

Modelling an Optimal Control Problem and Application of the Pontryagin Maximum Principle

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Outline

- 1 **Modelling an Optimal Control Problem**
- 2 **Application of the Pontryagin Maximum Principle**
- 3 **Shooting Algorithm**

Modelling an Optimal Control Problem

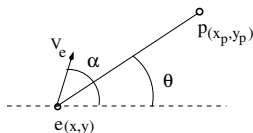


Figure: Evader and pursuer.

$$\begin{aligned} x_p &= x + L \cos \theta, \\ y_p &= y + L \sin \theta. \end{aligned} \quad (1)$$

$$\begin{aligned} \dot{x} &= V_e \cos \alpha, \\ \dot{y} &= V_e \sin \alpha. \end{aligned} \quad (2)$$

$$\begin{aligned} \dot{x}_p &= \dot{x} - L \sin \theta \dot{\theta}, \\ \dot{y}_p &= \dot{y} + L \cos \theta \dot{\theta}. \end{aligned} \quad (3)$$

Substituting Equation 2 in Equation 3

$$\begin{aligned} \dot{x}_p &= V_e \cos \alpha - L \sin \theta \dot{\theta}, \\ \dot{y}_p &= V_e \sin \alpha + L \cos \theta \dot{\theta}. \end{aligned} \quad (4)$$

Norm of V_p

Obtaining the norm L_2 of the pursuer's velocity vector.

$$V_p^2 = \dot{x}_p^2 + \dot{y}_p^2 \quad (5)$$

$$V_p^2 = V_e^2 + L^2 \dot{\theta}^2 + 2LV_e \dot{\theta} (\sin \alpha \cos \theta - \cos \alpha \sin \theta) \quad (6)$$

$$V_p^2 = V_e^2 + L^2 \dot{\theta}^2 + 2LV_e \dot{\theta} \sin(\alpha - \theta) \quad (7)$$

$$L^2 \dot{\theta}^2 + 2LV_e \sin(\alpha - \theta) \dot{\theta} + V_e^2 - V_p^2 = 0 \quad (8)$$

Solving for $\dot{\theta}$

$$L^2\dot{\theta}^2 + 2LV_e \sin(\alpha - \theta)\dot{\theta} + V_e^2 - V_p^2 = 0 \quad (9)$$

$$\dot{\theta} = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a} \quad (10)$$

$$a = L^2, \quad b = 2LV_e \sin(\alpha - \theta), \quad c = V_e^2 - V_p^2, \quad \rho = \frac{V_e}{V_p}. \quad (11)$$

$$\dot{\theta} = \frac{-2LV_e \sin(\alpha - \theta) \pm \sqrt{(2LV_e \sin(\alpha - \theta))^2 - 4L^2(V_e^2 - V_p^2)}}{2L^2} \quad (12)$$

After manipulation

$$\dot{\theta} = \frac{V_p}{L} \left(-\rho \sin(\alpha - \theta) \pm \sqrt{1 - \rho^2 \cos^2(\alpha - \theta)} \right) \quad (13)$$

The Optimal Control Problem

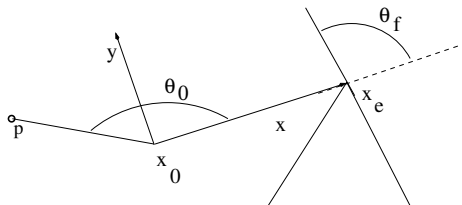


Figure: Reference frame defined with respect to the evader initial position.

- In formulating the optimal control problem, it is convenient to let x be the independent variable (instead of time t) and to parameterize the control α by x , since the evader will move from the point $x = 0$ to the point $x' = x_e$.
- This leads to a modified system description in which the state of the rod is given by $\zeta = (y, \theta)$, and the system equation is given by:

$$\frac{d\zeta}{dx} = f(\zeta, \alpha) \quad (14)$$

The State Transition Equation

$$\frac{dx}{dt} = V_e \cos \alpha \quad (15)$$

$$\frac{dy}{dt} = V_e \sin \alpha \quad (16)$$

$$\frac{dy}{dx} = \frac{dy}{dt} \left(\frac{dx}{dt} \right)^{-1} = \tan \alpha = f_1(\zeta, \alpha) \quad (17)$$

$$\frac{d\theta}{dx} = \frac{d\theta}{dt} \left(\frac{dx}{dt} \right)^{-1} = \frac{-\rho \sin(\alpha - \theta) \pm \sqrt{1 - \rho^2 \cos^2(\alpha - \theta)}}{L \rho \cos \alpha} = f_2(\zeta, \alpha) \quad (18)$$

The Cost Function

$$J = \int_{x_0}^{x_e} \frac{d\theta}{dx} dx = \int_{x_0}^{x_e} f_2(\zeta, \alpha) dx \quad (19)$$

