

# A Boundary Value Problem Approach to Autonomous Rendezvous in Industrial Human-Drone Interaction

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## Appendix A. BACKGROUND ON INDIRECT METHOD

Let  $[t_0, t_f]$  denote the finite time horizon, where  $t_0 \in \mathbb{R}$  is the initial time,  $t_f \in \mathbb{R}$  is the final time, and  $t \in [t_0, t_f]$  denotes the continuous time variable. Let  $\mathcal{U}$  denote the set of admissible control functions  $u : [t_0, t_f] \rightarrow U \subseteq \mathbb{R}^m$ , where  $U$  is a compact subset of the control space, and let  $\mathcal{S}$  denote the set of admissible state trajectories  $s : [t_0, t_f] \rightarrow \mathbb{R}^n$  with initial condition  $s(t_0) = s_0$ . The mapping  $f : \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R} \rightarrow \mathbb{R}^n$  defines the system dynamics, the mapping  $\xi : \mathbb{R}^n \rightarrow \mathbb{R}^q$  defines the terminal constraints at time  $t_f$ , the function  $L : \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R} \rightarrow \mathbb{R}$  defines the running cost, and the function  $\Phi : \mathbb{R}^n \rightarrow \mathbb{R}$  defines the terminal cost.

The optimal control problem, defined over the finite time horizon  $[t_0, t_f]$ , consists in determining  $u^*(\cdot) \in \mathcal{U}$  and the corresponding optimal state trajectory  $s^*(\cdot) \in \mathcal{S}$  that minimize the performance index

$$J[u(\cdot)] = \Phi(s(t_f)) + \int_{t_0}^{t_f} L(s(t), u(t), t) dt, \quad (\text{A.1})$$

subject to the dynamical system

$$\dot{s}(t) = f(s(t), u(t), t), \quad s(t_0) = s_0, \quad (\text{A.2})$$

and the terminal equality constraints

$$\xi(s(t_f)) = 0. \quad (\text{A.3})$$

According to the *Pontryagin Maximum Principle (PMP)* formulated in Kopp (1962), the Hamiltonian is introduced as

$$H(s(t), u(t), \lambda(t), t) = L(s(t), u(t), t) + \lambda^\top f(s(t), u(t), t), \quad (\text{A.4})$$

where  $\lambda(t) \in \mathbb{R}^n$  is the costate vector associated with the state dynamics. The necessary conditions of optimality consist of the state dynamics

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$$\dot{s}(t) = \frac{\partial H}{\partial \lambda}(s(t), u(t), \lambda(t), t), \quad s(t_0) = s_0, \quad (\text{A.5})$$

the costate dynamics

$$\dot{\lambda}(t) = -\frac{\partial H}{\partial s}(s(t), u(t), \lambda(t), t), \quad (\text{A.6})$$

and the Hamiltonian minimization condition

$$u^*(t) = \arg \min_{u(t) \in U} H(s(t), u(t), \lambda(t), t). \quad (\text{A.7})$$

For fixed final time and terminal constraints, the terminal costate satisfies

$$\lambda(t_f) = \frac{\partial \Phi}{\partial s}(s(t_f)) + \left( \frac{\partial \xi}{\partial s}(s(t_f)) \right)^\top \nu, \quad (\text{A.8})$$

where  $\nu \in \mathbb{R}^q$  is the multiplier vector associated with the terminal equality constraints. For free terminal state components, indexed by  $i \in \mathcal{I} := \{1, \dots, I\}$ , the corresponding costate components satisfy

$$\lambda_i(t_f) = 0, \quad \forall i \in \mathcal{I}. \quad (\text{A.9})$$

When the final time  $t_f$  is free, the Hamiltonian satisfies the additional transversality condition

$$H(s(t_f), u(t_f), \lambda(t_f), t_f) = 0. \quad (\text{A.10})$$

The coupled system composed of the state and costate equations, together with the Hamiltonian minimization and transversality conditions, defines a *two-point Boundary Value Problem (BVP)* (Avvakumov and Kiselev (2000)) in the unknown trajectories  $s : [t_0, t_f] \rightarrow \mathbb{R}^n$ ,  $\lambda : [t_0, t_f] \rightarrow \mathbb{R}^n$ , and  $u : [t_0, t_f] \rightarrow U$ . The boundary specification is mixed in nature, with initial conditions imposed on the state vector at time  $t_0$  and terminal conditions imposed on the costate vector and multipliers at time  $t_f$ . Hence, the BVP formulation constitutes the foundation of the *indirect method* (Wang (2009)), reformulating the optimal control problem as a differential system where the necessary conditions of the PMP are enforced through boundary conditions imposed at both the initial and final times.

