UNIVERSIDADE ESTADUAL PAULISTA "JÚLIO DE MESQUITA FILHO" FACULDADE DE ENGENHARIA DE ILHA SOLTEIRA DEPARTAMENTO DE ENGENHARIA MECÂNICA

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Modeling and Control of Multirotor UAVs Formation

Ilha Solteira 2023

UNIVERSIDADE ESTADUAL PAULISTA "JÚLIO DE MESQUITA FILHO"

Campus de Ilha Solteira

Modeling and Control of Multirotor UAVs Formation

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Dissertation presented to the School of Engineering of Ilha Solteira - UNESP - as part of the requirements to obtain the title of Master in Mechanical Engineering.

Knowledge Area: Solid Mechanics

Ilha Solteira - SP February/2023

FICHA CATALOGRÁFICA Desenvolvido pelo Serviço Técnico de Biblioteca e Documentação

S586m	Silva, Renan Cavenaghi. Modeling and control of multirotor UAVs formation / Renan Cavenaghi Silva. Ilha Solteira: [s.n.], 2023 65 f. : il.	
	Dissertação (mestrado) - Universidade Estadual Paulista. Faculdade de Engenharia de Ilha Solteira. Área de conhecimento: Mecânica dos Sólidos, 2023	
	Orientador: Douglas Domingues Bueno Coorientador: Rodrigo Borges Santos Inclui bibliografia	
	1. UAV. 2. Dynamics. 3. Control. 4. Formation flight. Raiane da Silva Santos	65

Supervisora Tenica de Seção Seção Técnica de Referência, Atendimento ao usuário e Documentação Diretoria Técnica de Biblioteca e Documentação CRB/8 - 9999



UNIVERSIDADE ESTADUAL PAULISTA

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CERTIFICADO DE APROVAÇÃO

TÍTULO DA DISSERTAÇÃO: Modeling and Control of Multirotor UAVs Formation

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Documento assinado digitalmente **GOV.D** ALINE SOUZA DE PAULA Data: 23/02/2023 17:20:40-0300 Verifique em https://verificador.iti.br

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Ilha Solteira, 23 de fevereiro de 2023

Faculdade de Engenharia - Câmpus de Ilha Solteira Avenida Brasil, 56, 15385000, Ilha Solteira - São Paulo www.ppgem.feis.unesp.brCNPJ: 48.031.918/0015-20.

ACKNOWLEDGMENTS

The authors thank the São Paulo State University (UNESP), the team of Intelligent Materials and Structures (GMSINT), and the National Council for Scientific and Technological Development (CNPq) for the financial support through the grant $n^{0}152165/2021$ -5.

This study was financed in part by the Coordenação de Aperfeiçoamento de Pessoal de Nível Superior - Brasil (CAPES) - Finance Code 001.

ABSTRACT

The Unmanned Aerial Vehicles (UAV) have been extensively researched due to their underlying characteristics that suit them for different applications such as border surveillance, object recognition, crop spraying and aerial transportation. Despite the enthusiasm from the general community, practical tasks are restricted since the specialized equipment that the UAV have to transport in a mission have limited weight. In this sense, a possible alternative consists of using multiple UAVs in a formation flight to collaboratively transport a heavy-payload. Then, in this dissertation, the dynamic model from an UAV formation is obtained using an approach to parameterize the equation of motion in terms of the number of vehicles, such that different formation configurations can be considered. Moreover, the presented controller involves a leader-follower strategy with virtual-constraints for the vehicles to keep each particular formation. The results show that the employed control strategy is feasible for transporting a heavy-payload despite showing a persistent error due to the unmodeled dynamic from the suspended payload in the formation controller.

Keywords: UAV. Dynamics. Control. Formation Flight.

RESUMO

Os Veículos Aéreos não Tripulados (VANT) têm sido amplamente pesquisados devido às suas características que permitem seu uso em aplicações como monitoramento de fronteiras, reconhecimento de objetos, pulverização de defensivos agrícolas e transporte aéreo de cargas. Apesar do entusiasmo da comunidade, as aplicações são restritas no sentido de que o equipamento especializado para cada missão que o VANT deve transportar tem peso limitado. Nesse sentido, uma possível alternativa consiste em utilizar múltiplos veículos em um voo em formação para colaborativamente transportar uma carga pesada. Assim, neste trabalho apresenta-se o modelo dinâmico de uma formação de VANTs usando uma abordagem que permite parameterizar as equações do movimento no número de veículos, de modo que diferentes configurações de formações podem ser consideradas. Ainda, apresentase um controlador implementando a estratégia líder-seguidor com vínculos virtuais para definir e manter os seguidores em suas respectivas posições na formação. Os resultados mostram que a estratégia empregada é viável para o transporte de cargas pesadas, apesar de apresentar um erro persistente decorrente da dinâmica da carga desconsiderada no controlador da formação.

Palavras-chave: VANT. Dinâmica. Controle. Voo em Formação.

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Chapter 1 INTRODUCTION

Unmanned Aerial Vehicles (UAV) have attracted attention from both researchers and the community due to their potential to perform different applications. The advantages include the lower cost of operation and acquisition, relatively simple training, feasible to use in dangerous environments, updated regulation from certification agents, besides the autonomous operation.

In Brazil, UAVs have been extensively employed in the agriculture, with emphasis to modernize the analysis of crop health using aerial photography (e.g., with the Normalized Difference Vegetation Index). Then, farmers can act in localized regions and reduce the use of water and chemicals. The logistic industry can also benefit from UAVs in regions with higher population density, low infrastructure, or difficult terrain, where UAVs can surpass obstacles for ground vehicles and allow fast transportation of critical supplies such as medical products.

The use of UAVs in challenging missions, such as those ones involving heavy payload transportation increases the UAV complexity due to the increasing maximum take-off weight (MTOW). Then, their costs increase and, in practice, there is a very limited number of these aerial vehicles for this type of application. An interesting solution that is also the main focus of this research consists of using a formation to transport heavy payloads, where multiple UAVs collaboratively lift a payload heavier than the MTOW from each single vehicle.

This chapter is divided as follows: Section 1.1 describes the context and technical challenges that motivates this work. The objectives are outlined in Section 1.2. The published works are presented in Section 1.3. A description from the discussed topics are presented in Section 1.4.

1.1 PROBLEM DESCRIPTION

Formations are solutions that benefits from the collective mainly to perform tasks that are difficult, time-consuming, and even impossible for single individuals. Occurrences of formations in nature can be observed from migrating birds and transportation of heavy weights by ants. Portugal et al. (2014) show that the northern bald ibises, *Geronticus eremita*, position themselves in an aerodynamically efficient "V" formation to exploit the upwash vortices from the birds ahead. This biological behavior inspires research in commercial aviation to reduce the carbon-footprint by smart scheduling flight routes to allow aircrafts to fly in formation (DURANGO; LAWSON; SHAHNEH, 2016). Other species such as the Humpback whales, *Megaptera novaeangliae*, can swim in formations to manage their prey to form clusters, then the whales efficiently eat large amounts of food (HAIN et al., 1981).

Cooperative transportation using UAVs is a strategy under development by research groups with the perspective to transport heavy-payloads (RASTGOFTAR; ATKINS, 2019). The main idea consists of distributing the payload weight such that the lift force of each one is in a feasible domain. A cable is used for constraining the payload to each UAV, and it transfers a pulling force which is used to guide the payload through the desired path. A typical mission performed by UAVs has common features with manned vehicles (RAYMER, 1989), indicated in Fig. 1.1 with a sequence of well-defined steps.

- Phase 1 take-off: consists of the UAVs increasing lift to compensate the weight;
- Phase 2 cruise: consists of moving the vehicles to the designated mission area. This motion need to occur in an power efficiently manner to increase the system endurance;
- Phase 3 mission: consist of operating the equipment on the desired areas to carry out a particular task. It can be searching a specific area, following a target, and hovering at certain spots;
- Phase 4 returning cruise: the UAVs returns towards the landing area;
- Phase 5 landing: the UAVs lands.



Figure 1.1: Illustration of the steps of a typical multi-lift UAV mission.

Aerial transportation with a single vehicle, denoted as single-lift system, with a cable suspension allow one to transport cargo that do not fit inside its cargo compartment (RAST-GOFTAR; ATKINS, 2019). However, this transporting configuration also poses additional challenges due to the introduction of pendulum modes (SHI; WU; CHOU, 2018). By suspending the payload with two UAVs, i.e., twin-lift, the pendulum motion occurs perpendicular to the plane formed by the payload and attachments points (BERRIOS et al., 2014). In contrast, by using three or more UAVs, i.e., multi-lift formation, the pendulum motion from the suspended payload is reduced.

Despite the dynamic advantages, remotely piloting an UAV formation poses additional challenges due to closeness from the vehicles requiring coordination to avoid collisions. The twin-lift and multi-lift formations show a change from the equilibrium configuration which requires a change of the attitude from each vehicle to compensate the additional weight and maintain a separation. This effect increases the complexity of manually piloting a multi-lift formation (RAZ; ROSEN, 2005). However, UAVs can autonomously perform multi-lift missions using feedback control.

The autonomous operation from an UAV requires the use of onboard controllers to manage the mission path and compute the adequate set of control inputs to guide the vehicle towards the desired trajectory. Modeling the UAV is an essential step to design a suitable controller. Although modeling the dynamic from a single UAV is widely presented in the specialized literature, this step increases in complexity for the aerial transportation mission, mainly due to kinematic constraints that connects the UAV to the payload. For the single-lift system, the common approach consists of applying a coordinate transformation to yield a set of independent generalized coordinates that maps directly to the constraint hypersurface in the original coordinate configuration (SHI; WU; CHOU, 2018; GUERRERO-SÁNCHEZ et al., 2017). Nonetheless, this approach introduces singular configurations and there is an important difficulty to scale it for an arbitrary number of vehicles. An alternative approach consists in obtaining the dynamics in the original coordinate space using the Euler-Lagrange formulation for constrained systems (SALETAN; CROMER, 1970). Expressing the dynamic from the multi-lift system in the original coordinate space requires satisfying both the obtained differential equations, modeled from the Euler-Lagrange formulation for constrained systems (SALETAN; CROMER, 1970), and the algebraic equations from the imposed constraints, resulting in a set of differential and algebraic equations (DAE). To solve them for the trajectories in time, the resulting system of DAE are transformed into a set of ODE through the time-differentiation from the algebraic constraints (ASCHER et al., 1995), and this process is known in the literature by index-reduction. This procedure, however, introduces numerical instabilities because the constraint equations are solved only on acceleration level, such that the error is accumulated in the velocity and position levels. This phenomenon is widely recognized in the literature and it is named constraint-drift. This problem is solved by using the technique proposed by Baumgarte (1972), which is interpreted by Bisgaard, Bendtsen, and Cour-Harbo (2009) as artificially introducing springs and dampers in parallel with each one of the rods. Then, the constraints equations are approximately satisfied during the solution.

Based on the dynamic model describing the multi-lift formation, it is possible to evaluate different strategies to control the formation. There is an extensive amount of strategies for controlling multi-agent systems such as the Potential Field Method (PAUL; KROGSTAD; GRAVDAHL, 2008; ZE-SU; JIE; JIAN, 2012) and Leader-Follower Architecture (YUN et al., 2010; HE et al., 2018; LIANG; DONG; ZHAO, 2020). However, there is narrow discussion in the literature regarding the effects of a formation controller for multi-lift system.

1.2 OBJECTIVE

The main objective of this research is to investigate the dynamic and control of an UAV formation transporting a heavy-payload.

1.3 PUBLISHED WORKS

The work entitled "Disturbance Observer of an UAV with a Suspended Payload" was presented in the 26th International Congress of Mechanical Engineering.

The work entitled "Quaternion-Based Attitude Control of a Multirotor UAV" was presented in the workshop to celebrate the 25th anniversary from the Graduate Program of Mechanical Engineering from the School of Engineering of Ilha Solteira (PPGEM).

The following items summarize the contributions from this work:

• A dynamic model is parameterized in the number of vehicles, which allows one to obtain the EOM from the single-lift, twin-lift, and multi-lift configurations;

• A controller based on the virtual constraints is developed to maintain a formation during all flight and the implication from the suspended payload on the "V" formation

geometry during the multi-lift transportation mission.

1.4 OUTLINE

The remaining of this work is divided in chapters. Chapter 2 presents the modeling procedure from first principles, where the Euler-Lagrange formalism for constrained systems is applied to describe the dynamics from a multi-lift UAV formation. Chapter 3 presents a literature review on techniques to control UAVs. The developed method to control a UAV formation is also presented. Chapter 4 presents the results for a single-lift, twin-lift and the multi-lift UAV formation. Chapter 5 presents the final conclusions from this work along suggestions for future work.

Chapter 2 MODELING THE UAV FORMATION

This chapter presents the dynamic modeling of the multi-lift UAV formation. The requirements are introduced, then, the model from the UAV formation is derived and technical challenges are discussed.

This chapter is divided into sections described by the following chapter content. Section 2.1 defines the mathematical symbols. Section 2.2 introduces the parameters used in the modeling development throughout this text. Section 2.3 presents particularities from the construction of multirotors UAVs and the major differences from other configurations. Section 2.4 introduces the dynamic modeling of a single multirotor UAV using the Euler-Lagrange formulation. Additional considerations for system with constraints are presented in Section 2.5. The modeling of the dynamic from the multi-lift system is presented in Section 2.7. Section 2.8 presents the procedure to analyze the multi-lift system dynamic without solving the EOM.

2.1 NOMENCLATURE

For the remaining of this text, scalar variables are defined by a lower-case letter (e.g., a), vectors by a lower-case bold letter (e.g., a), and matrices by an upper-case bold letter (e.g., A). The transpose is denoted by A^{\top} . The identity $n \times n$ matrix is defined by $I_{n \times n}$, and the $n \times m$ zero matrix is defined by $\mathbf{0}_{n \times m}$.

2.2 MODELING PRELIMINARIES

Rigid bodies are representations of objects whose dimensions are relevant for the problem. The underlying assumption in a rigid body is the distance between any two of its particles is constant (NETO, 2013). Figure 2.1 illustrates a generic rigid body and an arbitrary particle p.

