# Multi-Agent Path Optimization using Calculus of Variations

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# Contents

1	List of Symbols
2	Introduction
3	Theory
4	Cases
5	Results
6	Discussion
7	Conclusion
8	Acknowledgements
$\mathbf{A}_{]}$	ppendices
Δ	MATLAR Code

#### **Abstract**

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## 1 List of Symbols

N number of robots

M number of circular obstacles

 $x_i$  initial position of the *i*-th robot

 $y_i$  target position of the *i*-th robot

 $p_i$  path of the *i*-th robot

r radius or repulsive distance of robots

 $c_j$  center of the j-th circular obstacle

 $\hat{R}$  radius of circular obstacles

F objective function

L Lagrangian function

g barrier function

### 2 Introduction

## 3 Theory

#### 3.1 Problem Definition

#### Add description of problem here.

Let  $p_i \in C^1([0,1])^2$  be the vector valued function corresponding to the path of the *i*-th robot. Then we have that  $p_i(0) = x_i$  and  $p_i(1) = y_i$ . The length of the path is defined as

$$l_i = \int_0^1 ||p_i'(t)|| dt.$$
 (1)

#### 3.2 Objective Function

We wish to minimize the total length of all robot paths. Define F as the total path length

$$F(P) = \sum_{i=1}^{N} l_i = \int_0^1 \sum_{i=1}^{N} \|p_i'(t)\| dt,$$
 (2)

where  $P = [p_1, \dots, p_N]^{\mathsf{T}}$  is the path ensemble.

We will modify the objective function in order to represent the no-collision and obstacle constraints. As an example, consider the obstacle constraint. In order to check whether or not a robot collides with an obstacle, a function g can be defined such that it evaluates to infinity if the distance between the robot and the center of the obstacle is less than R and evaluates to 0 when not. However, g would be discontinuous and forbid the use of calculus of variations techniques.

A barrier function  $g_{d,\mu}$  is a continuous function that goes to infinity as it approaches a "barrier" value d. A parameter  $\mu$  can be varied in order for g to approach the discontinuous form. In this case, we wish to devise a function that goes to infinity when approached from above to act as a repulsor. In this project, we will consider two types of barrier functions

#### 1. Log function:

$$g_{d,\mu}(x) = \frac{-1}{\mu} \log(x - d), \quad g'_{d,\mu}(x) = \frac{-1}{\mu(x - d)}.$$
 (3)

2. Inverse function:

$$g_{d,\mu}(x) = \frac{1}{\mu(x-d)}, \quad g'_{d,\mu}(x) = \frac{-1}{\mu(x-d)^2}.$$
 (4)

Using this barrier function we can rewrite F as

$$F(P) = \int_{0}^{1} \left[ \sum_{i=1}^{N} \|p_{i}'(t)\| + \sum_{i,j=1; i \neq j}^{N} g_{r,\mu_{1}} (\|p_{i}(t) - p_{j}(t)\|) + \sum_{i=1}^{N} \sum_{j=1}^{M} g_{R,\mu_{2}} (\|p_{i}(t) - c_{j}\|) \right] dt, \quad (5)$$

and the optimization problem becomes an unconstrained optimization problem.

#### 3.3 First-Order Necessary Conditions

To derive first-order necessary condition, we first identify the Lagrangian L,

$$L(Y,Z) = \sum_{i=1}^{N} \|z_i\| + \sum_{i,j=1; i \neq j}^{N} g_{r,\mu_1} (\|y_i - y_j\|) + \sum_{i=1}^{N} \sum_{j=1}^{M} g_{R,\mu_2} (\|y_i - c_j\|).$$
 (6)

Applying the Euler–Lagrange equation,

$$L_{p'_{i,x}}(t) = \frac{p'_{i,x}(t)}{\|p'_{i}(t)\|} \implies \frac{d}{dt} \left[ L_{p'_{i,x}}(t) \right] = \frac{p''_{i,x}(t)}{\|p'_{i}(t)\|} - \frac{p'_{i,x}(t) \left( p'_{i}(t) \cdot p''_{i}(t) \right)}{\|p'_{i}(t)\|^{3}}$$
(7)

$$L_{p_{i,x}}(t) = \sum_{i,j=1; i \neq j}^{N} g'_{r,\mu_{1}} (\|p_{i}(t) - p_{j}(t)\|) \frac{p_{i,x}(t) - p_{j,x}(t)}{\|p_{i,x}(t) - p_{j,x}(t)\|} + \sum_{j=1}^{M} g'_{R,\mu_{2}} (\|p_{i}(t) - p_{j}(t)\|) \frac{p_{i,x}(t) - c_{j,x}}{\|p_{i}(t) - c_{j}\|}.$$
(8)

This differential equation will be used to find the optimal path ensemble.

- 4 Cases
- 5 Results
- 6 Discussion
- 7 Conclusion
- 8 Acknowledgements

# Appendices

# A MATLAB Code