

Mobile robotics: Inertial  
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# Contents

<b>1</b>	<b>Lie groups</b>	<b>7</b>
1.1	Rotation matrices . . . . .	7
1.2	Groups . . . . .	8
1.3	Lie groups . . . . .	10
1.4	Lie algebra . . . . .	12
1.5	Rotation vector . . . . .	14
1.6	Rodrigues rotation formulas . . . . .	16
1.7	Evolution of the rotation matrix . . . . .	19
1.8	Interpolation . . . . .	20
1.9	Coordinate system change . . . . .	20
<b>2</b>	<b>Euler angles</b>	<b>29</b>
2.1	Definition . . . . .	29
2.2	Gimbal lock . . . . .	31
2.3	Rotation matrix to Euler angles . . . . .	31
2.4	Rotation vector of a moving Euler matrix . . . . .	33
<b>3</b>	<b>Inertial unit</b>	<b>39</b>
3.1	Mechanization equations . . . . .	39
3.2	With gravity . . . . .	41
3.3	Integration scheme . . . . .	41
3.4	Dead reckoning . . . . .	42
<b>4</b>	<b>Dynamic modeling</b>	<b>51</b>
4.1	Principle . . . . .	51
4.2	Equation of a 3D robot . . . . .	52
4.3	Modeling a quadrotor . . . . .	53
4.4	Euler-Lagrange method . . . . .	55
4.5	Proof of the Euler-Lagrange-equations . . . . .	55
4.6	Modeling a simple pendulum using the Euler-Lagrange method . . . . .	57

<b>5</b>	<b>Inertial control</b>	<b>69</b>
5.1	Control the accelerations . . . . .	69
5.2	Positioner . . . . .	70
5.3	Orientator . . . . .	70
5.4	Controller . . . . .	72
5.5	Two-dimensional robots . . . . .	73

# Introduction

A *mobile robot* can be defined as a mechanical system capable of moving in its environment in an autonomous manner. For that purpose, it must be equipped with:

- *sensors* that will collect knowledge of its surroundings (which it is more or less aware of) and determine its location ;
- *actuators* which will allow it to move ;
- an *intelligence* (or algorithm, regulator), which will allow it to compute, based on the data gathered by the sensors, the commands to send to the actuators in order to perform a given task.

Finally, to this we must add the *surroundings* of the robot which correspond to the world in which it evolves and its *mission* which is the task it has to accomplish. Mobile robots are constantly evolving, mainly from the beginning of the 2000s, in military domains (airborne drones [1], underwater robots [2], etc.), and even in medical and agricultural fields. They are in particularly high demand for performing tasks considered to be painful or dangerous to humans. This is the case for instance in mine-clearing operations, the search for black boxes of damaged aircraft on the ocean bed and planetary exploration. Artificial satellites, launchers (such as Ariane V), driverless subways and elevators are examples of mobile robots. Airliners, trains and cars evolve in a continuous fashion towards more and more autonomous systems and will very probably become mobile robots in the following decades.

*Mobile robotics* is the discipline which looks at the design of mobile robots [3]. It is based on other disciplines such as automatic control, signal processing, mechanics, computing and electronics. The aim of this book is to give an overview of the tools and methods of robotics which will aid in the design of mobile robots. The robots will be modeled by *state equations*, *i.e.*, first order (mostly non-linear) differential equations. These state equations can be obtained by using the laws of mechanics. It is not in our objectives to teach, in detail, the methods of robot modeling (refer to [4] and [5] for more information on the subject), merely to recall its principles. By *modeling*, we mean obtaining the state equations. This step is essential for simulating robots as well as designing controllers. In this book, we will provide the principle and the tools for three-dimensional modeling of a solid (non-articulated) robot modeling in three-dimensions such as an airplane, a quadcopter, a submarine, and so forth.

Through this modeling we will introduce a number of fundamental concepts in robotics such as state representation, rotation matrices and Euler angles, Lie groups. We will show their applications for modeling robots and also for localization.

The robots we will consider are assumed to be put into a state representation form:

$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t)) \\ \mathbf{y}(t) = \mathbf{g}(\mathbf{x}(t), \mathbf{u}(t)) \end{cases}$$

where  $\mathbf{x}$  is the state vector,  $\mathbf{u}$  the input vector and  $\mathbf{y}$  the vector of measurements [4]. We will call *modeling* the step which consists of finding a more or less accurate state representation of the robot. In general, constant parameters may appear in the state equations (such as the mass and the moment of inertia of a body, viscosity, etc.). In such cases, an identification step might prove to be necessary. We will assume that all of the parameters are known. Of course, there is no systematic methodology that can be applied for modeling a mobile robot. We will present the tools which allow to reach a state representation of three-dimensional solid robots in order to acquire a certain experience which will be helpful when modeling his/her own robots. This modeling will also allow us to recall a number of important concepts in Euclidean geometry, which are fundamental in mobile robotics.