Consider an inertial frame of reference denoted by the tuple $\mathcal{I} : (O, \hat{i}, \hat{j}, \hat{k})$, where O denotes an arbitrary origin, and $\hat{i} = \{1 \ 0 \ 0\}^{\top}$, $\hat{j} = \{0 \ 1 \ 0\}^{\top}$ and $\hat{k} = \{0 \ 0 \ 1\}^{\top}$ the orthonormal basis that spans the Euclidean space \mathbb{R}^3 . Moreover, there is a body frame



Figure 2.1: Representation of a generic rigid body, where O is the coordinate system origin; \hat{i} , \hat{j} and \hat{k} are the basis from the inertial frame of reference and \hat{i}_b , \hat{j}_b and \hat{k}_b the basis from the body frame of reference, \boldsymbol{r} and \boldsymbol{r}_{ρ} are the position from the center of mass and the arbitrary particle respectively.

of reference $\mathcal{B}: (O_b, \hat{i}_b, \hat{j}_b, \hat{k}_b)$ with its origin O_b attached to the body center of mass at a distance $\mathbf{r}(t)$ with respect to the inertial frame, and with the same orientation from the body. Then, the position from any particle p from the rigid body relative to the center of mass is constant in the body reference frame \mathcal{B} , i.e.,

$${}^{\mathcal{B}}\boldsymbol{\rho} \equiv constant \tag{2.1}$$

The position from this arbitrary particle with respect to \mathcal{I} can be parameterized by the sum of the position from the center of mass and its distance relative to the body center of mass expressed in the inertial frame by means of the rotation matrix $\mathbf{R}(t)$, i.e.,

$$\boldsymbol{r}_{\rho}(t) = \boldsymbol{r}(t) + \boldsymbol{R}(t)^{\mathcal{B}}\boldsymbol{\rho}.$$
(2.2)

2.2.1 Rotation Matrix

The rotation matrix $\mathbf{R}(t)$ is a transformation from the special orthogonal group of transformations $SO(3) : \{\mathbf{R}(t) | \mathbf{R}(t) \in \mathbb{R}^{3 \times 3}, \mathbf{R}(t)^{\top} \mathbf{R}(t) = \mathbf{R}(t) \mathbf{R}(t)^{\top} = \mathbf{I}\}$ that has the property to rotate a vector around its origin while preserving the norm and handness (FOSSEN, 1994).

2.2.2 Euler Angles

The rotation matrix can be parameterized using Euler angles $\boldsymbol{\eta}(t) = \{\phi(t) \ \theta(t) \ \psi(t)\}^{\top}$ (FOS-SEN, 1994). By defining the rotations of yaw $\boldsymbol{T}_{\psi} : (\hat{\boldsymbol{i}}' \ \hat{\boldsymbol{j}}' \ \hat{\boldsymbol{k}}') \rightarrow (\hat{\boldsymbol{i}} \ \hat{\boldsymbol{j}} \ \hat{\boldsymbol{k}})$, pitch $\boldsymbol{T}_{\theta} : (\hat{\boldsymbol{i}}' \ \hat{\boldsymbol{j}}'' \ \hat{\boldsymbol{k}}'') \rightarrow (\hat{\boldsymbol{i}}' \ \hat{\boldsymbol{j}}' \ \hat{\boldsymbol{k}}')$, and rolling $\boldsymbol{T}_{\phi} : (\hat{\boldsymbol{i}}_b \ \hat{\boldsymbol{j}}_b \ \hat{\boldsymbol{k}}_b) \rightarrow (\hat{\boldsymbol{i}}'' \ \hat{\boldsymbol{j}}'' \ \hat{\boldsymbol{k}}'')$, the components from the rotation matrix can be obtained by

$$\boldsymbol{R} = \boldsymbol{T}_{\psi} \boldsymbol{T}_{\theta} \boldsymbol{T}_{\phi}. \tag{2.3}$$

Figure 2.2 illustrates each consecutive rotation of yaw, pitch and roll, which allows one to



Figure 2.2: Sequential transformation from the coordinate system through the yaw (a), pitch (b), and roll (c).

obtain the following equations

$$\hat{\boldsymbol{i}} = c_{\psi} \hat{\boldsymbol{i}}' - s_{\psi} \hat{\boldsymbol{j}}' \qquad \hat{\boldsymbol{i}}' = c_{\theta} \hat{\boldsymbol{i}}'' + s_{\psi} \hat{\boldsymbol{k}}'' \qquad \hat{\boldsymbol{i}}'' = \hat{\boldsymbol{i}}_{b}
\hat{\boldsymbol{j}} = s_{\psi} \hat{\boldsymbol{i}}' + c_{\psi} \hat{\boldsymbol{j}}' \qquad \hat{\boldsymbol{j}}' = \hat{\boldsymbol{j}}'' \qquad \hat{\boldsymbol{j}}'' = c_{\phi} \hat{\boldsymbol{j}}_{b} - s_{\phi} \hat{\boldsymbol{k}}_{b}$$

$$\hat{\boldsymbol{k}} = \hat{\boldsymbol{k}}' \qquad \hat{\boldsymbol{k}}' = -s_{\theta} \hat{\boldsymbol{i}}'' + c_{\theta} \hat{\boldsymbol{k}}'' \qquad \hat{\boldsymbol{k}}'' = s_{\phi} \hat{\boldsymbol{j}}_{b} + c_{\phi} \hat{\boldsymbol{k}}_{b}$$
(2.4)

where $c_{(\cdot)}$ and $s_{(\cdot)}$ denotes $\cos(\cdot)$ and $\sin(\cdot)$ respectively. The transformation matrices $T_{(\cdot)}$ are obtained from Eq. (2.4) and given by Eq. (2.5)

$$\boldsymbol{T}_{\phi} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & c_{\phi} & -s_{\phi} \\ 0 & s_{\phi} & c_{\phi} \end{bmatrix} \qquad \boldsymbol{T}_{\theta} = \begin{bmatrix} c_{\theta} & 0 & s_{\theta} \\ 0 & 1 & 0 \\ -s_{\theta} & 0 & c_{\theta} \end{bmatrix} \qquad \boldsymbol{T}_{\psi} = \begin{bmatrix} c_{\psi} & -s_{\psi} & 0 \\ s_{\psi} & c_{\psi} & 0 \\ 0 & 0 & 1 \end{bmatrix}$$
(2.5)

$$\boldsymbol{R} = \begin{bmatrix} c_{\psi}c_{\theta} & -s_{\psi}c_{\phi} + c_{\psi}s_{\theta}s_{\phi} & s_{\psi}s_{\phi} + c_{\psi}c_{\phi}s_{\theta} \\ s_{\psi}c_{\theta} & c_{\psi}c_{\phi} + s_{\phi}s_{\theta}s_{\psi} & -c_{\psi}s_{\phi} + s_{\theta}s_{\psi}c_{\phi} \\ -s_{\theta} & c_{\theta}s_{\phi} & c_{\theta}c_{\phi} \end{bmatrix} .$$
(2.6)

In this case, the transformation $\boldsymbol{R} : \left(\hat{\boldsymbol{i}}_b \ \hat{\boldsymbol{j}}_b \ \hat{\boldsymbol{k}}_b\right) \to \left(\hat{\boldsymbol{i}} \ \hat{\boldsymbol{j}} \ \hat{\boldsymbol{k}}\right)$ maps a vector from the body reference frame to the inertial reference frame. The inverse transformation, i.e., that maps a vector from the inertial reference system to the body reference system is then $\boldsymbol{R}^{\top} = \boldsymbol{T}_{\phi}^{\top} \boldsymbol{T}_{\theta}^{\top} \boldsymbol{T}_{\psi}^{\top}$.

The angular velocity vector from the body ${}^{\mathcal{B}}\boldsymbol{\omega} = {}^{\mathcal{B}}\boldsymbol{\omega}(t) = \{\omega_x(t), \omega_y(t), \omega_z(t)\}^{\top}$ is given in terms of the Euler angles, such that (FOSSEN, 1994):

$${}^{\mathcal{B}}\boldsymbol{\omega} = \mathbf{T}_{\phi}^{\top}\mathbf{T}_{\theta}^{\top} \begin{cases} 0\\ 0\\ \dot{\psi} \end{cases} + \mathbf{T}_{\phi}^{\top} \begin{cases} 0\\ \dot{\theta}\\ 0 \end{cases} + \begin{cases} \dot{\phi}\\ 0\\ 0 \end{cases}$$
(2.7)

that results in ${}^{\mathcal{B}}\omega = W_{\eta}\dot{\eta}$, with W_{η} denoting the transformation from the angular velocity in the inertial reference frame to the body reference frame (RAFFO; ORTEGA; RUBIO, 2010)

$$\boldsymbol{W}_{\boldsymbol{\eta}} = \begin{bmatrix} 1 & 0 & -s_{\theta} \\ 0 & c_{\phi} & s_{\phi}c_{\theta} \\ 0 & -s_{\phi} & c_{\phi}c_{\theta} \end{bmatrix}.$$

The Euler angles rate are obtained in terms of the body angular velocity through the inverse transformation defined by (WANG et al., 2016)

$$\dot{\boldsymbol{\eta}} = \boldsymbol{W}_{\boldsymbol{\eta}}^{-1\,\mathcal{B}}\boldsymbol{\omega},\tag{2.8}$$

such that

$$\boldsymbol{W}_{\boldsymbol{\eta}}^{-1} = \begin{bmatrix} 1 & \sin\phi \tan\theta & \cos\phi \tan\theta \\ 0 & \cos\phi & -\sin\phi \\ 0 & \sin\phi \sec\theta & \cos\phi \sec\theta \end{bmatrix}.$$

Euler angles are intuitive for understanding the geometric properties from rotations when considering isolated maneuvers. Nonetheless, once $det(\mathbf{W}_{\eta}) = c_{\theta}$, the inverse transformation is singular for $\theta = \pm \pi/2$.

2.3 MULTIROTOR UAV

A distinct feature from multirrotor UAVs is that the propulsion and control is performed by varying the angular velocity from the rotors composed of an electric motor and a fixed-pitch propeller. This characteristics allows one to achieve significant cost reduction in acquisition and maintenance since, in contrast with the helicopter configuration, the swash-plate mechanism is absent. However, the benefit of controlling the UAV by varying the rotor velocity is also challenging to scale for transporting heavier payloads, since this task usually requires bigger propellers which imply to increase the rotational inertia that slows the rotor dynamics and may render the vehicle uncontrollable (POUNDS; MAHONY, 2009; PORTER; SHIRINZADEH; CHOI, 2015). The thrust force and drag moment from the *i*th rotor can be reasonable modeled respectively by the following set of equations if small size propellers are considered.

$${}^{\mathcal{B}}\boldsymbol{f}_{i}(\omega_{r,i}) = -k_{T}\omega_{r,i}^{2}\hat{\boldsymbol{k}}_{b}$$
(2.9)

$${}^{\mathcal{B}}\boldsymbol{m}_{i}(\omega_{r,i}) = \operatorname{sign}(\omega_{r,i})k_{D}\omega_{r,i}^{2}\hat{\boldsymbol{k}}_{b}$$

$$(2.10)$$

with $\omega_{r,i}$ denotes the angular velocity from the corresponding *i*th rotor, k_T and k_D are respectively linear coefficients of thrust and moment, commonly determined from the propeller performance data. Note the negative sign in Eq. (2.9) indicates the thrust pointing to the negative \hat{k}_b axis. Moreover, the drag moment direction depends if the rotor is rotating clockwise (CW) or counterclockwise (CCW). Thus it is necessary to arrange the rotors to balance the yaw moment. This is one from the explanations of why multirotors are commonly built with an even number of rotors (e.g., quadrotors, hexarotors, and octorotors). In contrast, the yaw moment in a helicopter configuration is balanced using a tail-rotor.

In the planar multirotor configuration, the force from the *i*th rotor is applied at ${}^{\mathcal{B}}\mathbf{r}_i = \{l_r c_{\xi_i} - l_r s_{\xi_i} 0\}^{\top}$ in the body coordinate system, where ξ_i denotes the angle from the line connecting the UAV center of mass to the propeller rotation axis to the \hat{i}_b axis. Considering a symmetric arrangement of rotors, this angle is computed for the *i*th rotor by

$$\xi_i = \frac{2\pi}{n_r} i + \xi_c, \tag{2.11}$$

where the two distinct + and × configurations can be obtained by setting the offset angle ξ_c to 0 or $\frac{\pi}{n_r}$, respectively. Then, the net lift and moment exerted on the UAV center of mass are obtained respectively by

$${}^{\mathcal{B}}\boldsymbol{f} = \sum_{i=1}^{n_r} {}^{\mathcal{B}}\boldsymbol{f}_i \tag{2.12}$$

$${}^{\mathcal{B}}\boldsymbol{m} = \sum_{i=1}^{n_r} {}^{\mathcal{B}}\boldsymbol{r}_i \times {}^{\mathcal{B}}\boldsymbol{f}_i + {}^{\mathcal{B}}\boldsymbol{m}_i$$
(2.13)

The control input is defined in terms of the angular speed from each rotor and the control allocation matrix \mathbf{K}_r (KOTARSKI et al., 2021), representing a map from the quadratic angular velocity, denoted by the vector $\boldsymbol{\omega}_r^2 = \left\{ \omega_{r,1}^2 \; \omega_{r,2}^2 \; \ldots \; \omega_{r,n_r}^2 \right\}^{\top}$, of a given configuration to the control forces $\boldsymbol{u} = \left\{ u_T \; u_{\hat{i}_b} \; u_{\hat{j}_b} \; u_{\hat{k}_b} \right\}^{\top}$, with u_T denoting the $\hat{\boldsymbol{k}}_b$ component from the force vector in Eq. (2.12), and $u_{\hat{i}_b}, \; u_{\hat{j}_b}$ and $u_{\hat{k}_b}$ the moments on the UAV body axes given by Eq. (2.13) (PILJEK; KOTARSKI; KRZNAR, 2020).

$$\boldsymbol{u} = \boldsymbol{K}_r \boldsymbol{\omega}_r^2, \tag{2.14}$$

Equation (2.14) allows generalizing the forces for any configuration of multirotor. Then, the rotors speed for a particular configuration are obtained through the following equation

$$\boldsymbol{\omega}_r^2 = \boldsymbol{K}_r^{-1} \boldsymbol{u}. \tag{2.15}$$

Table 2.1 shows the matrix K_r for different multirrotor configurations. The different multirrotor configurations are named by Q, H, and O, where the letter Q denotes the quadrirrotor, H denotes the hexarrotor, and O the octorrotor configurations, respectively. The Q× configuration is widely used in practice since a camera aligned with the body coordinate system usually has its field of view clear from the rotor arm (RAO et al., 2022). Other configurations with more rotors, such as the hexacopter (H+) and octocopter (O+), are mainly used to provide more thrust and fail-safe features.