Application of the Pontryagin Maximum Principle

Pontryagin Maximum Principle

Hamiltonian

$$H(x, u, \lambda, t) = L(x, u, t) + \langle \lambda, \dot{x} \rangle \quad (20)$$

If (x^*, u^*) is an optimal trajectory-control pair, there exists λ^* such that

$$\dot{\lambda}^* = -\frac{\partial H}{\partial x} \quad (21)$$

The optimal control is given by

$$u^* = \underset{u}{\operatorname{argmin}} H(x^*, u, \lambda^*, t) \quad (22)$$

Application of the Pontryagin Maximum Principle

Hamiltonian

$$H(x, u, \lambda, t) = \frac{d\theta}{dx} + \begin{pmatrix} \lambda_1 & \lambda_2 \end{pmatrix} \begin{pmatrix} f_1(\zeta, \alpha) \\ f_2(\zeta, \alpha) \end{pmatrix} \quad (23)$$

$$H(x, u, \lambda, t) = f_2(\zeta, \alpha) + \lambda_1 f_1(\zeta, \alpha) + \lambda_2 f_2(\zeta, \alpha) \quad (24)$$

$$H(x, u, \lambda, t) = \lambda_1 f_1(\zeta, \alpha) + (1 + \lambda_2) f_2(\zeta, \alpha) \quad (25)$$

$$H(x, u, \lambda, t) = \lambda_1 \tan \alpha + (1 + \lambda_2) \frac{-\rho \sin(\alpha - \theta) \pm \sqrt{1 - \rho^2 \cos^2(\alpha - \theta)}}{L\rho \cos \alpha} \quad (26)$$

Application of the Pontryagin Maximum Principle

Adjoint Equation

$$\dot{\lambda}^* = -\frac{\partial H}{\partial x} \quad (27)$$

$$\begin{pmatrix} \dot{\lambda}_1 \\ \dot{\lambda}_2 \end{pmatrix} = \begin{pmatrix} -\frac{\partial H}{\partial y} \\ -\frac{\partial H}{\partial \theta} \end{pmatrix} \quad (28)$$

$$\begin{pmatrix} \dot{\lambda}_1 \\ \dot{\lambda}_2 \end{pmatrix} = \begin{pmatrix} 0 \\ -(1 + \lambda_2) \frac{\cos(\alpha - \theta)}{L \cos \alpha} \left(1 - \frac{\rho \sin(\alpha - \theta)}{\sqrt{1 - \rho^2 \cos^2(\alpha - \theta)}} \right) \end{pmatrix} \quad (29)$$

Application of the Pontryagin Maximum Principle

Therefore, λ_1 is constant and the Hamiltonian can be written as:

$$H(x, u, \lambda, t) = K \tan \alpha + (1 + \lambda_2) \frac{-\rho \sin(\alpha - \theta) \pm \sqrt{1 - \rho^2 \cos^2(\alpha - \theta)}}{L \rho \cos \alpha} \quad (30)$$

We seek the control input α that minimises J , subject to the system equation $\dot{\zeta} = f(\zeta, \alpha)$, and subject to the boundary conditions $\theta(0) = \theta$, $y(0) = y(x_e) = 0$

The optimal control is given by

$$u^* = \underset{u}{\operatorname{argmin}} H(x^*, u, \lambda^*, t) \quad (31)$$

Shooting Algorithm

Shooting Algorithm

Algorithm OptimalEvaderControl(θ, x_e)

- 1 Choose initial values for the Lagrange multipliers λ_i
- 2 Let $j \leftarrow 0$, $\theta(0) \leftarrow \theta$, $y(0) \leftarrow 0$, $x(0) \leftarrow 0$.
- 3 Choose $\alpha(j)$ that minimizes the Hamiltonian,
 $\alpha^*(j) = \operatorname{argmin}_{\alpha} H(\zeta^*(j), \alpha, \lambda^*(j), x(j))$
- 4 Integrate the state equations to determine $\zeta(j+1)$
 $y(j+1) \leftarrow y(j) + f_1(\zeta(j), \alpha(j))\Delta x$
 $\theta(j+1) \leftarrow \theta(j) + f_2(\zeta(j), \alpha(j))\Delta x$
- 5 Integrate the adjoint equation for λ_2 to determine $\lambda_2(j+1)$
- 6 $x(j+1) \leftarrow x(j) + \Delta x$
- 7 If $y(j)$ and $y(j+1)$ have different sign, then the system has crossed the x axis
If $|x(j+1) - x_e| < \epsilon$ then the optimal trajectory is given by $\alpha(0), \dots, \alpha(j)$
- 8 If $y(j)$ and $y(j+1)$ have different sign, but $|x(j+1) - x(j)| > \epsilon$ then we have missed the boundary condition.
In this case, adjust the initial values for λ and go to step 2.
- 9 If $y(j)$ and $y(j+1)$ have the same sign, then we have not crossed the x -axis, and we continue to iterate forward: $j \leftarrow j+1$, go to step 3.

Thanks... Questions?

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