For the localization, a fundamental sensor is the inertial units, commonly referred to as Inertial Measurement Units (IMUs). They provide critical data on a robot's motion and orientation. By measuring linear acceleration, angular velocity, and sometimes magnetic fields, these sensors enable robots to navigate, stabilize, and interact with their environment accurately. This course will explore the principles of inertial sensing, the mathematical tools, the components of IMUs, and their integration into robotic systems.

# Chapter 1

## Lie groups

This chapter recalls a number of important concepts in kinematics which will be useful for the modeling in three dimensions. To get the state equations of a mobile robot, it is important to characterize its geometry, its degrees of freedom. The theoretical foundations can be found in Lie groups to be introduced in this chapter. In this course, the use a Lie groups will mainly used to represent the rotations in a 3D-space.

### 1.1 Rotation matrices

For three-dimensional modeling, it is essential to have a good understanding of the concepts related to rotation matrices, which are recalled in this section. It is by using this tool that we will perform our coordinate system transformations and position our objects in space.

Let us recall that the  $j^{\text{th}}$  column of the matrix of a linear application of  $\mathbb{R}^n \rightarrow \mathbb{R}^n$  represents the image of the  $j^{\text{th}}$  vector  $\mathbf{e}_j$  of the standard basis (see Figure 1.1). Thus, the expression of a rotation matrix of angle  $\theta$  in the plane  $\mathbb{R}^2$  is given by:

$$\mathbf{R} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}.$$

Concerning rotations in the space  $\mathbb{R}^3$  (see Figure 1.2), it is important to specify the axis of rotation. We distinguish three main rotations: the rotation around the  $Ox$  axis, the one around the  $Oy$  axis and the one around the  $Oz$  axis.

The associated matrices are respectively given by:

$$\mathbf{R}_x = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \theta_x & -\sin \theta_x \\ 0 & \sin \theta_x & \cos \theta_x \end{pmatrix}, \quad \mathbf{R}_y = \begin{pmatrix} \cos \theta_y & 0 & \sin \theta_y \\ 0 & 1 & 0 \\ -\sin \theta_y & 0 & \cos \theta_y \end{pmatrix}, \quad \mathbf{R}_z = \begin{pmatrix} \cos \theta_z & -\sin \theta_z & 0 \\ \sin \theta_z & \cos \theta_z & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

Let us recall the formal definition of a rotation. A *rotation* is a linear application which is an isometry (*i.e.*, it preserves the scalar product) which is direct (it does not change the orientation in space).

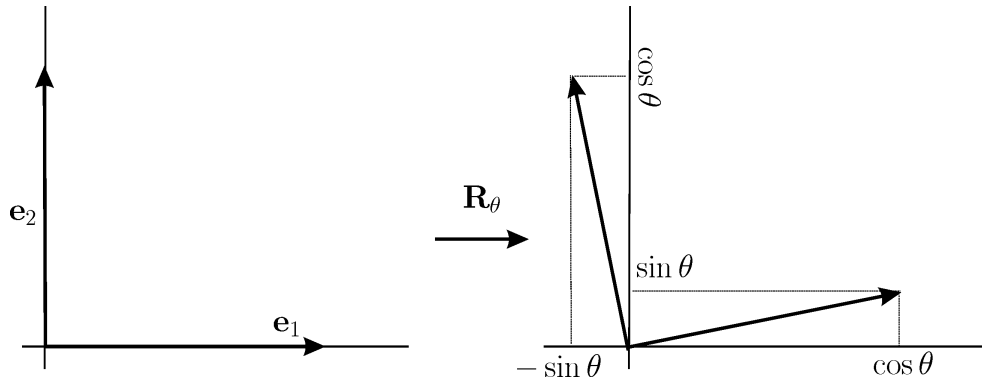


Figure 1.1: Rotation of angle  $\theta$  in a plane

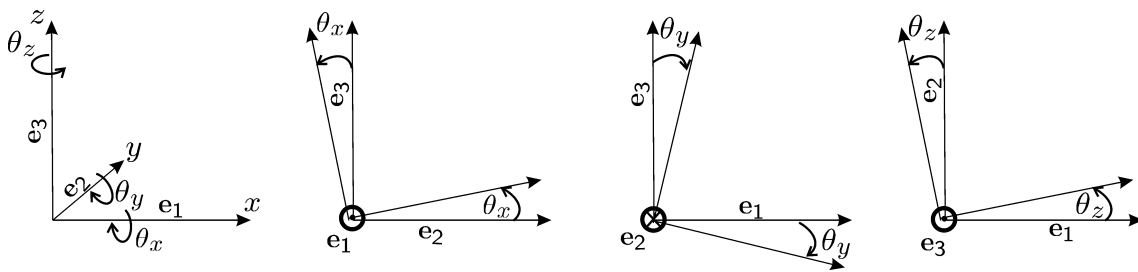


Figure 1.2: Rotations in  $\mathbb{R}^3$  following various viewing angles

**Theorem 1.** A matrix  $\mathbf{R}$  is a rotation matrix if and only if :

$$\mathbf{R}^T \cdot \mathbf{R} = \mathbf{I} \text{ and } \det \mathbf{R} = 1.$$

*Proof.* The scalar product is preserved by  $\mathbf{R}$  if, for any  $\mathbf{u}$  and  $\mathbf{v}$  in  $\mathbb{R}^n$ , we have :

$$(\mathbf{R}\mathbf{u})^T \cdot (\mathbf{R}\mathbf{v}) = \mathbf{u}^T \cdot \mathbf{R}^T \mathbf{R} \cdot \mathbf{v} = \mathbf{u}^T \mathbf{v}.$$