## Appendix B. STATE-COSTATE DYNAMICS

According to the PMP, the necessary conditions of optimality for a continuous-time system with Hamiltonian  $H^d(s^d(t), u^d(t), \lambda^d(t), t)$  consist of the canonical equations

$$\dot{s}^d(t) = \frac{\partial H^d}{\partial \lambda^d}(s^d(t), u^d(t), \lambda^d(t), t), \quad s^d(t_0) = s_0^d, \quad (\text{B.1})$$

$$\dot{\lambda}^d(t) = -\frac{\partial H^d}{\partial s^d}(s^d(t), u^d(t), \lambda^d(t), t), \quad (\text{B.2})$$

together with the Hamiltonian minimization condition

$$u^{d,*}(t) = \arg \min_{u^d(t) \in \mathcal{U}} H^d(s^d(t), u^d(t), \lambda^d(t), t). \quad (\text{B.3})$$

By substituting the explicit dynamics and cost function of the drone, the extended Hamiltonian reads

$$\begin{aligned} H^d(s^d(t), u^d(t), t) &= L^d(s^d(t), u^d(t), t) + \lambda_p^d(t)^\top v^d(t) \\ &\quad + \lambda_v^d(t)^\top \left( -\frac{k^d}{m^d} v^d(t) + \frac{1}{m^d} u^d(t) \right), \end{aligned} \quad (\text{B.4})$$

where the running cost is

$$\begin{aligned} L^d(s^d(t), u^d(t), t) &= \|u^d(t)\|_R^2 + \|v^d(t)\|_Q^2 \\ &\quad + (v^d(t) - v_{\text{safe}}(t) \hat{v}^d(t))^\top S(v^d(t) - v_{\text{safe}}(t) \hat{v}^d(t)). \end{aligned} \quad (\text{B.5})$$

The canonical state equations follow from the partial derivative of the Hamiltonian with respect to the costate variables:

$$\frac{\partial H^d}{\partial \lambda^d(t)} = \left( \frac{\partial H^d}{\partial \lambda_p^d(t)}, \frac{\partial H^d}{\partial \lambda_v^d(t)} \right) \Rightarrow \begin{cases} \dot{p}^d(t) = v^d(t), \\ \dot{v}^d(t) = -\frac{k^d}{m^d} v^d(t) + \frac{1}{m^d} u^d(t). \end{cases} \quad (\text{B.6})$$

The Hamiltonian minimization condition gives

$$\frac{\partial H^d}{\partial u^d(t)} = 0 \quad \Rightarrow \quad u^{d,*}(t) = \text{proj}_{\mathcal{U}} \left( -\frac{1}{2m^d} R^{-1} \lambda_v^d(t) \right), \quad (\text{B.7})$$

yielding the explicit state dynamics

$$\dot{p}^d(t) = v^d(t), \quad \dot{v}^d(t) = -\frac{k^d}{m^d} v^d(t) + u^{d,*}(t). \quad (\text{B.8})$$

The costate dynamics is obtained from the negative partial derivative of the Hamiltonian with respect to the state variables:

$$\dot{\lambda}^d(t) = - \begin{pmatrix} \frac{\partial H^d}{\partial p^d(t)} \\ \frac{\partial H^d}{\partial v^d(t)} \end{pmatrix}. \quad (\text{B.9})$$

Since  $H^d(\cdot)$  depends on the drone position  $p^d(t)$  only through the safety term  $v_{\text{safe}}(t) = \tilde{v}_{\text{safe}}(p^d(t), p^h(t))$ , its partial derivative with respect to position can be expressed as

$$\frac{\partial H^d}{\partial p^d(t)} = \frac{\partial L}{\partial v_{\text{safe}}(t)} \frac{\partial v_{\text{safe}}(t)}{\partial p^d(t)}, \quad \frac{\partial v_{\text{safe}}(t)}{\partial p^d(t)} = \tilde{v}'_{\text{safe}}(d(t)) \hat{r}(t), \quad (\text{B.10})$$

where the unit relative position vector is defined as  $\hat{r}(t) = \frac{p^d(t) - p^h(t)}{\|p^d(t) - p^h(t)\|}$ , and points from the human operator toward the drone and  $\tilde{v}'_{\text{safe}}(d(t))$  is equal to

$$\tilde{v}'_{\text{safe}}(d(t)) = \begin{cases} 0, & d(t) \leq d_{\text{stop}}, \\ \frac{v_{\text{cruise}}}{d_{\text{cruise}} - d_{\text{stop}}}, & d_{\text{stop}} < d(t) < d_{\text{cruise}}, \\ 0, & d(t) \geq d_{\text{cruise}}. \end{cases} \quad (\text{B.11})$$

