Generating Safe Paths in Dynamic Environments By Extracting Minimum Cost Trajectories Using Obstacle Position Probability Distributions and Replanning



CS4099: Major Software Project

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The main text of this project report is NNN words long, including project specification and plan.

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Chapter 1

Design

$$P_a(x, y, t_0, t_m) = \int_{t_0}^{t_m} \mathcal{N}(\zeta_a(t), \alpha \cdot (t - t_0)^2 + \beta, x, y) \cdot (t_m - t)^{\gamma} dt$$
(1.1)

Where $\mathcal{N}(\mu, \sigma^2, x, y)$ is the evaluation of a 3D normal distribution centered at (μ_x, μ_y) with a variance of σ^2 at (x, y).

$$P_A(x, y, t_0, t_m) = \frac{\sum_{a \in A} P_a(x, y, t_0, t_m)}{|A|}$$
(1.2)

$$C_A(i,j) = \int_0^1 \exp\left(P_A(x(\lambda), y(\lambda), i_t, j_t)\right) \cdot ||i - j||_2 d\lambda$$
(1.3)

Where $x(\lambda) = (j_x - i_x) \cdot \lambda + i_x$ and $y(\lambda) = (j_y - i_y) \cdot \lambda + i_y$ are the parametric equations of the line from i to j.

$$\tilde{\zeta}_a(t) = \begin{cases}
\tilde{\zeta}_a(t - \delta t) + \zeta_a'(t) \cdot \delta t + \rho & \text{if } t > 0 \\
\zeta_a^{(0)} & \text{if } t = 0
\end{cases}$$
(1.4)

Where $\rho \sim \mathcal{U}(-\epsilon, \epsilon)$, $\epsilon > 0$, and $\zeta_a^{(0)}$ is the initial position of the obstacle.

Algorithm 1 ROADMAP(n, d, w, h, O)

Input:

- n: Maximum number of samples
- d: Maximum distance between neighbouring nodes
- O: Set of obstacles

Output:

An unweighted graph of points describing the connectivity of the environment

```
1: for i=1 to n do

2: q \leftarrow \text{RANDOMPOINT2D}(w,h)

3: if \bigwedge_{o \in O} \neg \text{COLLISION}(o,q) then

4: V \leftarrow V \cup \{q\}

5: for all q_i \in V do

6: for all q_j \in V do

7: if q_i \neq q_j \wedge ||q_i - q_j|| \leq d then

8: E \leftarrow E \cup \{(q_i, q_j)\}

9: return (V, E)
```

Algorithm 2 GetPath $(n, d, w, h, \delta, p, g, O, A, R)$

d: Maximum distance between neighbouring nodes in the roadmap

```
Input:

n: Maximum number of samples for the roadmap
```

w: Width of the scene

```
h: Height of the scene
Output:
 1: (V, E) \leftarrow \text{ROADMAP}(n, d, w, h, O)
 2: \Pi \leftarrow \text{Set}()
 3: q \leftarrow p
 4: while ||\text{BACK}(\Pi) - g||_2 > R do
        \pi \leftarrow \text{SearchGraph}(V, E, R, A, q, g)
        for all i \in \pi do
 6:
            \Pi \leftarrow \Pi \cup \{i\}
 7:
            for all a \in A do
 8:
 9:
               STEP(a)
            if \bigvee_{a \in A} ||\zeta_a(i_t) - \zeta_a(i_t)|| > \delta then
10:
               for all a \in A do
11:
                  UPDATE(\zeta_a, \tilde{\zeta_a})
12:
               q \leftarrow i
13:
               break
15: return \Pi
```

$\overline{\mathbf{Algorithm}}$ 3 SearchGraph(V, E, R, A, p, g)

```
1: Q \leftarrow \text{PriorityQueue}()
 2: D \leftarrow \text{Dictionary}()
 3: \Pi \leftarrow \text{Dictionary}()
 4: Insert(Q, p, 0)
 5: while \neg \text{Empty}(Q) do
        q, w \leftarrow \text{Pop}(Q)
        if ||q-g||_2 \leq R then
 7:
           return BACKTRACKPATH(p, q, \Pi)
 9:
        N \leftarrow \text{GetTemporalNeighbours}(V, E, q)
        for all n \in N do
10:
           \Pi_n \leftarrow q
11:
           c \leftarrow \psi \cdot C_A(q, n) + \omega \cdot D_n
12:
           D_n \leftarrow D_n + 1
13:
           if w > c then
14:
15:
              c \leftarrow w
           Q \leftarrow \text{Insert}(Q, n, q)
```

Algorithm 4 GetTemporalNeighbours(V, E, q)

```
1: S \leftarrow Set()

2: N \leftarrow \text{Neighbours}(V, E, q)

3: for all n \in N do

4: t \leftarrow ||q - n||_2/s + q_t

5: S \leftarrow S \cup \{(n_x, n_y, t)\}
```

Algorithm 5 BACKTRACKPATH (p, g, Π)

```
1: q \leftarrow g

2: S \leftarrow \text{STACK}()

3: while \Pi_q \neq p do

4: S \leftarrow \text{PUSH}(S, q)

5: q \leftarrow \Pi_q

6: S \leftarrow \text{PUSH}(S, p)

7: return S
```

Chapter 2

Experimental Setup

Algorithm 6 PF(q, g, O, A, R)

```
1: q_{min} \leftarrow q
2: p_{min} \leftarrow \infty
3: \theta \leftarrow 0
4: while \theta \leq 2\pi do
5: q' \leftarrow q + \delta t \cdot s \cdot \text{ROT}(\theta)
6: p \leftarrow U_{rep}(q', O \cup A) + U_{att}(q', g)
7: if p < p_{min} then
8: p_{min} \leftarrow p
9: q_{min} \leftarrow q'
10: \theta \leftarrow \theta + \delta \theta
11: if ||q_{min} - g|| < R then
12: return \{p_{min}\}
13: return \{q_{min}\} \cup \text{PF}(q_{min}, g, O, R)
```

Chapter 3

Discussion