Therefore  $\mathbf{R}^T \mathbf{R} = \mathbf{I}$ . The symmetries relative to a plane, as well as all the other improper isometries (isometries that change the orientation of space, such as a mirror), also verify the property  $\mathbf{R}^T \mathbf{R} = \mathbf{I}$ . The condition  $\det \mathbf{R} = 1$  allows us to be limited to the isometries which are direct.  $\square$

## 1.2 Groups

The set of rotations in  $\mathbb{R}^n$  forms a group with respect to the multiplication [6]. This means that we can multiply two rotations, but adding them or multiplying by some real number  $\alpha$  is forbidden. In this section, we recall what is a group.

**Definition 2.** A *group* is a set  $G$  together with a binary operation  $\cdot : G \times G \rightarrow G$ , usually written as  $(a, b) \mapsto a \cdot b$ , such that the following axioms hold:

**Associativity:** For all  $a, b, c \in G$ ,  $(a \cdot b) \cdot c = a \cdot (b \cdot c)$

**Identity element:** There exists an element  $e \in G$  such that for all  $a \in G$ ,  $e \cdot a = a \cdot e = a$ .

**Inverse element:** For each  $a \in G$ , there exists an element  $a^{-1} \in G$  such that  $a \cdot a^{-1} = a^{-1} \cdot a = e$

To compute in groups, it is convenient to use a matrix representation. As an illustration, consider the hyperoctahedral group  $B_2$  which is the group of symmetries of the hypercube  $[-1, 1]^2$  [7] of  $\mathbb{R}^2$ . If we use the matrix representation, the elements of  $B_2$  are

$$\begin{aligned} \sigma_0 &= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} & \sigma_1 &= \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \\ \sigma_2 &= \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} & \sigma_3 &= \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix} \\ \sigma_4 &= \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} & \sigma_5 &= \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \\ \sigma_6 &= \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} & \sigma_7 &= \begin{pmatrix} 0 & -1 \\ -1 & 0 \end{pmatrix} \end{aligned}$$

It is trivial to check that  $B_2$  is a group with respect to the multiplication, or equivalently, for the composition  $\circ$ . A symmetry  $\sigma$  will be seen both as a linear function from  $\mathbb{R}^2$  to  $\mathbb{R}^2$  or as a  $2 \times 2$  matrix. For instance, we will write equivalently

$$\sigma_5 = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \text{ or } \sigma_5 : \begin{cases} \mathbb{R}^2 & \mapsto \mathbb{R}^2 \\ (x_1, x_2) & \mapsto (x_2, -x_1) \end{cases}$$

Figure 1.3 corresponds to the multiplication table (known as Cayley table [8]).

	$\sigma_0$	$\sigma_1$	$\sigma_2$	$\sigma_3$	$\sigma_4$	$\sigma_5$	$\sigma_6$	$\sigma_7$
$\sigma_0$	$\sigma_0$	$\sigma_1$	$\sigma_2$	$\sigma_3$	$\sigma_4$	$\sigma_5$	$\sigma_6$	$\sigma_7$
$\sigma_1$	$\sigma_1$	$\sigma_0$	$\sigma_6$	$\sigma_4$	$\sigma_3$	$\sigma_7$	$\sigma_2$	$\sigma_5$
$\sigma_2$	$\sigma_2$	$\sigma_5$	$\sigma_0$	$\sigma_7$	$\sigma_6$	$\sigma_1$	$\sigma_4$	$\sigma_3$
$\sigma_3$	$\sigma_3$	$\sigma_4$	$\sigma_7$	$\sigma_0$	$\sigma_1$	$\sigma_6$	$\sigma_5$	$\sigma_2$
$\sigma_4$	$\sigma_4$	$\sigma_3$	$\sigma_5$	$\sigma_1$	$\sigma_0$	$\sigma_2$	$\sigma_7$	$\sigma_6$
$\sigma_5$	$\sigma_5$	$\sigma_2$	$\sigma_4$	$\sigma_6$	$\sigma_7$	$\sigma_3$	$\sigma_0$	$\sigma_1$
$\sigma_6$	$\sigma_6$	$\sigma_7$	$\sigma_1$	$\sigma_5$	$\sigma_2$	$\sigma_0$	$\sigma_3$	$\sigma_4$
$\sigma_7$	$\sigma_7$	$\sigma_6$	$\sigma_3$	$\sigma_2$	$\sigma_5$	$\sigma_4$	$\sigma_1$	$\sigma_0$

Figure 1.3: Multiplication table

Now, due to the fact that we have a group, the transitivity and the inversion property should be satisfied. As a consequence, the entries of the table are dependent. We can easily check that from the second and the third lines, we can reconstruct the whole table. We say that the two elements  $\sigma_1, \sigma_2$  are generators of the group  $B_2$  or equivalently, we write  $B_2 = \langle \sigma_1, \sigma_2 \rangle$ .