 $oldsymbol{K}_r$ representation conf. ref. $\begin{bmatrix} k_T & k_T & k_T & k_T \\ 0 & -l_r k_T & 0 & l_r k_T \\ l_r k_T & 0 & -l_r k_T & 0 \end{bmatrix}$ Q +(LUO; DU; YU, 2019) k_T k_T k_T k_T $\frac{\sqrt{2}}{2}l_rk_T - \frac{\sqrt{2}}{2}l_rk_T - \frac{\sqrt{2}}{2}l_rk_T$ $Q \times$ (RAO et al., 2022) $\frac{\sqrt{2}}{2}l_rk_T$ $\frac{\sqrt{2}}{2}l_rk_T \quad -\frac{\sqrt{2}}{2}l_rk_T \quad -\frac{\sqrt{2}}{2}l_rk_T$ $\begin{bmatrix}
2 & 2 & -k_{D} & k_{D} \\
-k_{D} & k_{D} & -k_{D} & k_{D}
\end{bmatrix}$ $\frac{k_{T} & k_{T} & k_{T} & k_{T} & k_{T} \\
\frac{\sqrt{3}}{2}l_{r}k_{T} & \frac{\sqrt{3}}{2}l_{r}k_{T} & 0 & -\frac{\sqrt{3}}{2}l_{r}k_{T} & -\frac{\sqrt{3}}{2}l_{r}k_{T}$ k_T 0 H+ $\frac{1}{2} l_r k_T - \frac{1}{2} l_r k_T - l_r k_T - \frac{1}{2} l_r k_T$ $l_r k_T$ $-k_D$ k_T k_T k_T $\frac{\sqrt{2}}{2}l_rk_T \quad l_rk_T \quad \frac{\sqrt{2}}{2}l_rk_T \quad 0 \quad -\frac{\sqrt{2}}{2}l_rk_T \quad -l_rk_T \quad -\frac{\sqrt{2}}{2}l_rk_T$ 0 O+ $\frac{\sqrt{2}}{2} l_r k_T = 0 \qquad \frac{\sqrt{2}}{2} l_r k_T = l_r k_T = \frac{\sqrt{2}}{2} l_r k_T$ $0 \qquad -\frac{\sqrt{2}}{2}l_rk_T$ $-l_r k_T$ - $-k_D$ k_D $-k_D$ $-k_D$ k_D $-k_D$ k_D k_D

Table 2.1: Multirrotors configurations and the associated control allocation matrix. Propellers indicated in red and blue represent a CW and CCW rotation respectively.

2.4 MODEL DERIVATION

The quadrotor is illustrated in Fig. 2.3. Two reference systems are defined according to Sec. 2.2 and the generalized coordinates that uniquely describes the UAV configuration is given by the vector $\boldsymbol{q}(t) = \{\boldsymbol{r}(t) \ \boldsymbol{\eta}(t)\}^{\top}$, with $\boldsymbol{r}(t) = \{x(t) \ y(t) \ z(t)\}^{\top}$ denoting the position from the UAV center of mass and $\boldsymbol{\eta}(t) = \{\phi(t) \ \theta(t) \ \psi(t)\}^{\top}$ describing the attitude parameterized using Euler angles.



Figure 2.3: Quadrotor configuration of a multirotor UAV.

The kinetic energy from the UAV is composed of the translational energy due the motion of the center of mass, and the kinetic energy due to body angular velocity, such that $\mathcal{T} = \mathcal{T}_r + \mathcal{T}_\eta$. The potential energy due to the gravitational field is computed by $\mathcal{U} = \mathcal{U}_r$. Then, the Lagrangian is given by: $\mathcal{L} = \mathcal{T} - \mathcal{U}$ (GOLDSTEIN, 2002), which allows one to obtain the equation of motion for each generalized coordinate $q_i \in q$ through the Euler-Lagrange equation (LEMOS, 2007), i.e.,

$$\frac{d}{dt}\left(\frac{\partial \mathcal{L}}{\partial q_i}\right) - \frac{\partial \mathcal{L}}{\partial q_i} = f_i, \qquad (2.16)$$

with f_i denoting the generalized force vector.

2.4.1 Translational Dynamics

The kinetic energy from the translating rigid body is computed by

$$\mathcal{T}_{\boldsymbol{r}} = \frac{1}{2} \dot{\boldsymbol{r}}^{\top} \boldsymbol{M}_{\boldsymbol{r}} \dot{\boldsymbol{r}}.$$
 (2.17)

Due to the gravitational potential field, the UAV is also subjected to a potential energy $\mathcal{U}_r = -mgz$. The partial derivative of Eq. (2.17) with respect to the velocity is obtained by $\frac{\partial \mathcal{T}_r}{\partial \dot{r}} = m\dot{r}$. Differentiating with respect to the time, the resulting equation is

$$\frac{d}{dt} \left(\frac{\partial \mathcal{T}_{\boldsymbol{r}}}{\partial \dot{\boldsymbol{r}}} \right) = m \ddot{\boldsymbol{r}}.$$
(2.18)

The potential energy depends only on the position vector. Nonetheless, its partial derivative is computed by

$$\frac{\partial \mathcal{U}_{\boldsymbol{r}}}{\partial \boldsymbol{r}} = -mg\hat{\boldsymbol{k}}.$$
(2.19)

Then, the equations for the translational dynamics in matrix form are given by

$$\boldsymbol{M_r}\ddot{\boldsymbol{r}} + \boldsymbol{f_g} = \boldsymbol{B_r}\boldsymbol{u_T} + \boldsymbol{d_r} \tag{2.20}$$

with $M_r = m I_{3\times 3}$ representing the translation inertia, $f_g = -mg\hat{k}$ is the gravitational force, and $d_r = \{d_x \ d_y \ d_z\}^{\top}$ is the disturbance forces vector acting on the UAV. The components of matrix B_r are the projections from the thrust acting on the UAV \hat{k}_b axis to the inertial reference frame, such that

$$\boldsymbol{B_r} = \begin{bmatrix} c_{\phi} s_{\theta} c_{\psi} + s_{\phi} s_{\psi} \\ c_{\phi} s_{\theta} s_{\psi} - s_{\phi} c_{\psi} \\ c_{\phi} c_{\theta} \end{bmatrix}.$$

Note that for the multirotor there is only one control term to affect the three translational coordinates. This render the multirotor UAV as an underactuated vehicle (BRANDÃO; FILHO; CARELLI, 2013).

2.4.2 Attitude Dynamics

The kinetic energy due to the rotational motion is computed by

$$\mathcal{T}_{\boldsymbol{\eta}} = \frac{1}{2} \,{}^{\mathcal{B}} \boldsymbol{\omega}^{\mathsf{T}} \boldsymbol{J}^{\mathcal{B}} \boldsymbol{\omega}. \tag{2.21}$$

Substituting Eq. (2.8) in the expression above, the kinetic energy is written in terms of the Euler angles

$$\mathcal{T}_{\boldsymbol{\eta}} = \frac{1}{2} \dot{\boldsymbol{\eta}}^{\top} \boldsymbol{W}_{\boldsymbol{\eta}}^{\top} \boldsymbol{J} \boldsymbol{W}_{\boldsymbol{\eta}} \dot{\boldsymbol{\eta}}.$$
 (2.22)

The EOM from the attitude dynamic is decoupled from the translational dynamic, such that

$$\frac{\partial \mathcal{T}_{\eta}}{\partial \dot{\eta}} = \frac{\partial}{\partial \dot{\eta}} \left(\frac{1}{2} \dot{\eta}^{\top} W_{\eta}^{\top} J W_{\eta} \dot{\eta} \right) = W_{\eta}^{\top} J W_{\eta} \dot{\eta} = M_{\eta} \dot{\eta}$$
(2.23)

and the symmetric and positive definite inertia matrix $M_{\eta} = W_{\eta}^{\top} J W_{\eta}$ is obtained, such that

$$\boldsymbol{M}_{\boldsymbol{\eta}} = \begin{bmatrix} I_{xx} & 0 & -I_{xx}s_{\theta} \\ 0 & I_{yy}c_{\phi}^{2} + I_{zz}s_{\phi}^{2} & s_{\phi}c_{\phi}c_{\theta}(I_{yy} - I_{zz}) \\ -I_{xx}s_{\theta} & s_{\phi}c_{\phi}c_{\theta}(I_{yy} - I_{zz}) & I_{xx}s_{\theta}^{2} + c_{\theta}^{2}(I_{yy}s_{\phi}^{2} + I_{zz}c_{\phi}^{2}) \end{bmatrix}.$$
 (2.24)

Differentiating Eq. (2.23) with respect to the time using the chain rule, the following equation is obtained

$$\frac{d}{dt} \left(\frac{\partial \mathcal{T}_{\eta}}{\partial \dot{\eta}} \right) = \dot{M}_{\eta} \dot{\eta} + M_{\eta} \ddot{\eta}.$$
(2.25)

The second term of the Euler-Lagrange equation corresponds to the partial derivative from the kinetic energy with respect to the coordinates. Then,

$$\frac{\partial \mathcal{T}_{\boldsymbol{\eta}}}{\partial \boldsymbol{\eta}} = \frac{\partial}{\partial \boldsymbol{\eta}} \left(\frac{1}{2} \dot{\boldsymbol{\eta}}^{\top} \boldsymbol{W}_{\boldsymbol{\eta}}^{\top} \boldsymbol{J} \boldsymbol{W}_{\boldsymbol{\eta}} \dot{\boldsymbol{\eta}} \right) = \frac{1}{2} \dot{\boldsymbol{\eta}}^{\top} \frac{\partial \boldsymbol{M}_{\boldsymbol{\eta}}}{\partial \boldsymbol{\eta}} \dot{\boldsymbol{\eta}}$$
(2.26)

By grouping the angular velocity dependent terms, the Coriolis matrix is obtained as the following (BRANDÃO; FILHO; CARELLI, 2013)

$$\boldsymbol{C}_{\boldsymbol{\eta}} = \dot{\boldsymbol{M}}_{\boldsymbol{\eta}} - \frac{1}{2} \dot{\boldsymbol{\eta}}^{\top} \frac{\partial \boldsymbol{M}_{\boldsymbol{\eta}}}{\partial \boldsymbol{\eta}}.$$
 (2.27)

The first term on the right-hand side of Eq. (2.27) corresponds to the derivative of a matrix with respect to a scalar, which consists of differentiating every component of M_{η} with respect to time. The second term consists of differentiating a matrix with respect to a vector, which is a tensor. The result is a 3 × 3 matrix C_{η} with its elements c_{ij} given by

$$\begin{split} c_{11} &= 0 \\ c_{12} &= s_{\phi} c_{\phi} (I_{yy} - I_{zz}) \dot{\theta} + s_{\phi}^{2} c_{\theta} (I_{yy} - I_{zz}) \dot{\psi} - I_{xx} \dot{\psi} c_{\theta} \\ c_{13} &= s_{\phi} c_{\phi} c_{\theta}^{2} (I_{zz} - I_{yy}) \dot{\psi} + c_{\phi}^{2} c_{\theta} (I_{zz} - I_{yy}) \dot{\theta} - I_{xx} c_{\theta} \dot{\theta} \\ c_{21} &= I_{zz} \dot{\psi} c_{\theta} + (I_{yy} - I_{xx}) (-\dot{\theta} s_{\theta} c_{\phi} + \dot{\psi} c_{\theta} c_{\phi}^{2} - \dot{\psi} c_{\theta} s_{\phi}^{2}), \ c_{22} &= -(I_{yy} - I_{xx}) s_{\phi} c_{\phi} \dot{\phi} \\ c_{23} &= \dot{\psi} s_{\theta} c_{\theta} (-I_{zz} + I_{yy} s_{\phi}^{2} + I_{xx} c_{\phi}^{2}) \\ c_{31} &= -(I_{xx} \dot{\theta} c_{\theta} - (I_{yy} - I_{xx}) (\dot{\psi} c_{\theta}^{2} s_{\phi} c_{\phi})) \\ c_{32} &= I_{zz} \dot{\psi} s_{\theta} c_{\theta} - (I_{yy} - I_{xx}) (\dot{\theta} s_{\theta} c_{\phi} s_{\phi} + \dot{\phi} c_{\theta} s_{\phi}^{2}) - (I_{yy} + I_{xx}) (\dot{\psi} s_{\theta} c_{\theta} c_{\phi}^{2} - \dot{\phi} c_{\theta} c_{\phi}^{2}) \\ c_{33} &= I_{zz} \dot{\theta} s_{\theta} c_{\theta} - (I_{yy} + I_{xx}) (\dot{\theta} s_{\theta} c_{\theta} s_{\phi}^{2}) + (I_{yy} - I_{xx}) \dot{\phi} c_{\theta}^{2} s_{\phi} c_{\phi}. \end{split}$$

Substituting the inertia and the Coriolis matrices, the model for the UAV attitude dynamic is given by

$$M_{\eta}\ddot{\eta} + C_{\eta}\dot{\eta} = u_{\eta} + d_{\eta}. \tag{2.28}$$

with $\boldsymbol{u}_{\boldsymbol{\eta}} = \{u_{\phi} \ u_{\theta} \ u_{\psi}\}$ representing the moments with respect to the inertial frame of reference, and $\boldsymbol{d}_{\boldsymbol{\eta}} = \{d_{\phi} \ d_{\theta} \ d_{\psi}\}^{\top}$ the disturbance moments vector action on the UAV.