Let  $e(t) = v^d(t) - v_{\text{safe}}(t) \hat{v}^d(t)$ , where  $\hat{v}^d(t) = v^d(t) / \|v^d(t)\|$  represents the instantaneous heading direction of the drone. The derivative of the running cost  $L^d$  with respect to  $v_{\text{safe}}(t)$  is then

$$\frac{\partial L}{\partial v_{\text{safe}}(t)} = -2 \hat{v}^d(t)^\top S(v^d(t) - v_{\text{safe}}(t) \hat{v}^d(t)), \quad (\text{B.12})$$

which, when combined with the chain rule above, gives

$$\frac{\partial H^d}{\partial p^d(t)} = -2 \tilde{v}'_{\text{safe}}(d(t)) \hat{r}(t) \hat{v}^d(t)^\top S(v^d(t) - v_{\text{safe}}(t) \hat{v}^d(t)). \quad (\text{B.13})$$

Thus, the position costate dynamics is defined as

$$\dot{\lambda}_p^d(t) = 2 \tilde{v}'_{\text{safe}}(d(t)) \hat{r}(t) \hat{v}^d(t)^\top S(v^d(t) - v_{\text{safe}}(t) \hat{v}^d(t)). \quad (\text{B.14})$$

For the velocity component, the Hamiltonian depends explicitly and implicitly on  $v^d(t)$  through the running cost  $L^d$  and the dynamic term. The partial derivative is

$$\frac{\partial H^d}{\partial v^d(t)} = \frac{\partial L^d}{\partial v^d(t)} + \lambda_p^d(t) - \frac{k^d}{m^d} \lambda_v^d(t), \quad (\text{B.15})$$

which leads to the costate dynamics

$$\dot{\lambda}_v^d(t) = \frac{k^d}{m^d} \lambda_v^d(t) - \lambda_p^d(t) - \frac{\partial L^d}{\partial v^d(t)}. \quad (\text{B.16})$$

Considering

$$L^d = \|u^d(t)\|_R^2 + \|v^d(t)\|_Q^2 + e(t)^\top S e(t), \quad (\text{B.17})$$

the derivative of  $L^d$  with respect to  $v^d(t)$  reads

$$\frac{\partial L^d}{\partial v^d(t)} = 2Q v^d(t) + 2S e(t) \frac{\partial e(t)}{\partial v^d(t)}. \quad (\text{B.18})$$

Since  $e(t)$  depends on  $v^d(t)$  both directly and through  $\hat{v}^d(t)$ , its derivative is

$$\begin{aligned} \frac{\partial e(t)}{\partial v^d(t)} &= I - v_{\text{safe}}(t) \frac{\partial \hat{v}^d(t)}{\partial v^d(t)}, \\ \frac{\partial \hat{v}^d(t)}{\partial v^d(t)} &= \frac{1}{\|v^d(t)\|} (I - \hat{v}^d(t) \hat{v}^d(t)^\top). \end{aligned} \quad (\text{B.19})$$

Assuming that  $v^d(t)$  varies slowly so that its direction changes negligibly, the operator  $(I - \hat{v}^d \hat{v}^d{}^\top)$  acts as a projection onto the plane orthogonal to  $\hat{v}^d(t)$ . Hence,

$$(I - \hat{v}^d \hat{v}^d{}^\top) (v^d(t) - v_{\text{safe}}(t) \hat{v}^d(t)) \simeq 0, \quad (\text{B.20})$$

which implies that only the component of  $v^d(t)$  along its heading direction is relevant in the derivative. Under this assumption, (B.20) simplifies to the approximate form

$$\frac{\partial L^d}{\partial v^d(t)} \simeq 2(Q + S) v^d(t) - 2S v_{\text{safe}}(t) \hat{v}^d(t). \quad (\text{B.21})$$

Since  $v_{\text{safe}}(t)$  depends only on the drone position and not explicitly on the velocity, no additional coupling term appears in the velocity costate dynamics. Substituting the expression of  $\frac{\partial L^d}{\partial v^d(t)}$  into the canonical costate equation yields

$$\dot{\lambda}_v^d(t) = \frac{k^d}{m^d} \lambda_v^d(t) - \lambda_p^d(t) - 2(Q+S) v^d(t) + 2S v_{\text{safe}}(t) \hat{v}^d(t). \quad (\text{B.22})$$

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