# Formulation of the problem

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## I. PROBLEM FORMULATION

## A. Robot and Target Motion Model

Unicycle motion model for the mobile robot:

$$z_{k+1} = h(z_k, u_k^r), (1)$$

where

$$f(z_k, u_k^r) = z_k + \begin{bmatrix} \cos \theta_k^r \Delta t & 0\\ \sin \theta_k^r \Delta t & 0\\ \Delta t & 0\\ 0 & \Delta t \end{bmatrix} u_k^r.$$

Motion model of the target:

$$x_{k+1} = f(x_k) + w_k, \ w_t \sim \mathcal{N}(0, Q)$$
 (2)

(3)

#### B. Modeling Sensing Domain

Sensor sensing domain is represented as  $\mathcal{F}_k$   $\{[x_{1,k},x_{2,k}]\in\mathbb{R}^2|\ \|v\|_2\leq r, \angle v\in[\theta_1,\theta_2]\}$ , where v  $[x_{1,k}-z_{1,k},x_{2,k}-z_{2,k}]$ .

## C. Sensor Measurement Model

Measurement model:

$$y_k = g(x_k) + v_k, \ v_k \sim \begin{cases} \mathcal{N}(0, R) & \text{if } \gamma_k = 1\\ \mathcal{N}(0, \sigma^2 I) & \text{if } \gamma_k = 0 \end{cases}$$
 (4)

$$\gamma_k = \mathbb{1}_{\{x_k \in \mathcal{F}_k\}} \tag{5}$$

## II. MPC-BASED PATH PLANNING

## A. EKF with Limited Sensing Domain

$$\hat{x}_{k+1|k}^t = f(\hat{x}_{k|k}^t) \tag{6a}$$

$$P_{k+1|k} = A_k^i P_{k|k}^i A_k^{i'} + Q (6b)$$

$$K_{k+1}^{i} = P_{k+1|k}^{i} C_{k+1}^{i'} (C_{k+1}^{i} P_{k+1|k}^{i'} C_{k+1}^{i'} + R)^{-1}$$
 (6c)

$$\hat{x}_{k+1|k+1}^{i} = \hat{x}_{k+1|k}^{i} + \gamma_{k+1} K_{k+1}^{i} (y_{k+1} - h(\hat{x}_{k+1|k}^{t})) \tag{6d}$$

$$P_{k+1|k+1}^{i} = P_{k+1|k}^{i} - \gamma_{k+1} K_{k+1}^{i} C_{k+1}^{i} P_{k+1|k}^{i}, \tag{6e}$$

where  $A_k^i=\frac{\partial f}{\partial x}|_{x=\hat{x}_{k|k}^i}$  and  $C_{k+1}^i=\frac{\partial g}{\partial x}|_{x=\hat{x}_{k+1|k}^i}$ . The  $\hat{x}_{k|k}^t$  and  $P_{k|k}$  represent the estimated target position and covariance matrix. For notational simplicity, we define  $b_k=[\hat{x}_k^t,P_{k|k}]$  and let  $b_{k+1}=g(b_k,u_k^r)$  represent the Kalman filter defined in Eq. (6).

#### $\gamma$ is approximated by

$$\gamma_{k} \approx \frac{1}{1 + \alpha_{1} \| [x_{1,k}, x_{2,k}] - [z_{1,k}, z_{2,k}] \|_{2}^{2}} \times \frac{1}{1 + \exp\left\{-\alpha_{2} (\cos(\theta_{k}^{r} - \tilde{\theta}_{k}) - \cos(\theta_{0}))\right\}},$$
(7)

where  $\tilde{\theta}_k = \angle([x_{1,k},x_{2,k}] - [z_{1,k},z_{2,k}])$  is the direction angle from the sensor position to target position;  $\theta_0 = \frac{\theta_2 - \theta_1}{2}$  is half of the sensing angle;  $\alpha_1$  and  $\alpha_2$  are tunning parameters that controls the shape of the function. Eq. (7) can be interpreted as follows: when the robot is close to the target, it is more likely that the target can be detected; besides, the closer the target direction aligns with the center direction of the sensor, the higher possibility that the target will get detected.

## B. Path Planning for Target Search and Tracking

The MPC-based path planner with planning horizon N can be formulated as:

$$\min_{u_{1:N}} J(b_{1:N+1}, u_{1:N})$$
(8a)

s.t. 
$$z_{k+1} = f(z_k, u_k^r),$$
 (8b)

$$b_{k+1} = g(b_k, u_k^r),$$
 (8c)

$$z_{k+1} \in \mathcal{X}, \ u_{k+1}^r \in \mathcal{U}, \tag{8d}$$

$$k = 1, \dots, N, \tag{8e}$$

The objective function is

$$J(b_{1:N+1}) = \sum_{k=1}^{N+1} H(b_k)$$
(9)

$$\approx -\sum_{k=1}^{N+1} \sum_{i=1}^{L} w_i \log b_k^i.$$
 (10)

The approximation is the 0-order approximation of entropy.

### C. Possible linearization in the iterative planning process

- 1) target motion matrix A
- 2) sensor measurement matrix C
- 3) initial solution
- 4) approximate the sensor boundary
- 5) linearize robot motion model

#### III. POSSIBLE EXTENSIONS

- 1) GSF with good weight update law
- 2) efficient computation of the objective function
- 3) how to represent and compute  $\gamma$  (incorporating  $b_t$  or just using a point estimate (e.g. MAP))
- 4) incorporate negative info in a better way than  $\gamma$ .
- 5) the way to do iterative planing: updating  $\gamma$  in SQP or outside SQP; whether updating w. if not updating, can I obtain an upper bound of the error?
- 6) control of nonholonomic vehicle
- 7) make the problem a cvx optimization or some other form (e.g., proximal gradient) to better utilize the form of the problem.

1,3,4 are the possible main contributions of the work.