot{x} = f(x, u)$  is stable, then we will design controller  $u$  such that conditions 1, 2, 3 are met.

For given Lyapunov function,  $X = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

1)  $x=0$  then  $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \therefore v(0, 0) = 0$

2) for  $v(x_1, x_2) = \frac{1}{2} x_1^2 + \frac{1}{4} x_2^4$   
 $\therefore v(x_1, x_2) > 0 \quad [x_1 \neq 0 \text{ and } x_2 \neq 0]$

$$\begin{aligned}
 3) \dot{V}(x_1, x_2) &= x_1 \cdot \dot{x}_1 + x_2^3 \dot{x}_2 \\
 &= x_1(-x_1 + x_2^3) + x_2^3(x_2 + u) < 0 \\
 \text{[for } x_1 \neq 0, x_2 \neq 0 \text{ if } u < -\frac{1}{2} \cdot 2x_1 \frac{d}{dt}(x_1) + \frac{1}{4} \cdot 4x_2^3 \frac{d}{dt}(x_2)] \\
 &= x_1 \dot{x}_1 + x_2^3 \dot{x}_2
 \end{aligned}$$

$$\begin{aligned}
 \text{So, } x_1(-x_1 + x_2^3) + x_2^3(-x_2 + u) < 0 \\
 \Rightarrow -x_1^2 + x_1 x_2^3 - x_2^4 + x_2^3 u < 0 \\
 \therefore u < \frac{x_1^2 - x_1 x_2^3 + x_2^4}{x_2^3} \text{ will result in a} \\
 \text{stable system where control } u \text{ used for} \\
 \text{stabilization.}
 \end{aligned}$$

Answer to question NO-4

A\* grid-based search algorithm: A\* algorithm is a 2D based shortest path planning algorithm which is also known as a label correcting algorithm. It is a modified version of Dijkstra's algorithm. In Dijkstra's algorithm the goal ~~vector~~ vertex  $g_a$  is not taken into account, potentially leading



to wasted effort in case where the greedy choice makes no <sup>progress</sup> ~~program~~ towards the goal.

A\*: Minimize cost of arrival and cost to go. So, A\* prioritized based on cost of arrival  $c(q)$  plus an approximate cost to go  $h(q)$ . This provides a better estimate of the total quality of a path than just using the cost of arrival alone.

A\* algorithm:  $f(q) = c(q) + h(q)$

here,  $f(q)$ ,  $c(q)$  and  $h(q)$  are respectively estimated cost, cost of arrival and minimum cost to go.

Data $\rightarrow q, q_g, G$	$q_i \rightarrow$ Initial vertex
Result $\rightarrow$ Path	$q_g \rightarrow$ Goal vertex
	$G \rightarrow$ Graph

Here,

$c(q) = \infty, f(q) = \infty, \forall q;$

here all the vertex  $q$ , cost of arrival and estimated cost are infinite.



$$c(q) = 0, f(q) = 0 = h(q)$$

for initial vertex  $q_1$ , cost of arrival are zero,

For that estimated cost are equal to minimum cost to go.

$$Q = \{q_1\}$$

while  $Q$  is not empty

$$q = \arg \min_{q' \in Q} f(q')$$

$$\text{if } q = q_1$$

return path

$$Q.remove(q)$$

$$\text{for } q' \in \{q' \mid (q, q') \in E\} \text{ do}$$

$$\tilde{c}(q') = c(q) + c(q, q')$$

$$\text{if } \tilde{c}(q') < c(q') \text{ then}$$

$$q'.parent = q$$

$$c(q') = \tilde{c}(q')$$

$$f(q') = c(q') + h(q')$$

$$\text{if } q' \notin Q \text{ then}$$

$$Q.add(q')$$

return failure,

The algorithm respectively show that enter starting node in graph. Then if the graph is empty, it return failure.

The select node from graph where smallest path value and if node = goal, return path.

Expand node and generate succession, compute  $g$ th for each node.

The attach next node to back to point then again go to the step for empty graph.

### Cons:

1) In this method if the grid resolution is not enough small, there is no guarantee to find solution.

2) Limitation to simple robots where grid size is exponential in the number of DOFs.

## Answer to question no 5

Probabilistic Road Map (PRM): Probabilistic Road Algorithm is conceptually quite similar to combinatorial planners. In particular, the PRM algorithm also generated a topological graph  $G$  called a roadmap where the vertices are configurations  $q$  in free part of configuration space free and edge connect the vertices. The key insight of the PRM algorithm is that a complete characterization of the free configuration space can be avoided by sampling configuration  $q$  at random then using a collision checker to validate if  $q \in C_{free}$ .

The algorithm follows -

- 1) Randomly sample  $n$  configurations  $q_i$  from the configuration space.
- 2) Query a collision checker for each  $q_i$  to determine if  $q_i \in C_{free}$  then it is removed from the sample set.

3] Create a graph  $G = (V, E)$  with ~~vertex~~ vertices from the sampled configuration  $q_i \in C_{free}$ .

Define a radius  $r$  and create edges for every pair vertices  $q$  and  $q'$  where,

i]  $\|q - q'\| \leq r$

ii] the straight line path between  $q$  and  $q'$  is also collision free.

Cons:

1] Insufficient sampling may result in wrong path

2] PRM can be computationally intensive, especially in high-dimensional spaces.



### Answer to question NO-6

Rapidly-exploring Random Tree (RRT): RRT is used for single query problem where it grows a tree and rooted at the start configuration and embedded in  $C_{free}$ .

This RRT algorithm solves the single query problem by incrementally sampling and building the graph, starting at the initial configuration  $q$ , until the goal configuration  $q_g$  is needed.

In general, the RRT algorithm begins by initialize a tree  $T = (V, E)$  with a vertex at the initial configuration. At each iteration it follows,

- 1) Randomly sample a configuration  $q \in C$
- 2) Find the vertex  $q_{near} \in V$  that is closest to the sample configuration  $q$ .

3) Compute a new configuration  $q_{new}$  that lies on line connecting  $q_{near}$  and  $q_{such}$  that the entire line from  $q_{near}$  to  $q_{new}$  is contained in the free configuration space  $C_{free}$ .

4) Add a vertex  $q_{new}$  and an edge  $(q_{near}, q_{new})$  to the Tree  $T$ .

Cons:

1) RRT can be arbitrary bad with non-negligible probability

2) The generate suboptimal in only terms of length.

3) Difficulty for handling dynamic obstacles.

Answer to question no - 7