Compute for instance  $\sigma_6 \circ \sigma_7$  using these two lines (red and blue in the figure). We get:

$$\sigma_6 \circ \sigma_7 = \sigma_1 \circ \sigma_2 \circ \sigma_1 \circ \underbrace{\sigma_2 \circ \sigma_1}_{\sigma_5} = \sigma_4$$

$$\underbrace{\underbrace{\sigma_1 \circ \sigma_2}_{\sigma_7} \circ \sigma_1}_{\sigma_3} = \sigma_4$$

We can easily check that  $\{\sigma_1, \sigma_2\}$  is a generator pair of  $B_2$ .

## 1.3 Lie groups

In robotics, Lie groups are often used to describe transformations (such as translation or rotations). The Lie algebra (following section) corresponds to velocities or equivalently to infinitesimal transformations. Lie group theory is useful, but requires some non trivial mathematical backgrounds.

A *Lie group* is a group  $G$  that is also a smooth manifold such that the multiplication and the inversion are both smooth maps. Some examples are

- $GL(n, \mathbb{R})$ : the group of invertible  $n \times n$  real matrices.
- $SO(n)$ : special orthogonal group, i.e. rotation matrices (to be detailed later).
- $SU(n)$ : special unitary group, important in quantum physics.

In this course, we will mainly consider  $SO(n)$ , the *special orthogonal group* (*special* because  $\det \mathbf{R} = 1$ , *orthogonal* because  $\mathbf{R}^T \cdot \mathbf{R} = \mathbf{I}$ ). It is trivial to check that  $(SO(n), \cdot)$  is a group where  $\mathbf{I}$  is the neutral element. Moreover, the multiplication and the inversion are both smooth. This makes  $SO(n)$  a *Lie group* which is a manifold of the set of matrices  $\mathbb{R}^{n \times n}$ .

The set  $\mathbb{R}^{n \times n}$  of  $n \times n$ -matrices is of dimension  $n^2$ . Since the matrix  $\mathbf{R}^T \cdot \mathbf{R}$  is always symmetric, the matrix equation  $\mathbf{R}^T \cdot \mathbf{R} = \mathbf{I}$  can be decomposed into  $\frac{n(n+1)}{2}$  independent scalar equations. For instance, for  $n = 2$ , we have  $\frac{2(2+1)}{2} = 3$  scalar equations:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \cdot \begin{pmatrix} a & b \\ c & d \end{pmatrix}^T = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \Leftrightarrow \begin{cases} a^2 + b^2 = 1 \\ ac + bd = 0 \\ c^2 + d^2 = 1 \end{cases}$$

As a consequence, the set  $SO(n)$  forms a manifold of dimension  $d = n^2 - \frac{n(n+1)}{2}$ .

- For  $n = 1$ , we get  $d = 0$ . The set  $SO(1)$  is a singleton which contains a single rotation matrix:  $R = 1$ .
- For  $n = 2$ , we get  $d = 1$ . We need a unique parameter (or angle) to represent the rotations of  $SO(2)$ .
- For  $n = 3$ , we get  $d = 3$ . We need 3 parameters (or angles) to represent  $SO(3)$ .

## Matrix Lie groups

A *matrix Lie group* is a subgroup  $G \subseteq GL(n, \mathbb{R})$  or  $GL(n, \mathbb{C})$  (the group of invertible  $n \times n$  real or complex matrices) such that  $G$  is a smooth manifold and the group operations (matrix multiplication and inversion) are smooth maps. Such groups are automatically smooth manifolds.

Matrix Lie groups are used to compute inside the group. Classical examples are

- The General Linear group:  $GL(n, \mathbb{R}) = \{\mathbf{A} \in M_n(\mathbb{R}) \mid \det \mathbf{A} \neq 0\}$ .
- The Orthogonal group:  $O(n) = \{\mathbf{A} \in GL(n, \mathbb{R}) \mid \mathbf{A}^T \mathbf{A} = \mathbf{I}\}$ .

We now give two examples: the complex numbers and the quaternions. These examples illustrate how familiar number systems naturally fit into the framework of matrix Lie groups.