2.4.3 Dynamic Model

Considering the equations from the translational dynamics (Sec. 2.4.1), and rotational dynamics (Sec. 2.4.2), the dynamic model from the UAV is presented by the following system of ODEs

$$\boldsymbol{M}\ddot{\boldsymbol{q}} + \boldsymbol{C}\dot{\boldsymbol{q}} + \boldsymbol{f}_g = \boldsymbol{B}\boldsymbol{u} + \boldsymbol{d}, \qquad (2.29)$$

where the term $\boldsymbol{d} = \left\{ \boldsymbol{d}_r^{\top} \ \boldsymbol{d}_{\eta}^{\top} \right\}^{\top}$ represents the disturbances forces acting on the model. Moreover, the model matrices for the full UAV dynamic are defined by the following equations

$$\boldsymbol{M} = \begin{bmatrix} \boldsymbol{m}\boldsymbol{I}_{3\times3} & \boldsymbol{0}_{3\times3} \\ \boldsymbol{0}_{3\times3} & \boldsymbol{M}_{\boldsymbol{\eta}} \end{bmatrix} \qquad \boldsymbol{C} = \begin{bmatrix} \boldsymbol{0}_{3\times3} & \boldsymbol{0}_{3\times3} \\ \boldsymbol{0}_{3\times3} & \boldsymbol{C}_{\boldsymbol{\eta}} \end{bmatrix} \qquad \boldsymbol{B} = \begin{bmatrix} \boldsymbol{B}_{r} & \boldsymbol{0}_{3\times3} \\ \boldsymbol{0}_{3\times1} & \boldsymbol{W}_{\boldsymbol{\eta}}^{\top} \end{bmatrix}$$
(2.30)

Note that the rotational dynamic is independent of the translational coordinates, however, the translational dynamic depends on the attitude, such as the control forces in these direction are projections from the thrust force.

Equation (2.29) allows one to compute the UAV trajectory depending on the commanded input history $\boldsymbol{u}(t)$ over time. It is also used to design the controller that stabilizes the system in a required position.

2.5 EULER-LAGRANGE EQUATIONS FOR CONSTRAINED SYSTEMS

The typical problem of dynamics consists of finding the trajectory that a system takes from an instant t_1 to an instant t_2 . For a system with independent coordinates, the solution is obtained by solving the Euler-Lagrange equations (Sec. 2.4), which minimizes the action computed through $S = \int_{t_1}^{t_2} \mathcal{L}(\mathbf{q}, \dot{\mathbf{q}}) dt$. This process is illustrated in Fig. 2.4, where the trajectory that satisfies the Euler-Lagrange equations is represented by a solid line and other trajectories are indicated by dashed lines.



Figure 2.4: Generic trajectories connecting the configurations at the instants t_1 and t_2 . The system takes the trajectory that satisfies the Euler-Lagrange equation indicated by the solid line.

Mechanical systems can be subject to restrictions on their motion, which consists to kinematic restraints imposed on them, such as the rod connecting a pendulum to its pivot joint (LEMOS, 2007). Considering the illustrative pendulum from Fig. 2.5, the mass attached to the rod is constrained to the perimeter of a circle with radius equal to the length l, which is equivalent to state that the mass position $(x_p \ y_p)$ is constrained by the equation $\Theta = x_p^2 + y_p^2 - l^2 = 0$.

A convenient approach used to model the constrained system consists of reparameterizing the coordinates to obtain a new set of independent coordinates, such that the



Figure 2.5: Representation of the pendulum.

constraint equations are identically satisfied in terms of the new coordinates. In the case of the illustrated pendulum, this strategy is alternatively achieved by parameterizing the pendulum mass using polar coordinates. Table 2.3 shows the constraint equations in terms of the Cartesian and polar coordinates. The first row indicates the constraint equation in the position level, whereas second and third rows represent the constraint equations in the velocity and acceleration levels, respectively. By using Cartesian coordinates, the solution consists to find the values of x_p and y_p such that the left-hand side of those equations are satisfied. On the other hand, using polar coordinates by defining $x_p = lc_{\theta}$ and $y_p = ls_{\theta}$, the constraint equations are satisfied for any value of θ .

Table 2.3:	Comparative of the ki	nematic constrain	ts from the	e pendulum	geometry	using
Cartesian an	nd polar coordinates.					
Car	tesian coordinates	nole	r coordina	tes		

	Cartesian coordinates	polar coordinates	
Θ	$x_p^2 + y_p^2 - l^2 = 0$	$(ls_{\theta})^{2} + (lc_{\theta})^{2} - l^{2} = 0$ $l^{2} (s_{\theta}^{2} + c_{\theta}^{2}) - l_{2} = 0$ 0 = 0	
Ġ	$2\left(x_p\dot{x}_p + y_p\dot{y}_p\right) = 0$	$2\left[(ls_{\theta}) l\dot{\theta}c_{\theta} - (lc_{\theta}) l\dot{\theta}s_{\theta} \right] = 0$ $2\left[l^{2}\dot{\theta}s_{\theta}c_{\theta} - l^{2}\dot{\theta}s_{\theta}c_{\theta} \right] = 0$ 0 = 0	
Ö	$2\left(\dot{x}_p^2 + \dot{y}_p^2 + x_p\ddot{x}_p + y_p\ddot{y}_p\right) = 0$	$2\left[l^{2}\dot{\theta}^{2}c_{\theta}^{2} + l^{2}\dot{\theta}^{2}s_{\theta}^{2} + l^{2}\ddot{\theta}c_{\theta}s_{\theta} - l^{2}\dot{\theta}^{2}s_{\theta}^{2}\right] = 0$ $2\left(l^{2}\dot{\theta}^{2} - l^{2}\dot{\theta}^{2}\right) = 0$ $0 = 0$	

Although this illustrative system is described above, the change of coordinates to obtain an independent set of equation usually is not a trivial task. For these cases, the Lagrangian is modified to include the holonomic constraints, such that the associated Euler-Lagrange equations for such system are given by (SALETAN; CROMER, 1970; NETO, 2013)

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{q}_i} \right) - \frac{\partial \mathcal{L}}{\partial q_i} + \frac{\partial \Theta}{\partial q_i}^\top \mathbf{\Lambda} = f_i$$
(2.31)

with $\mathbf{\Lambda} = \{\lambda_1 \ \lambda_2 \ \dots \ \lambda_m\}$ denotes the *m* Lagrange multipliers associated to each constraint equations $\mathbf{\Theta} = \{\Theta_1 \ \Theta_2 \ \dots \ \Theta_m\}^{\top}$. The result is a system of both differential and algebraic

equations (DAE) given by

$$\boldsymbol{M}\ddot{\boldsymbol{q}} + \frac{\partial \boldsymbol{\Theta}}{\partial \boldsymbol{q}}^{\top} \boldsymbol{\Lambda} = \boldsymbol{f}$$

$$\boldsymbol{\Theta} = \boldsymbol{0}$$
(2.32)

The first row from Eq (2.31) corresponds to the *n* second order differential equations to solve for the n + m variables q and Λ . The second row consists of the remaining *m* algebraic equations to be solved simultaneously, which introduces additional challenges to find an accurate solution (BRAUN; GOLDFARB, 2009).

2.6 SINGLE-LIFT DYNAMIC

The EOM from the single-lift system using the reparameterized coordinates is obtained in Silva, Bueno, and Santos (2021) and reproduced herein. Figure 2.6 illustrates the single-lift model, where the payload position is parameterized with spherical coordinates using the rod length l, and angles α and β .



Figure 2.6: Single-lift configuration with an UAV transporting a suspended payload.

The position in terms of the reparameterized coordinates $\tilde{\boldsymbol{q}} = \{x \ y \ z \ \phi \ \theta \ \psi \ \alpha \ \beta\}^{\top}$ is computed through Eq. (2.33).

$$\boldsymbol{r}_{p}\left(\tilde{\boldsymbol{q}}\right) = \begin{cases} x + lc_{\alpha}s_{\beta} \\ y - ls_{\alpha} \\ z + lc_{\alpha}c_{\beta} \end{cases}$$
(2.33)

The equation of motion for from the single-lift system with reparameterized coordinates is given by the following equation

$$\tilde{\boldsymbol{M}}\tilde{\tilde{\boldsymbol{q}}} + \tilde{\boldsymbol{C}}\tilde{\tilde{\boldsymbol{q}}} + \tilde{\boldsymbol{f}}_g = \tilde{\boldsymbol{B}}_u \boldsymbol{u}$$
(2.34)

such that

$$\tilde{\boldsymbol{M}} = \begin{bmatrix} (m+m_p) \, \boldsymbol{I}_{3\times3} & \boldsymbol{0}_{3\times3} & \tilde{\boldsymbol{M}}_r \\ \boldsymbol{0}_{3\times3} & \boldsymbol{M}_\eta & \boldsymbol{0}_{3\times2} \\ \tilde{\boldsymbol{M}}_r^\top & \boldsymbol{0}_{2\times3} & \tilde{\boldsymbol{M}}_p \end{bmatrix} \quad \tilde{\boldsymbol{C}} = \begin{bmatrix} \boldsymbol{0}_{3\times3} & \boldsymbol{0}_{3\times3} & \tilde{\boldsymbol{C}}_r \\ \boldsymbol{0}_{3\times3} & \boldsymbol{C}_\eta & \boldsymbol{0}_{3\times3} \\ \boldsymbol{0}_{2\times3} & \boldsymbol{0}_{2\times3} & \tilde{\boldsymbol{C}}_p \end{bmatrix}$$
(2.35)

$$\tilde{\boldsymbol{M}}_{\boldsymbol{r}} = \begin{bmatrix} -m_p l s_\alpha s_\beta & m_p l c_\alpha c_\beta \\ -m_p l c_\alpha & 0 \\ -m_p l s_\alpha c_\beta & -m_p l c_\alpha s_\beta \end{bmatrix}$$
(2.36)

$$\tilde{\boldsymbol{C}}_{\boldsymbol{r}} = \begin{bmatrix} m_p l(-c_\alpha s_\beta \dot{\alpha} - s_\alpha c_\beta \dot{\beta}) & m_p l(-s_\alpha c_\beta \dot{\alpha} - c_\alpha s_\beta \dot{\beta}) \\ m_p l s_\alpha \dot{\alpha} & 0 \\ m_p l \left(-c_\alpha c_\beta \dot{\alpha} + s_\alpha s_\beta \dot{\beta}\right) & m_p l \left(s_\alpha s_\beta \dot{\alpha} - c_\alpha c_\beta \dot{\beta}\right) \end{bmatrix}$$
(2.37)

$$\tilde{\boldsymbol{M}}_{p} = \begin{bmatrix} m_{p}l^{2} & 0\\ 0 & m_{p}l^{2}c_{\alpha}^{2} \end{bmatrix} \qquad \tilde{\boldsymbol{C}}_{p} = \begin{bmatrix} 0 & m_{p}l^{2}s_{\alpha}c_{\alpha}\dot{\beta}\\ -m_{p}l^{2}s_{\alpha}c_{\alpha}\dot{\beta} & -m_{p}l^{2}s_{\alpha}c_{\alpha}\dot{\alpha} \end{bmatrix}$$
(2.38)

$$\tilde{\boldsymbol{f}}_{g} = \begin{cases} \boldsymbol{0}_{2 \times 1} \\ -(m+m_{p}) g \\ \boldsymbol{0}_{3 \times 1} \\ m_{p} g l s_{\alpha} c_{\beta} \\ m_{p} g l c_{\alpha} s_{\beta} \end{cases}$$
(2.39)
$$\tilde{\boldsymbol{B}}_{u} = \begin{bmatrix} \boldsymbol{B} \\ \boldsymbol{0}_{2 \times 4} \end{bmatrix}$$
(2.40)

Note that in the single-lift with inelastic suspension model, the suspended payload exert no influence on the UAV rotational dynamic. On the other hand, rigidly coupling the payload to the UAV body increases the rotational inertia. A disadvantage is the introduction of a pendulum mode in the system dynamic which is discussed in Section 2.8.

2.7 MULTI-LIFT FORMATION DYNAMIC

Figure 2.7 shows the multi-lift formation in which the *i*th UAV is denoted by \mathcal{V}_i , with $i = 1, \ldots, n$. The system consists of an arbitrary number $n \geq 1$ of UAVs transporting a single suspended payload of mass m_p . Attached to each UAV center of mass there is an inelastic suspension cable (i.e., a no mass rigid rod) of length l_i connecting the payload to each *i*th center of mass.



Figure 2.7: Multi-lift UAV formation with three illustrative UAVs transporting the suspended payload. Note the first, second and *i*th UAVs illustrated.

The inertial reference frame \mathcal{I} is located at the origin O and it is defined by the orthonormal basis $\{\hat{i}, \hat{j}, \hat{k}\}$. A body reference frame \mathcal{B}_i with orthonormal basis $\{\hat{i}_{b,i}, \hat{j}_{b,i}, \hat{i}_{k,i}\}$ attached to each *i*th UAV center of mass O_i and rotating solidarity to the vehicle is defined. The coordinates that unique define the multi-lift configuration is denoted by,

$$\boldsymbol{q}_{f}(t) = \left\{ \boldsymbol{q}_{1}(t)^{\top} \ \boldsymbol{q}_{2}(t)^{\top} \ \dots \ \boldsymbol{q}_{n}(t)^{\top} \ \boldsymbol{q}_{p}(t)^{\top} \right\}^{\top}$$
(2.41)

where the vector $\boldsymbol{q}_i(t) = \{\boldsymbol{r}_i(t)^\top \ \boldsymbol{\eta}_i(t)^\top\}^\top$ corresponds to the position and attitude from the *i*th UAV. Moreover, $\boldsymbol{q}_p(t) = \boldsymbol{r}_p(t) = \{x_p \ y_p \ z_p\}^\top$ denotes the position of the payload with respect to the inertial frame.