## Complex numbers

Complex numbers forms also a Lie group ans can be represented as a Matrix Lie Group. Every complex number  $z = x + iy \in \mathbb{C}$  can be represented as a  $2 \times 2$  real matrix:

$$z \longleftrightarrow \begin{pmatrix} x & -y \\ y & x \end{pmatrix}.$$

Matrix multiplication reproduces complex multiplication. If

$$z_1 = x_1 + iy_1, \quad z_2 = x_2 + iy_2,$$

then

$$\begin{pmatrix} x_1 & -y_1 \\ y_1 & x_1 \end{pmatrix} \begin{pmatrix} x_2 & -y_2 \\ y_2 & x_2 \end{pmatrix} = \begin{pmatrix} x_1x_2 - y_1y_2 & -(x_1y_2 + y_1x_2) \\ x_1y_2 + y_1x_2 & x_1x_2 - y_1y_2 \end{pmatrix},$$

which corresponds to  $(x_1 + iy_1)(x_2 + iy_2)$ .

The group of unit complex numbers,

$$U(1) = \{z = x + iy \mid x^2 + y^2 = 1\}$$

corresponds to rotation matrices

$$\mathbf{R}(\theta) = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}.$$

It is a Lie groups diffeomorphic to the circle.

## Quaternions

The quaternions  $\mathbb{H}$  are a 4-dimensional algebra over  $\mathbb{R}$  with basis  $1, i, j, k$ , satisfying the multiplication rules

$$\begin{aligned} i^2 &= j^2 = k^2 = ijk = -1 \\ ij &= k, \quad jk = i, \quad ki = j, \\ ji &= -k, \quad kj = -i, \quad ik = -j \end{aligned}$$

Every quaternion  $q = a + bi + cj + dk$  can be represented by the real  $4 \times 4$  matrix

$$q \longleftrightarrow \begin{pmatrix} a & -b & -c & -d \\ b & a & -d & c \\ c & d & a & -b \\ d & -c & b & a \end{pmatrix}.$$

Quaternion multiplication corresponds exactly to matrix multiplication under this representation.

The set of unit quaternions is

$$Sp(1) = \{q = a + bi + cj + dk \in \mathbb{H} : a^2 + b^2 + c^2 + d^2 = 1\}.$$

This group is diffeomorphic to the 3-sphere  $S^3 \subset \mathbb{R}^4$ , making it a compact 3-dimensional Lie group.

## 1.4 Lie algebra

**Definition 3.** An *algebra* is an algebraic structure  $(\mathcal{A}, +, \times, \cdot)$  over the field  $\mathbb{R}$ , if

- (i)  $(\mathcal{A}, +, \cdot)$  is a vector space over  $\mathbb{R}$  ;
- (ii) the multiplication rule  $\times$  of  $\mathcal{A} \times \mathcal{A} \rightarrow \mathcal{A}$  is left- and right-distributive with respect to  $+$
- (iii) for all  $\alpha, \beta \in \mathbb{R}$ , and for all  $x, y \in \mathcal{A}$ ,  $\alpha \cdot x \times \beta \cdot y = (\alpha\beta) \cdot (x \times y)$ .

Note that in general, an algebra is non-commutative ( $x \times y \neq y \times x$ ) and non-associative ( $(x \times y) \times z \neq x \times (y \times z)$ ).

**Definition 4.** A *Lie algebra*  $(\mathfrak{g}, +, [], \cdot)$  is a non-commutative and non-associative algebra in which multiplication, denoted by a so-called *Lie bracket*, verifies

- (i)  $[\cdot, \cdot]$  that is bilinear, i.e., linear with respect to each variable ;
- (ii)  $[x, y] = -[y, x]$  (antisymmetry)
- (iii)  $[x, [y, z]] + [y, [z, x]] + [z, [x, y]] = 0$  (Jacobi relation).

For every Lie group we can define the associated *Lie algebra* [6] as stated by the Lie's third theorem. Lie algebras allow us to consider infinitesimal motions around a given element if the Lie group in order to use derivatives or differential methods.

### Lie algebra of a matrix Lie group

Consider a matrix Lie Group  $G \subseteq \mathbb{R}^{n \times n}$ . Consider the identity matrix  $\mathbf{I}$  which belongs to  $G$ . If we move  $\mathbf{I}$  by adding a small matrix say  $\mathbf{A} \cdot dt$  of  $\mathbb{R}^{n \times n}$ , we generally do not obtain an element of  $G$ . The Lie algebra  $\mathfrak{g}$  corresponds to the matrices  $\mathbf{A}$  such that  $\mathbf{I} + \mathbf{A} \cdot dt \in G$  or equivalently to the matrices  $\mathbf{A}$  such that  $e^{dt\mathbf{A}} \in G$  for a small  $dt$ . Formally, the Lie algebra is defined by

$$\mathfrak{g} = \{\mathbf{A} \in \mathbb{R}^{n \times n} \mid e^{\mathbf{A} \cdot dt} \in G \text{ for all small } dt\}.$$

To get the Lie algebra of  $G$ ,

- We write down the defining equation of  $G$ . We check that this condition is consistent with the fact that  $G$  is a group.
- Substitute  $e^{dt\mathbf{A}}$  and differentiate at  $dt = 0$  (we get  $\mathbf{I} + \mathbf{A} \cdot dt$  since  $e^{dt\mathbf{A}} = \mathbf{I} + \mathbf{A} \cdot dt + o(dt)$ ).
- The resulting linear conditions on  $\mathbf{A}$  give  $\mathfrak{g}$ .
- The Lie bracket is the commutator:  $[\mathbf{X}, \mathbf{Y}] = \mathbf{X}\mathbf{Y} - \mathbf{Y}\mathbf{X}$ .

This is illustrated by Figure 1.4.

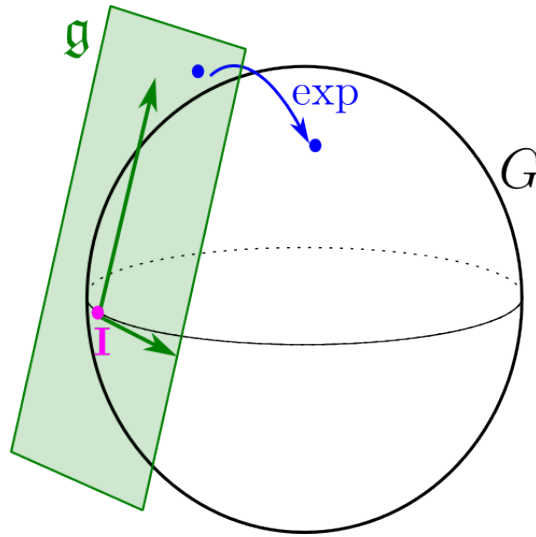


Figure 1.4: The Lie algebra of the Lie group  $G$  is the tangent space at the identity

### Example: the rotation matrices

As an example, let us find the Lie algebra associated with the the Lie group  $G = SO(n)$ . The defining equation is  $\mathbf{R}^T \cdot \mathbf{R} = \mathbf{I}$ . We check that if  $\mathbf{R}_1^T \cdot \mathbf{R}_1 = \mathbf{I}$  and if  $\mathbf{R}_2^T \cdot \mathbf{R}_2 = \mathbf{I}$ , we also have  $(\mathbf{R}_1\mathbf{R}_2)^T \cdot (\mathbf{R}_1\mathbf{R}_2) = \mathbf{I}$  and  $(\mathbf{R}_1^{-1})^T \cdot \mathbf{R}_1^{-1} = \mathbf{I}$ .

Take a small matrix  $\mathbf{A} \cdot dt$  of  $\mathbb{R}^{n \times n}$ . We have

$$e^{\mathbf{A} \cdot dt} \in SO(n)$$

if

$$(e^{\mathbf{A} \cdot dt})^T \cdot e^{\mathbf{A} \cdot dt} = \mathbf{I},$$

i.e.,

$$(\mathbf{I} + \mathbf{A} \cdot dt)^T \cdot (\mathbf{I} + \mathbf{A} \cdot dt) = \mathbf{I} + o(dt)$$

*i.e.*,  $\mathbf{A} \cdot dt + \mathbf{A}^T \cdot dt = \mathbf{0} + o(dt)$ . Therefore,  $\mathbf{A}$  should be skew-symmetric. It means that we are able to move in  $SO(n)$  around  $\mathbf{I}$  by adding infinitesimal skew-symmetric matrices  $\mathbf{A} \cdot dt$  that are not elements of  $SO(n)$ . Formally, we define the *Lie algebra* associated to  $SO(n)$  as follows

$$\mathfrak{so}(n) = \text{Lie}(SO(n)) = \{\mathbf{A} \in \mathbb{R}^{n \times n} \mid \mathbf{I} + \mathbf{A} \cdot dt \in SO(n)\}$$

and it corresponds skew-symmetric matrices of  $\mathbb{R}^{n \times n}$ .

If now we want to move around any matrix  $\mathbf{R}$  of  $SO(n)$ , we generate a rotation matrix around  $\mathbf{I}$  and we transport it to  $\mathbf{R}$ . We get  $\mathbf{R}(\mathbf{I} + \mathbf{A} \cdot dt)$  where  $\mathbf{A}$  is skew-symmetric. It means that we add to  $\mathbf{R}$  the matrix  $\mathbf{R} \cdot \mathbf{A} \cdot dt$ .

## Interpretation of the Lie bracket in the Lie group

**Proposition 5.** *Consider a matrix Lie group  $G$ . Take two matrices  $g_a = e^{\mathbf{A}dt}$  and  $g_b = e^{\mathbf{B}dt}$  near the identity in  $G$ . We have*

$$g_a^{-1} g_b^{-1} g_a g_b = e^{-\mathbf{A}dt} \cdot e^{-\mathbf{B}dt} \cdot e^{\mathbf{A}dt} \cdot e^{\mathbf{B}dt} = e^{[\mathbf{A}, \mathbf{B}]dt^2 + o(dt^2)}$$

This proposition shows that if in  $G$ , we can only move following  $g_a$  and  $g_b$  we can also move along the new direction  $g_c = e^{[\log g_a, \log g_b]}$  still staying at the neighborhood of the identity.