Due to the rigid rods assumption, the multi-lift formation coordinates are constrained to satisfy the constant length equation, which is defined for each *i*th rod by the equation $g_i(\mathbf{q}_f) = (x_p - x_i)^2 + (y_p - y_i)^2 + (z_p - z_i)^2 - l_i^2 = 0$. The constraint equation from the *n* rods are combined such that Eq. (2.42) represents the set of holonomic constraints from the multi-lift formation.

$$\boldsymbol{g}_f = \{g_1 \ g_2 \ \dots \ g_n\}^\top = \boldsymbol{0}$$
(2.42)

The dynamics of the multi-lift formation is obtained by using the Euler-Lagrange equations and the holonomic constraints handled using Lagrange multipliers (GOLDSTEIN, 2002). This approach allows one to generalize the equation of motion for an arbitrary number of UAVs. The Lagrangian from the multi-lift formation system is composed of the

kinetic and potential energies from the UAVs and payload.

$$\mathcal{T}_{f} = \frac{1}{2} m_{p} \dot{\boldsymbol{q}}_{p}^{\top} \dot{\boldsymbol{q}}_{p} + \frac{1}{2} \sum_{i=1}^{n} m \dot{\boldsymbol{r}}_{i}^{\top} \dot{\boldsymbol{r}}_{i} + \boldsymbol{\omega}_{i}^{\top} \boldsymbol{J}_{i} \boldsymbol{\omega}_{i}, \qquad (2.43)$$

$$\mathcal{U}_f = -m_p g z_p - \sum_{i=1}^n m g z_i. \tag{2.44}$$

Additionally, there are *n* Lagrange multipliers denoted by $\Lambda_f = \{\lambda_1 \ \lambda_2 \ \dots \ \lambda_n\}$ from the holonomic constraints that are added such that the Euler-Lagrange equations yield the correct equation of motion for the multi-lift system (SALETAN; CROMER, 1970). Then, the Lagrangian from the multi-lift system is given by

$$\mathcal{L}_f = \mathcal{T}_f - \mathcal{U}_f + \boldsymbol{g}_f^{\top} \boldsymbol{\Lambda}_f, \qquad (2.45)$$

Applying the Euler-Lagrange equation for each coordinate, and considering Eq. (2.42), the equation of motion from the multi-lift formation is given by the following system of second order differential and algebraic equations

$$M_f \ddot{\boldsymbol{q}}_f + \boldsymbol{C}_f \dot{\boldsymbol{q}}_f + \boldsymbol{f}_g + \boldsymbol{G}_f^\top \boldsymbol{\Lambda}_f = \boldsymbol{B}_f \boldsymbol{u}_f + \boldsymbol{d}_f$$

$$\boldsymbol{g}_f = \boldsymbol{0}$$
(2.46)

where M_f is the formation inertia, C_f is the Coriolis matrix, f_g is the gravitational force vector. The matrix $G_f = \frac{\partial g_i}{\partial q_f}$ is the Jacobian from the constraint equations and the term $G_f^{\top} \Lambda_f$ corresponds to the constraint forces. B_f is the formation input matrix and u_f is the input vector for all UAVs. In addition, d_f represents the disturbance force vector acting on the system. The matrices from Eq. (2.46) are computed as

$$\boldsymbol{M}_f = \operatorname{diag}\left(\boldsymbol{M}_1 \; \boldsymbol{M}_2 \; \dots \; \boldsymbol{M}_n \; \boldsymbol{M}_p\right) \tag{2.47}$$

$$\boldsymbol{C}_f = \operatorname{diag}\left(\boldsymbol{C}_1 \ \boldsymbol{C}_2 \ \dots \ \boldsymbol{C}_n \ \boldsymbol{0}_{3\times 3}\right) \tag{2.48}$$

$$\boldsymbol{B}_f = \operatorname{diag}\left(\boldsymbol{B}_1 \ \boldsymbol{B}_2 \ \dots \ \boldsymbol{B}_n\right) \tag{2.49}$$

$$\boldsymbol{f}_{g} = \left\{ \boldsymbol{f}_{g,1}^{\top} \ \boldsymbol{f}_{g,2}^{\top} \ \dots \ \boldsymbol{f}_{g,n}^{\top} \ \boldsymbol{f}_{g,p}^{\top} \right\}^{\top}, \qquad (2.50)$$

$$\boldsymbol{u}_f = \left\{ \boldsymbol{u}_1^\top \ \boldsymbol{u}_2^\top \ \dots \ \boldsymbol{u}_n^\top \right\}^\top$$
(2.51)

$$\boldsymbol{d}_{f} = \left\{ \boldsymbol{d}_{1}^{\top} \ \boldsymbol{d}_{2}^{\top} \ \dots \ \boldsymbol{d}_{n}^{\top} \ \boldsymbol{d}_{p}^{\top} \right\}^{\top}$$
(2.52)

The formation gravitational forces from the *i*th UAV and the payload are computed by $\mathbf{f}_{g,i} = -m_i g \hat{\mathbf{k}}$ and $\mathbf{f}_{g,p} = -m_p g \hat{\mathbf{k}}$, respectively. Table 2.4 summarizes the dimensions of these involved matrices.

matrix	dimension
M_{f}	$6n + 3 \times 6n + 3$
C_f	$6n + 3 \times 6n + 3$
G_{f}	$n \times 6n + 3$
$oldsymbol{B}_{f}$	$6n + 3 \times 4n$
$oldsymbol{f}_g$	6n + 3
$oldsymbol{u}_f$	4n
$oldsymbol{d}_f$	6n + 3

Table 2.4: Dimensions of the matrices and vectors for the multi-lift equation of motion.

2.7.1 Aerodynamic Drag Model

The disturbance force vector acting on the payload is included in the model from Eq. (2.46) through $\boldsymbol{d}_p = \{d_{p,x} \ d_{p,y} \ d_{p,z}\}^{\top}$. The drag force due to for the wind resistance on the payload is given by (ANDERSON, 1991)

$$\boldsymbol{d}_p = -c_{pd} \dot{\boldsymbol{r}}_p | \dot{\boldsymbol{r}}_p |, \qquad (2.53)$$

where c_{pd} is the coefficient of drag, whose value is given in terms of the Reynolds number and the shape from the payload. The Reynolds number is an adimensional parameters which expresses the ratio between inertial and viscous forces, and it is given by (ANDERSON, 1991, p.32)

$$Re = \frac{\rho_{\infty} v_{\infty} l_{ref}}{\mu_{\infty}}, \qquad (2.54)$$

where ρ_{∞} is the specific mass from the fluid, v_{∞} is the freestream velocity, μ the dynamic viscosity from the fluid, and l_{ref} a reference dimension, typically the chord for wings or the diameter in the case of a sphere.

2.7.2 Gust Model

The 1-cosine gust model presented in Eq. 2.55 is a discrete gust for evaluating the impact of atmospheric turbulence on the system dynamics. It is parameterized by the gust gradient distance H, a reference gust velocity U_{ref} and the parameter r representing the position with respect to the gust. The gust acts either vertically or laterally and it is consist of a pulse gradually increasing from 0 to U_{ref} at the center of the gust. Figure 2.8 illustrates the normalized gust profile acting vertically in a position $(x_{go} y_{go})$ on the grid. The model is incorporated to the equation of motion by defining 1-cosine gust in specific positions on the environment. Considering a gust centered at $(x_{go} y_{go})$ and a distance $r = \sqrt{(x - x_{go})^2 + (y - y_{go})^2}$ from the gust origin, then the following equation model the gust velocity

$$\frac{U}{U_{ref}} = \begin{cases} \frac{1}{2} \left[1 + \cos\left(\frac{\pi r}{H}\right) \right] & \text{for } r \le H \\ 0 & \text{for } r > H \end{cases}$$
(2.55)



Figure 2.8: Representation from the normalized 1-cosine gust equation at $(x_0 y_0)$.

2.7.3 Index Reduction

The DAE system requires an especial strategy to be solved (BRAUN; GOLDFARB, 2009). The algebraic constraint imposes a restriction on the configuration manifold and violate this constraint implies changing the length of the suspension cables. A typical strategy consists of differentiating the algebraic equations to change the problem from solving a system of DAE to solve a set of ODEs, which is known in the literature as an index reduction. However, the resulting system is mildly numerically unstable and an accurate solution requires the use of a constraint stabilization technique (BAUMGARTE, 1972).

Differentiating the constraint equation twice in respect to the time t result in the constraint equations on the acceleration level (ASCHER et al., 1995), i.e.,

$$\frac{d\boldsymbol{g}_f(\boldsymbol{q}_f)}{dt} = \frac{d\boldsymbol{g}_f}{d\boldsymbol{q}_f}\frac{d\boldsymbol{q}_f}{dt} = \boldsymbol{G}_f \dot{\boldsymbol{q}}_f = \boldsymbol{0}, \qquad (2.56)$$

$$\frac{d^2 \boldsymbol{g}_f(\boldsymbol{q}_f)}{dt^2} = \boldsymbol{G}_f \ddot{\boldsymbol{q}}_f + \dot{\boldsymbol{G}}_f \dot{\boldsymbol{q}}_f = \boldsymbol{0}.$$
(2.57)

Then, considering $f_f = B_f u_f + d_f - f_g - C_f \dot{q}_f$, the resulting model is the following system of differential equations

$$\begin{bmatrix} \boldsymbol{M}_{f} & \boldsymbol{G}_{f}^{\mathsf{T}} \\ \boldsymbol{G}_{f} & \boldsymbol{0} \end{bmatrix} \begin{bmatrix} \ddot{\boldsymbol{q}}_{f} \\ \boldsymbol{\Lambda}_{f} \end{bmatrix} = \begin{bmatrix} \boldsymbol{f}_{f} \\ -\dot{\boldsymbol{G}}_{f} \dot{\boldsymbol{q}}_{f} \end{bmatrix}, \qquad (2.58)$$

2.7.4 Constraint Stabilization

Equation (2.58) usually is numerically unstable. This numerical issue occurs since the problem of solving the algebraic equations is transformed in an equivalent equation in terms of acceleration $\ddot{g}_f = 0$. However, numerical approaches commonly produce residuals that are integrated and accumulated on the velocity and position levels, resulting in the problem know as constraint drift (MASARATI, 2011), which is violation from the constraints equation on position and velocity levels, i.e., Eq. (2.42) and Eq. (2.56) respectively.

Baumgarte (1972) suggests the addition of the numerically stabilizing terms $-2\alpha_f \dot{g}_f$ and $-\beta_f^2 g_f$ to Eq. (2.57) to modify the constraint hypersurface to an attractor. Bisgaard, Bendtsen, and Cour-Harbo (2009) interpret this modification as an artificial introduction of spring and dampers in parallel with the rigid rods. This method is illustrated by Figure 2.9, where the non-stabilized solution diverges from the exact solution over time, whereas the stabilized case, the solution remains close to accurate value. However, note



Figure 2.9: Illustration from the constraint drift and the stabilization method.

that both new terms are equal to zero. Then, the model from the multi-lift including the stabilizing terms assumes the following form

$$\begin{bmatrix} \boldsymbol{M}_{f} & \boldsymbol{G}_{f}^{\mathsf{T}} \\ \boldsymbol{G}_{f} & \boldsymbol{0} \end{bmatrix} \begin{pmatrix} \ddot{\boldsymbol{q}}_{f} \\ \boldsymbol{\Lambda}_{f} \end{pmatrix} = \begin{pmatrix} \boldsymbol{f}_{f} \\ -\dot{\boldsymbol{G}}_{f} \dot{\boldsymbol{q}}_{f} - 2\alpha_{f} \dot{\boldsymbol{g}}_{f} - \beta_{f}^{2} \boldsymbol{g}_{f} \end{pmatrix}.$$
 (2.59)

Then, the Lagrange multiplier with the numerically stabilizing terms can be written by

$$\boldsymbol{\Lambda}_{f} = -\left(\boldsymbol{G}_{f}\boldsymbol{M}_{f}^{-1}\boldsymbol{G}_{f}^{\top}\right)^{-1}\left(-\dot{\boldsymbol{G}}_{f}\dot{\boldsymbol{q}}_{f} - 2\alpha_{f}\dot{\boldsymbol{g}}_{f} - \beta_{f}^{2}\boldsymbol{g}_{f} - \boldsymbol{G}_{f}\boldsymbol{M}_{f}^{-1}\boldsymbol{f}_{f}\right), \qquad (2.60)$$

The matrix $G_f M_f^{-1} G_f^{\top}$ is invertible since rank $(G_f) = n, \forall q_f | g_f = 0$, and this property is demonstrated by the following Theorem.

Theorem 1. the constraint Jacobian G_f has rank n for all q_f such that $g_f(q_f) = 0$.