*Proof.* We compute a second order approximation of  $g_a^{-1} g_b^{-1} g_a g_b$  : □

$$\begin{aligned} e^{\mathbf{B}dt} &= \mathbf{I} + \mathbf{B} \cdot dt + \frac{1}{2} \mathbf{B}^2 \cdot dt^2 + o(dt^2) \\ e^{\mathbf{A}dt} \cdot e^{\mathbf{B}dt} &= \left( \mathbf{I} + \mathbf{A} \cdot dt + \frac{1}{2} \mathbf{A}^2 \cdot dt^2 + o(dt^2) \right) \cdot e^{\mathbf{B}dt} \\ &= \mathbf{I} + (\mathbf{A} + \mathbf{B}) \cdot dt + \left( \frac{1}{2} \mathbf{A}^2 + \frac{1}{2} \mathbf{B}^2 + \mathbf{A}\mathbf{B} \right) \cdot dt^2 + o(dt^2) \\ e^{-\mathbf{B}dt} \cdot e^{\mathbf{A}dt} \cdot e^{\mathbf{B}dt} &= \left( \mathbf{I} - \mathbf{B} \cdot dt + \frac{1}{2} \mathbf{B}^2 \cdot dt^2 \right) \cdot e^{\mathbf{A}dt} \cdot e^{\mathbf{B}dt} \\ &= \mathbf{I} + \mathbf{A} \cdot dt + \left( \frac{1}{2} \mathbf{A}^2 + \mathbf{A}\mathbf{B} - \mathbf{B}\mathbf{A} \right) \cdot dt^2 + o(dt^2) \\ e^{-\mathbf{A}dt} \cdot e^{-\mathbf{B}dt} \cdot e^{\mathbf{A}dt} \cdot e^{\mathbf{B}dt} &= \left( \mathbf{I} - \mathbf{A} \cdot dt + \frac{1}{2} \mathbf{A}^2 \cdot dt^2 \right) \cdot e^{-\mathbf{B}dt} \cdot e^{\mathbf{A}dt} \cdot e^{\mathbf{B}dt} \\ &= \left( \mathbf{I} + (\mathbf{A}\mathbf{B} - \mathbf{B}\mathbf{A}) \cdot dt^2 \right) + o(dt^2) \\ &= e^{[\mathbf{A}, \mathbf{B}]dt^2 + o(dt^2)} \end{aligned}$$

## 1.5 Rotation vector

Consider a rotation matrix  $\mathbf{R}$  depending on time  $t$ . By differentiating the relation  $\mathbf{R}\mathbf{R}^T = \mathbf{I}$ , we get

$$\dot{\mathbf{R}} \cdot \mathbf{R}^T + \mathbf{R} \cdot \dot{\mathbf{R}}^T = \mathbf{0}.$$

Thus, the matrix  $\dot{\mathbf{R}} \cdot \mathbf{R}^T$  is a skew-symmetric matrix (*i.e.*, it satisfies  $\mathbf{A}^T = -\mathbf{A}$  and therefore its diagonal contains only zeros, and for each element of  $\mathbf{A}$ , we have  $a_{ij} = -a_{ji}$ ). We may therefore write, in the case where  $\mathbf{R}$  is of dimension  $3 \times 3$ :

$$\dot{\mathbf{R}} \cdot \mathbf{R}^T = \begin{pmatrix} 0 & -\omega_z & \omega_y \\ \omega_z & 0 & -\omega_x \\ -\omega_y & \omega_x & 0 \end{pmatrix}. \quad (1.1)$$

The vector  $\boldsymbol{\omega} = (\omega_x, \omega_y, \omega_z)$  is called the *rotation vector* or *Euler vector* associated with the pair  $(\mathbf{R}, \dot{\mathbf{R}})$ . It must be noted that  $\dot{\mathbf{R}}$  is not a matrix with good properties (such as for instance the fact of being skew-symmetric). On the other hand, the matrix  $\dot{\mathbf{R}} \cdot \mathbf{R}^T$  has the structure of Equation (1.1) since it allows to be positioned within the coordinate system in which the rotation is performed and this, due to the change of basis performed by  $\mathbf{R}^T$ . We will define the *cross product* between two vectors  $\boldsymbol{\omega}$  and  $\mathbf{x} \in \mathbb{R}^3$  as follows:

$$\boldsymbol{\omega} \wedge \mathbf{x} = \begin{pmatrix} \omega_x \\ \omega_y \\ \omega_z \end{pmatrix} \wedge \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_3\omega_y - x_2\omega_z \\ x_1\omega_z - x_3\omega_x \\ x_2\omega_x - x_1\omega_y \end{pmatrix} = \begin{pmatrix} 0 & -\omega_z & \omega_y \\ \omega_z & 0 & -\omega_x \\ -\omega_y & \omega_x & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}.$$

For each vector  $\boldsymbol{\omega} = (\omega_x, \omega_y, \omega_z)$ , we may associate the skew-symmetric matrix:

$$\wedge(\boldsymbol{\omega}) = \begin{pmatrix} 0 & -\omega_z & \omega_y \\ \omega_z & 0 & -\omega_x \\ -\omega_y & \omega_x & 0 \end{pmatrix}$$

which can be interpreted as the matrix associated with a cross product by the vector  $\boldsymbol{\omega}$ . The matrix  $\wedge(\boldsymbol{\omega})$  is also written  $\boldsymbol{\omega}\wedge$ . It is also called the *small adjoint* of  $\boldsymbol{\omega}$  and denoted by  $\text{ad}(\boldsymbol{\omega})$ . It should not be confused with the *large adjoint*  $\text{Ad}(\mathbf{R})$  also used in this context, but not here in this book.