Proof. The rows from G_f are linearly independent if, and only if, the trivial solution $\alpha_1 = \alpha_2 = \cdots = \alpha_n = 0$ is the unique solution to this following equation

$$\alpha_1 \frac{dg_1}{d\boldsymbol{q}_f} + \alpha_2 \frac{dg_2}{d\boldsymbol{q}_f} + \dots + \alpha_n \frac{dg_n}{d\boldsymbol{q}_f} = 0.$$
(2.61)

The equation from the *i*th rod depends only on the coordinates from the *i*th UAV and payload, i.e., $g_i \equiv g_i(\mathbf{q}_i \ \mathbf{q}_p)$, such that the partial differentiation with respect to the coordinates from the other UAVs is zero, i.e.,

$$\frac{\partial g_i}{\partial \boldsymbol{q}_j} = \boldsymbol{0}, \forall \ i \neq j.$$
(2.62)

This implies that the *i*th row $\frac{\partial g_i}{\partial q_i}$ from G_f can not be expressed as a linear combination of $\frac{g_j}{\partial q_i}$, $i \neq j$. Then, the trivial solution is the only possible solution for Eq. (2.61). An

additional condition is that the rows from G_f do not correspond to be the null-vector, i.e., $\frac{dg_i}{dq_f} \neq \mathbf{0}, i = 1 \ 2 \ \dots \ n$. Consider, by contradiction that

$$\frac{dg_i(\boldsymbol{q}_f)}{d\boldsymbol{q}_f} = \mathbf{0},\tag{2.63}$$

then, if Eq. (2.63) is satisfied, the dot product is such that

$$\frac{dg_i(\boldsymbol{q}_f)}{d\boldsymbol{q}_f} \frac{dg_i(\boldsymbol{q}_f)}{d\boldsymbol{q}_f}^{\top} = 0.$$
(2.64)

Writing

$$\frac{dg_i(\boldsymbol{q}_f)}{d\boldsymbol{q}_f} = \begin{bmatrix} \frac{\partial g_i}{\partial \boldsymbol{q}_1} & \frac{\partial g_i}{\partial \boldsymbol{q}_2} & \dots & \frac{\partial g_i}{\partial \boldsymbol{q}_n} & \frac{\partial g_i}{\partial \boldsymbol{q}_p} \end{bmatrix}, \qquad (2.65)$$

and from the holonomic constraint written in terms of the corresponding UAV and payload coordinates, i.e., $g_i(\boldsymbol{q}_f) \equiv g_i(\boldsymbol{q}_i, \boldsymbol{q}_p)$, the non-null terms are:

$$\frac{\partial g_i(\boldsymbol{q}_i, \boldsymbol{q}_p)}{\partial \boldsymbol{q}_i} = \begin{bmatrix} -2(x_p - x_i) & -2(y_p - y_i) & -2(z_p - z_i) & 0 & 0 \end{bmatrix}$$
(2.66)

and

$$\frac{\partial g_i(\boldsymbol{q}_i, \boldsymbol{q}_p)}{\partial \boldsymbol{q}_p} = \begin{bmatrix} 2(x_p - x_i) & 2(y_p - y_i) & 2(z_p - z_i) \end{bmatrix}.$$
(2.67)

which allows one to rewrite Eq. (2.64) by

$$\frac{dg_i(\boldsymbol{q}_f)}{d\boldsymbol{q}_f} \frac{dg_i(\boldsymbol{q}_f)}{d\boldsymbol{q}_f}^{\top} = 4\left(x_p - x_i\right)^2 + 4\left(y_p - y_i\right)^2 + 4\left(z_p - z_i\right)^2.$$
(2.68)

Substituting the holonomic constraint from the ith rod result in:

$$\frac{dg_i(\boldsymbol{q}_f)}{d\boldsymbol{q}_f} \frac{dg_i(\boldsymbol{q}_f)}{d\boldsymbol{q}_f}^{\top} = 4l_i^2, \qquad (2.69)$$

which is zero only if $l_i = 0$. However, $l_i \neq 0$, and then, the rows $G_f(q_f)$ are non null for all q_f that satisfies $g_f(q_f) = 0$, which demonstrates that the matrix G_f is invertible quod erat demonstrandum.

Substituting the stabilized Lagrange multiplier from Eq.(2.60) in the first row of Eq. (2.58), and defining the variable $v_f = \dot{q}_f$ leads to a system of first order differential equations:

$$\dot{\boldsymbol{q}}_{f} = \boldsymbol{v}_{f}$$

$$\dot{\boldsymbol{v}}_{f} = \boldsymbol{M}_{f}^{-1} \left(\boldsymbol{f}_{f} + \boldsymbol{G}_{f}^{\top} \left(\boldsymbol{G}_{f} \boldsymbol{M}_{f}^{-1} \boldsymbol{G}_{f}^{\top} \right)^{-1} \left(- \dot{\boldsymbol{G}}_{f} \dot{\boldsymbol{q}}_{f} - \boldsymbol{G}_{f} \boldsymbol{M}_{f}^{-1} \boldsymbol{f}_{f} - 2\alpha_{f} \dot{\boldsymbol{g}}_{f} - \beta_{f}^{2} \boldsymbol{g}_{f} \right) \right)$$
(2.70)

which represents the EOM for the multi-lift formation. Integrating Eq. (2.70) using a numerical method, such as the Runge-Kutta method, the trajectories of the multi-lift formation system can be obtained over time.

2.7.5 Algorithm

Algorithm 1 summarizes the procedure for obtaining the trajectories from the multi-lift system. The algorithm consists of defining the number n of UAVs in the formation and their initial configuration, such that: the UAVs are on the constraint manifold. Then, the state from the formation at instant $(t + \Delta t)$ is obtained by integrating Eq. (2.70) using the Runge-Kutta algorithm. The process is repeated until t achieves t_{max} .

Algorithm 1 Compute the trajectories from the multi-lift system

1: $n \leftarrow \mathbb{Z}^+$. 2: define the acceptable error in position ϵ_{g_f} and velocity $\epsilon_{\dot{g}_f}$ levels. 3: define the simulation and stabilization parameters t_{max} , Δt , α_f , and β_f . 4: define the initial system configuration $q_f(0)$ satisfying $q_f(q_f) = 0$ and their derivatives according to Eq. (2.42), Eq. (2.56), Eq. (2.57). 5: *main loop*: 6: for $t \leq t_{max}$ do $\boldsymbol{q}_f(t + \Delta t) \leftarrow integrate \ Eq. \ (2.70) \ (\boldsymbol{q}_f(t), \boldsymbol{u}_f(t), n)$ 7: $\boldsymbol{u}_f(t + \Delta t) \leftarrow$ update the control signal. 8: $t \leftarrow t + \Delta t$. 9: 10: check if the constraints are satisfied for all t. 11: if $|\boldsymbol{g}_f(t)| < \epsilon_{\boldsymbol{g}_f}$ and $|\dot{\boldsymbol{g}}_f(t)| < \epsilon_{\dot{\boldsymbol{g}}_f}$ then accept result and plot trajectories. 12:13: else redefine the simulation and stabilization parameters. 14:goto main loop. 15:

2.7.6 Null Space Formulation

The null-space from G_f , i.e., ker (G_f) is one subspace that maps any vector from it to the null vector. If one select a matrix N_f whose columns forms the basis from the null-space of G_f , then the following equation holds

$$\boldsymbol{G}_f \boldsymbol{N}_f = \boldsymbol{0} \tag{2.71}$$

where N_f is a $6n + 3 \times 5n + 3$ matrix that spans the null-space from G_f . Applying the null-space in the EOM from the multi-lift dynamics (Eq.(2.46)), the result is the projection such that the constraint forces are zero, as represented by the following system of equations

$$\boldsymbol{N}_{f}^{\top}\boldsymbol{M}_{f}\boldsymbol{\ddot{q}}_{f} = \boldsymbol{N}_{f}^{\top}\boldsymbol{f}_{f}$$

$$(2.72)$$

2.8 FORMATION MODAL ANALYSIS

The geometric coupling of the multi-lift formation implies to a dynamics for which an equivalent modal analysis can be performed. This analysis can be carried out for the single-lift configuration, since the coordinates are reparameterized analytically and the EOM linearized. However, considering the multi-lift system from Eq. (2.46), there are n constraints, such that only (5n + 3) from those (6n + 3) coordinates are independent from each other. This increases the complexity since a choice of reparameterization is not trivial.

Yang (1992) noted that a linear system under linear and time-invariant constraints is stiffer than the unconstrained system. For the nonlinear constraints, such as Eq. (2.57), a stable and linear system remains stable for small vibrations around a equilibrium condition. In addition, the constraints introduce additional excitation sources and modify the disturbances forces, such that their effects on the system dynamics must be evaluated. Berrios et al. (2014) present such analysis for the twin-lift system and find unstable formation modes.

Sohoni and Whitesell (1986) linearizes the EOM around an operating point and assumes a linear and time-invariant map from the independent and dependent coordinates. Then, they eliminate both the dependent coordinates and Lagrange multipliers to analyze the modal properties. Liang and Lance (1987) use the Gram-Schmidt algorithm to generate an orthonormal basis using the constraint Jacobian such that both the null-space and its time-derivative are efficiently obtained. They obtain the EOM using the velocity transformation in terms of the independent coordinates which allows one to compute the trajectories without using a constraint stabilization method González et al. (2017). The main disadvantage from using independent coordinates is the extra-step required to solve the non-linear algebraic equations to find the dependent coordinates at each time-step, which is prone to numerical convergence issues.

2.8.1 Single-Lift Modal

The linearized EOM for the single-lift system with independent coordinates is obtained in Silva, Bueno, and Santos (2021). The modal properties from the single-lift system are obtained from the eigenvalues of matrix \boldsymbol{A} defined by

$$oldsymbol{A} = egin{bmatrix} oldsymbol{0}_{8 imes 8} & oldsymbol{I}_{8 imes 8} \ oldsymbol{A}_p & oldsymbol{0}_{8 imes 8} \end{bmatrix}$$

where

Two of the eigenvalues from \mathbf{A} are identical purely imaginary pairs $\lambda_1 = \lambda_2 = \pm 2.13j$ and they correspond to the pendulum modes from the single-lift system with a period of 2.94 s. The coordinate vector $\mathbf{q}_e = \{x \ y \ z \ 0 \ 0 \ \psi \ x \ y \ z + l\}^{\top}$ is a configuration in equilibrium as illustrated in Fig. 2.10.



Figure 2.10: Schematic representation of the equilibrium configuration for the single-lift.

2.8.2 Twin-Lift Modal

The formation modal analysis for the twin-lift configuration is realized by identifying a subset of dependent coordinates from the formation coordinates, i.e., q_f , such that the constraint Jacobian with respect to these dependent coordinates has full rank. For the twin-lift system, the altitude from the leader and follower vehicles are choosen as dependent coordinates. The multi-lift equilibrium configuration is the solution from Eq. 2.46 with $\dot{q}_f = 0$

$$\begin{aligned} \boldsymbol{f}_g + \boldsymbol{G}_f^\top \boldsymbol{\Lambda}_f = & \boldsymbol{B}_f \boldsymbol{u}_f \\ \boldsymbol{g}_f = & \boldsymbol{0} \end{aligned} \tag{2.73}$$

For the twin-lift system, the equilibrium presented in Figure 2.11 is a solution.



Figure 2.11: Schematic representation of the equilibrium configuration for the twin-lift.

The dynamic equations from the twin-lift system are linearized with respect to the equilibrium and the modal properties obtained. Figure 2.12 shows the eigenvalues from

the linearized twin-lift configuration.



Figure 2.12: Eigenvalues from the twin-lift configuration.

Chapter 3 CONTROL OF UAV FORMATIONS

The use of electronic controllers is fundamental for multirotors UAVs due to their unstable dynamics, nonlinearities, and underactuated system configuration, which render this type of vehicle challenging for piloting without computer assistance (EMRAN; NAJJARAN, 2018).

A common requirement from electronic controllers is a reference model from the plant to be controlled, which contains the information in the form of mathematical equations or nonparametric functions to predict the future states depending on the applied control.

An assumption usually employed in literature is that the model is time-invariant and the differences in relation to its real dynamic are regarded as disturbances, which are classified into two categories: a) internal disturbances from unmodeled dynamics (e.g., simplifications of aerodynamics (SANZ et al., 2016), time delays in control (FRIDMAN; SEURET; RICHARD, 2004), and model uncertainties (LU; REN; PARAMESWARAN, 2020)); b) external disturbances due to different other sources, such as due to wind gusts (GUO et al., 2020).

In addition to the disturbances that a single UAV is subjected to, in aerial transportation mission using a multi-lift UAV formation, the coupled payload introduces additional forces on the system, since the UAV needs to provide extra thrust to balance the constraint force, which may saturate the UAV actuators.

This chapter presents a discussion from control of UAVs, with reference to the control literature in Section 3.1. Section 3.2 introduces the developed strategy to control a multi-lift UAV formation.

3.1 RELATED WORK

The PID controller is maybe the first controller that one encounters while researching for control of UAVs. This occurs due to its extensive use both in the hobbyist and scientific communities due to its ease of implementation, low computational requirements, and simple understanding from the controller parameters. To design a PID controller, the multirotor UAV dynamic is linearized in a hover condition, then one typically choose the PID gains such that the closed-loop error dynamic from the linearized model is asymptotically stable. Sliding Mode Control is also a strategy present in diverse literature to control UAVs. One of its features is their excellent capability of rejecting disturbances by rapidly switching the actuator to stabilize the system. However, the switching also introduces high frequency vibrations on the UAV which is denominated as "chattering".

Artificial Intelligence (AI) is also in research to overcome the limitation from classical controllers, that include the dependence of models and online adapting to a changing environment. Waslander et al. (2005) model the UAV dynamic using input and output data with a non-parametric model. Then, they optimize the controller gains by adding a Gaussian noise to it in each iteration. Hwangbo et al. (2017) uses a neural network as the controller whose gains are updated through reinforcement learning.

Due to the underactuated characteristics of multirotor UAV, the control is designed in a cascade fashion, where an inner-loop is responsible of stabilizing the actuated degrees of freedom, and an outer-loop responsible of finding adequate set-points to allow the vehicle track the underactuated degrees of freedom. Raffo, Ortega, and Rubio (2011) and Brandão, Filho, and Carelli (2013) uses the partial-feedback linearization to design an inner-loop controller (SPONG, 1994).

3.1.1 Propulsion Saturation

Multirrotor UAVs commonly uses multiple BLDC (brushless DC) eletric motors for both propulsion and control. The strategy is to keep the mechanical simplicity and reduce overall costs. The motor uses an electronic controller, such as an ESC (Electronic Speed Controller), to switch the DC voltage from the batteries, resulting in a changing magnetic field that the permanent magnets follows. Then, the motor speed can be controlled by conditioning the signal sent to the ESC.

3.2 FORMATION CONTROL

Formation control consists of coordinating the motion from each UAV to achieve a common goal. In the multi-lift transportation mission, the proximity from each vehicle also introduces a collision hazard that requires implementation to detect and avoid capability. There are different ways to achieve this goal, which requires that an UAV acknowledge the position from its peers by: a) using the onboard cameras to estimate the obstacle path, then plan a route to avoid a collision (PEDRO et al., 2021); b) creating a communication network to share the information from the inertial navigation system; c) using a dedicated sensor such as ultrasonic or lidar (TAHIR et al., 2019).

The individuals position are known over time, and they are used to develop a formation controller based on leader-follower strategy. The formation control consists of considering two types of UAVs: i) the leader, which is responsible of communicating with ground

control station and tracking the desired mission path, and *ii*) the followers, which are responsible to establish the formation geometry. The leader UAV has its control structure unaltered such that the stability and robustness properties are preserved. The followers UAVs have their outer-loop modified to use the leader position and determine their attitude set-points based on their current position within the formation.

The desired formation geometry is defined using virtual constraints. If a distance d from the *i*th to the *j*th UAVs is desired, the corresponding constraint equation is simply the Euclidian distance given by

$$(x_i - x_j)^2 + (y_i - y_j)^2 + (z_i - z_j)^2 - d^2 = 0.$$
(3.1)

If a linear separation in the \hat{i} direction of d_x units is desired, then the equation is

$$(x_i - x_j) - d_x = 0. (3.2)$$

Similarly, for the \hat{j} and \hat{k} directions, the constraint equations are given by the following equations

$$(y_i - y_j) - d_y = 0, (3.3)$$

$$(z_i - z_j) - d_z = 0. (3.4)$$

Table 3.1 presents different formation configurations and their corresponding virtual constraints. The first row presents the linear separation among the vehicles such that they keep the distance in both the \hat{i} and \hat{j} axes. In the second row of the table, the UAVs are separated from each other by the Euclidian distance. The "V" and the circular formation geometries are obtained by combining the constraint equations such that every vehicle is constrained in the formation. The "V" formation is parameterized through the vertex angle 2γ and the distance d between UAVs.

The formation control consists of keeping each UAV in its position, such that the virtual constraint equations are satisfied. If the virtual constraints exist, then a constraint force is considered to keep each UAV on its desired position according to the formation. Since there is no such force, it is artificially introduced by the followers, by projecting the thrust such that its components provide the computed virtual constraint forces. The virtual constraint equations are grouped in the following virtual constraints vector

$$\boldsymbol{g}_{vc} = \boldsymbol{g}_{vc}(\boldsymbol{r}_f) = \boldsymbol{0}, \tag{3.5}$$

which is written in terms of the formation configuration, described by the formation position vector $\mathbf{r}_f = \{\mathbf{r}_1 \ \mathbf{r}_2 \ \dots \ \mathbf{r}_n\}^{\top}$. This equation is linear in terms of the coordinates in "V" formations, and nonlinear for the circular formation.

Figure 3.1 illustrates the formation controller. The formation dynamics is used in the integrator block, which computes the future configuration based on the previous control. This future state is validated to check if the constraints are approximately satisfied within

a tolerance, if not, the solution is interrupted for adjusting the constraint stabilization parameters. Then, the information is processed and the state from the leader is updated. The leader is updated with the mission trajectory, then, it can compute its control for tracking the desired trajectory. The formation controller receives the leader state, then, the followers can compute their control through the virtual constraints. The control from the leader and the followers are validated through the saturation block. The saturation is applied if a control that result in angular velocity outside the operational range from the multirotors, . The saturated formation control is informed to the integration block and the process is repeated.



Figure 3.1: Diagram illustrating the process for the obtaining the formation trajectories.

representation	geometry	virtual constraints
$d_y \underbrace{ \begin{array}{c} & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_2 \\ & \mathcal{V}_1 \\ & \mathcal{V}_2 \\ & \mathcal{V}_2$	linear separation	$\begin{aligned} x_1 - x_2 - d_x \\ y_1 - y_2 - d_y \\ z_1 - z_2 - d_z \end{aligned}$
\mathcal{V}_2 \mathcal{V}_1	Euclidian distance	$(x_1 - x_2)^2 + (y_1 - y_2)^2 + (z_1 - z_2)^2 - d$
$\begin{array}{c} \begin{array}{c} \mathcal{V}_2 \\ \mathcal{V}_1 \\ \mathcal{V}_1 \\ \mathcal{V}_2 \\ \mathcal{V}_3 \\ \mathcal{V}_3 \end{array}$	"V"	$ \begin{array}{c} x_1 - x_2 - dc_{\gamma} \\ y_1 - y_2 - ds_{\gamma} \\ z_1 - z_2 \\ x_1 - x_3 - dc_{\gamma} \\ y_1 - y_3 + ds_{\gamma} \\ z_1 - z_3 \end{array} $
\mathcal{V}_3 d $+$ \mathcal{V}_2 $'$ \mathcal{V}_2	circular	$(x_1 - x_2)^2 + (y_1 - y_2)^2 + (z_1 - z_2)^2 - d$ $(x_1 - x_3)^2 + (y_1 - y_3)^2 + (z_1 - z_3)^2 - d$ $(x_2 - x_3)^2 + (y_2 - y_3)^2 + (z_2 - z_3)^2 - d$

Table 3.1: Different formation geometries and their corresponding virtual constraint equations.

3.2.1 "V" formation

The geometry of the desired "V" formation is obtained by setting the coefficients v_{1i} using the geometry parameters of the vertex angle 2γ and the desired distance d between each two UAVs on the "V", such as illustrated in Fig. 3.2. The "V-formation" geometry is obtained by considering a linear separation of v_{1j} units from \mathcal{V}_i to \mathcal{V}_1 , such that the following system of equations is obtained

$$\boldsymbol{g}_{vc} = \boldsymbol{J}_{vc} \boldsymbol{r}_f + \boldsymbol{v}_c \tag{3.6}$$

where $\boldsymbol{g}_{vc} = \{\boldsymbol{g}_{1i} \dots \boldsymbol{g}_{1n}\}^{\top}, i = 2, 3, \dots, n$, denotes virtual constraint equations relating the desired position from each *i*th follower to the leader. Vector $\boldsymbol{v}_c = \{\boldsymbol{v}_{12}^{\top} \dots \boldsymbol{v}_{1n}^{\top}\}^{\top}, i = 2, 3, \dots, n$, introduces the desired distance from the follower to the leader, which for the "V" formation it is parameterized by $\boldsymbol{v}_{1i} = \frac{i}{2} \{-dc_{\gamma} - ds_{\gamma} \ 0\}^{\top}$ for an even number of vehicles, and $\boldsymbol{v}_{1i} = \frac{i-1}{2} \{-dc_{\gamma} \ ds_{\gamma} \ 0\}^{\top}$ otherwise.



Figure 3.2: Schematic representation of the "V" formation with a vertex angle 2γ and the relative distance d for every two consecutive UAVs represented by \mathcal{V}_i , $i = 1, \ldots, 5$.

$$\boldsymbol{J}_{vc} = \begin{bmatrix} \boldsymbol{I}_{3\times3} & -\boldsymbol{I}_{3\times3} & \boldsymbol{0}_{3\times3} & \dots & \boldsymbol{0}_{3\times3} \\ \boldsymbol{I}_{3\times3} & \boldsymbol{0}_{3\times3} & -\boldsymbol{I}_{3\times3} & \dots & \boldsymbol{0}_{3\times3} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \boldsymbol{I}_{3\times3} & \boldsymbol{0}_{3\times3} & \boldsymbol{0} & \dots & -\boldsymbol{I}_{3\times3} \end{bmatrix}.$$
(3.7)

Differentiating twice Eq. (3.6) with respect to the time, the equation obtained is $J_{vc}\ddot{\boldsymbol{r}}_f = \mathbf{0}$. The terms $2\alpha_c (J_{vc}\dot{\boldsymbol{r}}_f)$ and $\beta_c^2 (J_{vc}\boldsymbol{r}_f + \boldsymbol{v}_c)$, such that $\alpha_c > 0$ and $\beta_c > 0$, are added to stabilize the controller (TARRAF; ASADA, 2002).

$$\boldsymbol{J}_{vc}\ddot{\boldsymbol{r}}_{f} + 2\alpha_{c}\left(\boldsymbol{J}_{vc}\dot{\boldsymbol{r}}_{f}\right) + \beta_{c}^{2}\left(\boldsymbol{J}_{vc}\boldsymbol{r}_{f} + \boldsymbol{v}_{c}\right) = \boldsymbol{0}$$
(3.8)

The translational dynamics from the UAV formation is given by

$$\boldsymbol{M}_{f,r}\ddot{\boldsymbol{r}}_f + \boldsymbol{f}_{f,r} + \boldsymbol{J}_{vc}^{\top}\boldsymbol{\Lambda}_c = \boldsymbol{u}_{f,r}, \qquad (3.9)$$

where $M_{f,r} = \text{diag}(m_1 I_{3\times 3} \ m_2 I_{3\times 3} \ \dots \ m_n I_{3\times 3})$ and $f_{f,r} = \{f_{1,r}^{\top} \ f_{2,r}^{\top} \ \dots \ f_{n,r}^{\top}\}^{\top}$ with $f_{i,r} = -m_i g \hat{k}$, and $u_{f,r} = \{u_{1,r} \ \dots \ u_{n,r}\}^{\top}$ with $u_{i,r} = B_{r,i} u_{T,i}$ denoting the lift from the *i*th UAV projected in the *x*, *y* and *z* directions. The formation acceleration is substituted in Eq. (3.8) resulting the following equations

$$\boldsymbol{J}_{vc}\boldsymbol{M}_{f,\boldsymbol{r}}^{-1}\left(\boldsymbol{u}_{f,\boldsymbol{r}}-\boldsymbol{f}_{f,\boldsymbol{r}}-\boldsymbol{J}_{c}^{\top}\boldsymbol{\Lambda}_{c}\right)+2\alpha_{c}\left(\boldsymbol{J}_{vc}\dot{\boldsymbol{r}}_{f}\right)+\beta_{c}^{2}\left(\boldsymbol{J}_{vc}\boldsymbol{r}_{f}+\boldsymbol{v}_{c}\right)=\boldsymbol{0}$$
(3.10)

$$-\left(\boldsymbol{J}_{vc}\boldsymbol{M}_{f,\boldsymbol{r}}^{-1}\boldsymbol{J}_{vc}^{\top}\right)\boldsymbol{\Lambda}_{c} = -\boldsymbol{J}_{vc}\boldsymbol{M}_{c}^{-1}\left(\boldsymbol{u}_{f,\boldsymbol{r}}-\boldsymbol{f}_{f,\boldsymbol{r}}\right) - 2\alpha_{c}\boldsymbol{J}_{vc}\dot{\boldsymbol{r}}_{f} - \beta_{c}^{2}\left(\boldsymbol{J}_{vc}\boldsymbol{r}_{f}-\boldsymbol{c}_{c}\right)$$
(3.11)

$$\boldsymbol{\Lambda}_{c} = \left(\boldsymbol{J}_{vc}\boldsymbol{M}_{f,\boldsymbol{r}}^{-1}\boldsymbol{J}_{vc}^{\top}\right)^{-1} \left(\boldsymbol{J}_{vc}\boldsymbol{M}_{c}^{-1}\left(\boldsymbol{u}_{f,\boldsymbol{r}} - \boldsymbol{f}_{f,\boldsymbol{r}}\right) + 2\alpha_{c}\boldsymbol{J}_{vc}\dot{\boldsymbol{r}}_{f} + \beta_{c}^{2}\left(\boldsymbol{J}_{vc}\boldsymbol{r}_{f} - \boldsymbol{c}_{c}\right)\right). \quad (3.12)$$

The term $\boldsymbol{u}_{v_f} = -\boldsymbol{J}_{vc}^{\top} \boldsymbol{\Lambda}_c = \left\{ \boldsymbol{u}_{1,\boldsymbol{v}_f}^{\top} \ \boldsymbol{u}_{2,\boldsymbol{v}_f}^{\top} \ \dots \ \boldsymbol{u}_{n,\boldsymbol{v}_f}^{\top} \right\}^{\top}$ from translational dynamics acts as the virtual forces and they are artificially introduced by the followers to keep the UAVs

on the formation constraint hypersurface defined by Eq.(3.6). The virtual force developed by the *i*th UAV is written in terms of its components in the *x*, *y* and *z* directions, such that $\boldsymbol{u}_{i,\boldsymbol{v}_f} = \{u_{i,x,\boldsymbol{v}_f} \ u_{i,y,\boldsymbol{v}_f} \ u_{i,z,\boldsymbol{v}_f}\}^{\top}$. Due to the underactuated configuration from multirotors UAVs, these computed virtual forces are obtained by considering the following attitude set-points:

$$\phi_{d,i} = \sin^{-1} \left(\frac{u_{i,x,v_f} s_{\psi,i} - u_{i,y,v_f} c_{\psi,i}}{u_{T,i}} \right), \ \theta_{d,i} = \sin^{-1} \left(\frac{u_{i,x,v_f} c_{\psi,i} + u_{i,y,v_f} s_{\psi,i}}{u_{T,i} c_{\phi_{d,i}}} \right).$$
(3.13)

These attitude set-points in addition to the leader altitude $z_{d,i} = z_1$ are then transmitted to the inner-loop controller.

Chapter 4 RESULTS AND DISCUSSION

This chapter presents the results to discuss the introduced approach. Section 4.1 presents a time-domain analysis from the trajectories of the single-lift system using the method of independent coordinates and the method of Lagrange multipliers for constrained systems. Section 4.2 presents the parameters employed to fully define the mission environment. Section 4.3 presents numerical data from the twin-lift configuration. Section 4.4 presents the trajectories from the multi-lift system, and Section 4.5 introduces a comparative study.

4.1 SINGLE-LIFT

The proposed methodology employed for solving the system of constrained differential equations is evaluated in the time-domain by integrating the equation of motion, by comparing the trajectories obtained using the methods of independent coordinates and Lagrange multipliers. The simulations start with identical initial conditions and a constant thrust is imposed to statically balance the combined UAV and payload weight. This specific control input is motivated to avoid compromising the trajectories due to the closed-loop control. Then, any difference in the results is attributed to the solving methodology.

Table 4.1 presents the physical parameters which define the single-lift system. The trajectories from the single-lift with reparameterized coordinates (see Eq. 2.34) is obtained by integrating the EOM using the 5th order Runge-Kutta algorithm and it is a reference solution used for comparison. The trajectory from the single-lift system using the Lagrange multiplier method is obtained by setting n = 1 and the resulting EOM is solved through the Algorithm 1.

parameter	description	value	unit
m	UAV mass	2.5	kg
m_p	payload mass	0.4	kg
J	inertia tensor	diag $(0.957 \ 0.486 \ 0.355)$	$\rm kg.m^2$
l	suspension length	2.5	m

Table 4.1: Single-lift system parameters.

Figure 4.1 shows the UAV trajectories and a snapshot from the configuration at t = 0 s. The payload is initially placed at $\alpha|_{t=0} = 15^{\circ}$ and $\beta|_{t=0} = 25^{\circ}$ and these initial conditions correspond to $\mathbf{r}_p|_{t=0} = \{1.02 - 0.65 \ 2.19\}^{\top}$ m. The oscillation exhibits a period equal to 2.94 s, corresponding to the eigenvalue obtained from the formation modal analysis.



Figure 4.1: Trajectories from the single-lift system.

Figure 4.2 presents the constraint equation from the single-lift system in position level, i.e., $g = (x - x_p)^2 + (y - y_p)^2 + (z - z_p)^2 - l^2$ for different values of stabilization parameters. The solution starts from a configuration consistent with the imposed constraints (BRAUN; GOLDFARB, 2009). The constraint rapidly increases when computing the non-stabilized case, which is represented by the curve $g(q)|_{\alpha_f=0}$, and then, the resulting trajectory becomes inaccurate. Including the constraint stabilization, represented by the curves $g(q)|_{\alpha_f=20}$ and $g(q)|_{\alpha_f=100}$, the constraint equation is satisfied within $\epsilon_{q_f} < 10^{-1}$.

The UAV and payload positions obtained by integrating the dynamic model using the Euler-Lagrange formulation for constrained system are approximately equal to the positions obtained by using reparameterized coordinates. As noted by Neto and Ambrósio (2003), the constraint stabilization method simply keeps the constraint violations under control, such that the error shown in Fig. 4.2 is integrated resulting in the increasing differences over time from the trajectories shown in Fig. 4.3. However, the UAV and payload system is designed to operate under a closed-loop controller, and then, the observed differences related to these different approaches to model the dynamics are insignificant.



Figure 4.2: Constraint equation from the single-lift system for different stabilization parameters.



Figure 4.3: Position error from the UAV and payload for $\alpha_f = 100$.

4.2 MISSION DEFINITION

The transportation mission considered herein consists of assigning a desired path for the leader UAV, whereas the followers collaboratively lift the suspended payload and maintain the relative distances through virtual constraints. The mission trajectory assigned to the leader is composed by four different parts: i) ascending with a climb rate of -0.3 m.s^{-1} for 10 seconds, ii) cruise in the \hat{i} direction with velocity 0.5 m.s⁻¹ for 15 seconds, iii) cruise in the \hat{j} direction with velocity 0.5 m.s⁻¹ by 15 seconds, and iv) cruise in the $-\hat{i}$ direction with velocity 0.5 m.s⁻¹ until t = 100 s, until the formation achieves the final flight time, i.e., $t_{max} = 120$ s.

The atmospheric disturbance is a lateral gust acting on the payload in the $+\hat{j}$ direction. This gust effect is positioned at $x_{go,1} = -4$ m, $y_{go,1} = 6$ m, with a gradient of 2 m and reference velocity equal to 3 m.s⁻¹. A gust load with similar characteristics action in the $+\hat{i}$ direction is defined at $x_{go,2} = -12$ m and $y_{go,2} = 6$ m.

The system parameters are further presented in Tab. 4.2, and the UAVs for both the following investigations are assumed to be identical. The maximum thrust due to the rotor parameter is 32.28 N, which limit the maximum payload to $m_p = 0.79$ kg. However, this upper limit is not practical because the attitude control needs to balance the rotors speed to rotate the UAV by reducing the lift and the UAV altitude control can fail. Then, a payload with weight exceeding such limit is defined, and the formation controller parameters α_{vc} and β_{vc} are arbitrary chosen.

narameter	description	value	unit
parameter	description	value	um
m	UAV mass	2.5	kg
J	inertia tensor	diag $(0.957 \ 0.486 \ 0.355)$	$\rm kg.m^2$
l_r	arm length	0.2	m
k_T	thrust coefficient	$1.83 \cdot 10^{-6}$	$N.s^2/rad^2$
k_D	moment coefficient	$1.85 \cdot 10^{-7}$	$N.m.s^2/rad^2$
$\omega_{r,min}$	minimum rotor speed	0	rad/s
$\omega_{r,max}$	maximum rotor speed	2100	rad/s
m_p	payload mass	0.8	kg
l	suspension length	10.26	m
c_{pd}	payload drag coefficient	0.12	-
$lpha_{vc}$	formation derivative gain	2.1	-
β_{vc}	formation proportional gain	1.3	-

 Table 4.2: Aerial transportation mission parameters.

4.3 TWIN-LIFT

The dynamic model for describing the twin-lift system is obtained by setting n = 2 in Algorithm 1, and it consists of a formation with two identical UAVs coordinated by using the leader-follower strategy. The formation is parameterized with a fixed separation of $d_x = 3.46 \text{ m}, d_y = -2.00 \text{ m}$ and $d_z = 0.00 \text{ m}$ between the leader and follower aerial vehicles (see Sec. 3.2).

Figure 4.4 presents the resulting constraints in position level (see Eq. 2.42) from the leader and follower, denoted by the curves g_1 and g_2 , respectively. The constraints are satisfied within an error of $2 \cdot 10^{-2}$ m², which is considered as an acceptable result for practical applications. During the hover phase, the constraints asymptotically converges to zero due to the stabilization method, such as previously noted by Baumgarte (1972).



Figure 4.4: Constraints over time from the twin-lift system.

Figure 4.5 shows the top-view from the trajectories. The payload is set to an initial angle with respect to the oscillation plane formed by the UAVs and the vertical direction. This condition introduces an initial pendulum motion that is gradually suppressed due to the included drag force (see Eq. 2.53). According to these results, this oscillatory motion is further excited when the UAVs changes directions and when the payload moves through regions with gusts.



Figure 4.5: Top-view from the twin-lift simulation.

Figure 4.6 shows the roll and pitch angles from the UAVs. Note that the UAVs keeps its non-zero attitude. This behavior is required to balance the forces to constraint the payload, which are at an angle with respect to the vertical direction resulting in forces in the \hat{i} and \hat{j} directions. These constraint forces are not included in the control design architecture, and they are considered as disturbances instead. They pull the UAVs towards each other, and this behavior increases the errors from the virtual constraints for the follower controller, and from the leader outer-loop. Then, a compensating control is considered (see Eq. 3.13), which results in that observed non-zero attitude.



Figure 4.6: Attitude from the UAVs considering the twin-lift configuration.

Figure 4.7 shows the virtual constraint from the twin-lift formation control. The UAVs start at the desired formation, and during the climb phase the follower UAV exhibits a time delay to keep at the same altitude as the leader vehicle, such as indicated by the curve $g_{vc}|_{\hat{k}}$. This behavior occurs since the reference for this parameter is updated in the outer-loop instead of the inner-loop, such as verified to the leader UAV. The follower

UAV accurately keeps the leader altitude for the remaining of the mission, even with the additional weight from the payload, and such performance is achieved due to the inner-loop controller robustness. In addition, regarding to the virtual constraints, i.e., $g_{vc}|_{\hat{j}}$ and $g_{vc}|_{\hat{j}}$, the payload pulls the UAVs close to each other, as shown by the difference from the follower UAV with respect to its position in the formation.



Figure 4.7: Virtual constraints from the twin-lift formation control.

Figure 4.8 shows the thrust forces for both UAVs. During the climb phase, since the follower vehicle is flying at a lower altitude than the leader, the required thrust from the leader is 6.14% higher than that force required from the follower. The required thrust force is approximately distributed equally for the remaining of the mission.



Figure 4.8: Thrust force required from the UAVs in the twin-lift configuration.



Figure 4.9: UAVs trajectories from the twin-lift configuration.

4.4 MULTI-LIFT

The multi-lift system configuration is evaluated by considering n = 3, i.e., three UAVs. Two of them are defined as followers and their outer-loop controllers are replaced by a single controller designed according to Sec. 3.2. The desired formation geometry for this case is the "V" formation, and the geometric parameters are d = 4 m and $\gamma = 30^{\circ}$. Figure 4.10 shows the corresponding constraint equation. Similarly to the twin-lift simulation, the constraint stabilization parameters reduces the constraint error over time until they stabilize at zero.



Figure 4.10: Constraints from the multi-lift system.

Figure 4.11 presents the virtual constraints from the multi-lift system. There are 6

virtual constraints to define the "V" formation, such that 3 of them constrain the position of the first follower to the leader, and the other ones constrain the position of the second follower to the leader. Note that, there is a persistent error in this multi-lift configuration, but its value decreases in comparison with the twin-lift system. This behavior occurs because the payload weight is also distributed to the additional UAV.



Figure 4.11: Virtual constraints from the multi-lift formation control.

Figure 4.12 presents the thrust forces computed for this multi-lift configuration. Their magnitudes change more substantially when the formation changes the phase of flight.



Figure 4.12: Thrust forces from the UAVs in the multi-lift configuration.

Figure 4.13 illustrates the top-view of the trajectories. The payload weight changes the formation geometry, resulting in the UAVs flying closer to each other. Figure 4.14 presents

the multi-lift system trajectories of the UAVs simultaneously lifting the heavy payload.



Figure 4.13: Trajectories (Top-view) for the multi-lift configuration.



Figure 4.14: Trajectories from the multi-lift system.

4.5 RESULTS

The multi-lift system in general can reduce the payload oscillations. Figure 4.15 presents the payload position computed when considering the twin-lift configuration $(x_p|_{TL})$, and multi-lift system $(x_p|_{ML})$. Note that the twin-lift system exhibits a decaying oscillation over time, due to the effect of the drag force. On the oher hand, no significant oscillation is observed when considering multi-lift systems. In addition, note that the gust load generates higher oscillations when considering the first gust, which shows that payload dynamic depends on formation geometry.



Figure 4.15: Comparison between the payload position when considering the twin-lift and multi-lift configurations.

Figure 4.16 shows a comparison of the virtual constraint error in the twin-lift and multi-lift configurations. Additional vehicle in the multi-lift configuration results in the decrease of the disturbance force exerted to each one from the vehicles. The effect is that the virtual constraint error is decreased in comparison with the twin-lift configuration.



Figure 4.16: Comparison between the virtual constraints when considering the twin-lift and multi-lift configurations.

Figure 4.17 shows the comparison from the rotor speed from the leader vehicle for the twin-lift configuration $(\omega_{r,1}|_{TL})$, and for the multi-lift configuration $(\omega_{r,1}|_{ML})$. Note that the rotor speed for both configurations is below the saturation region, as indicated by the gray region. The multi-lift configuration allows one to reduce 2.37% from the rotor speed, for the illustrated time-instant. The immediate advantage is a reduction from the required power, as the rotor speed directly correlates with the energy consumption, thus increasing the multi-lift flight time.



Figure 4.17: Comparison from the rotor speed from the leader in the twin-lift and multi-lift formations.

Chapter 5 FINAL REMARKS

The current technology readness level of UAVs allows one to use multirotor aerial vehicles for applications in which the cost and complexity of conventional aviation are prohibitive. However, due to the characteristics from the multirotor configuration, the applications are limited to those ones involving small weight and short-range flights. Formation flight can address the first limitation by collaboratively transporting heavy-payloads, i.e., distributing the additional weight on multiple vehicles, such that the force exerted by each one is within its operational range.

Formation flight also provides means for including redundancy, since a failure of an individual UAV occurs, the failed vehicle can decouple from the formation, and the mission can be continued. However, the design of such features require a robust computational environment to evaluate the controller performance and stability before experimental tests with prototypes. This type of tool allows one to collect data and analyze the controller in different scenarios.

This present work introduces a methodology to investigate the dynamics of a multilift UAV formation, by using the Euler-Lagrange formalism combined with Lagrange multipliers for dealing with the geometric constraints. The result contribute to future works involving designing controllers by providing a complete modeling for evaluating the system dynamics for a formation with an arbitrary number of UAVs. A leader-follower formation controller is investigated using virtual constraints to guide the UAVs in a formation through the desired trajectory.

5.1 CONCLUSIONS

The strategy of using multi-lift systems offers advantages in comparison with the use of a single high-lift vehicle such as a lower acquisition and operational costs of a smaller UAV. The use of multiple vehicles is also more convenient to define a redundant configuration to increase the mission robustness, in relation to failure events for an instance, because the failed UAV can decouple from the formation and the mission can continue with the remaining UAVs. In addition, the use of three or more vehicles can reduce the pendulum

dynamics commonly verified in both single and twin-lift configurations. This reduction is benefit since the periodic disturbance introduced by the pendulum motion may adverse affect the system performance. The payload pendulum motion verified in both single-lift and twin-lift configurations is introduced when the UAVs change their directions or due to the wind gusts effects.

The obtained dynamic model for the multi-lift configuration is written in terms of an arbitrary number of vehicles which allows one to investigate formations with any particular aerial transportation configuration, such as the single-lift, twin-lift, and multi-lift. The model is obtained by transforming a system of DAE into a system of ODE and involving an index-reduction with a constraint stabilization procedure. The approach introduces an extra-step of analyzing and tuning the stabilization parameters to obtain the constraints satisfied within a specified tolerance.

The twin-lift and multi-lift configurations require a non-zero attitude from each UAV to stabilize the system, due to the additional weight from the suspended payload. This condition usually is very difficult to be achieved by a human remotely controlled formation. However, the results herein show that a multi-lift configuration using a formation controller using virtual constraints can be successfully employed to transport heavy-payloads. It is assumed that each follower vehicle knows its relative distance from the leader, and then, the followers can move to their designed position in the "V" formation geometry by artificially introducing virtual constraint forces.

The formation controller reference model consists of disconnected UAVs moving without the suspended payload. This design consideration in a multi-lift mission considers the payload as a disturbance source. The result show that the payload pulls the UAVs close to each other. This approximation also increases the formation virtual constraints errors, and then a compensating control force is included to limit the pulling effect.

5.2 SUGGESTIONS FOR FUTURE WORK

The following suggestions are proposed for future works in this field.

- extend the modeling for rigid-body payloads and cable-suspension attachment off the UAVs center of mass;
- design of a payload-based controller to reduce pendulum oscillations and keep the formation geometry;
- design a controller to account for the propulsion-saturation and reorganize the formation to improve the payload-weight distribution;
- implement formation fault-tolerant controllers to increase mission robustness.

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