**Proposition 6.** *If  $\mathbf{R}(t)$  is a rotation matrix that depends on time, its rotation vector is given by:*

$$\boldsymbol{\omega} = \wedge^{-1}(\dot{\mathbf{R}} \cdot \mathbf{R}^T). \quad (1.2)$$

*Proof.* This relation is a direct consequence of Equation (1.1).  $\square$

**Proposition 7.** *If  $\mathbf{R}$  is a rotation matrix in  $\mathbb{R}^3$  and if  $\mathbf{a}$  is a vector of  $\mathbb{R}^3$ , we have:*

$$\wedge(\mathbf{R} \cdot \mathbf{a}) = \mathbf{R} \cdot \wedge(\mathbf{a}) \cdot \mathbf{R}^T. \quad (1.3)$$

*Remark 8.* We sometimes define the *capitalized Adjoint* in  $\text{SO}(3)$  as

$$\text{Ad}_{\mathbf{R}} : \begin{array}{l} \mathfrak{so}(3) \mapsto \mathfrak{so}(3) \\ \mathbf{a}\wedge \mapsto \mathbf{R} \cdot (\mathbf{a}\wedge) \cdot \mathbf{R}^T \end{array}$$

This operation is linear in  $\mathbf{a}$ . We can thus define the *Adjoint Matrix*  $\text{Ad}_{\mathbf{R}}$ :

$$\text{Ad}_{\mathbf{R}}(\mathbf{a}) = \mathbf{Ad}_{\mathbf{R}} \cdot \mathbf{a}.$$

Since

$$\mathbf{Ad}_{\mathbf{R}} \cdot \mathbf{a} = \text{Ad}_{\mathbf{R}}(\mathbf{a}) = \wedge^{-1}(\mathbf{R} \cdot (\mathbf{a}\wedge) \cdot \mathbf{R}^T)$$

we have

$$\mathbf{Ad}_{\mathbf{R}} \cdot \mathbf{a} = \wedge^{-1}(\mathbf{R} \cdot (\mathbf{a}\wedge) \cdot \mathbf{R}^T) = \wedge^{-1}(\wedge(\mathbf{R} \cdot \mathbf{a})) = \mathbf{R} \cdot \mathbf{a}$$

Thus here  $\mathbf{Ad}_{\mathbf{R}} = \mathbf{R}$ . Therefore, whereas the *Adjoint* is an important tool in general Lie group, it is not essential if we limit ourselves to the Lie group  $\text{SO}(3)$  as in this book. This is why it will not be mentioned anymore.

*Proof.* Let  $\mathbf{x}$  be a vector of  $\mathbb{R}^3$ . We have:

$$\begin{aligned}\wedge(\mathbf{R} \cdot \mathbf{a}) \cdot \mathbf{x} &= (\mathbf{R} \cdot \mathbf{a}) \wedge \mathbf{x} = (\mathbf{R} \cdot \mathbf{a}) \wedge (\mathbf{R} \cdot \mathbf{R}^T \mathbf{x}) \\ &= \mathbf{R} \cdot (\mathbf{a} \wedge \mathbf{R}^T \cdot \mathbf{x}) = \mathbf{R} \cdot \wedge(\mathbf{a}) \cdot \mathbf{R}^T \cdot \mathbf{x}.\end{aligned}$$

□

**Proposition 9.** (*duality*). We have:

$$\mathbf{R}^T \dot{\mathbf{R}} = \wedge(\mathbf{R}^T \boldsymbol{\omega}). \quad (1.4)$$

This relation expresses the fact that the matrix  $\mathbf{R}^T \dot{\mathbf{R}}$  is associated with the rotation vector  $\boldsymbol{\omega}$  but expressed in the coordinate system of  $\mathbf{R}$  whereas  $\dot{\mathbf{R}} \cdot \mathbf{R}^T$  is associated to the same vector, expressed in the coordinate system of the standard basis.

*Proof.* We have:

$$\mathbf{R}^T \dot{\mathbf{R}} = \mathbf{R}^T (\dot{\mathbf{R}} \cdot \mathbf{R}^T) \mathbf{R} \stackrel{(1.2)}{=} \mathbf{R}^T \cdot \wedge(\boldsymbol{\omega}) \cdot \mathbf{R} \stackrel{(1.3)}{=} \wedge(\mathbf{R}^T \boldsymbol{\omega}).$$

□

## 1.6 Rodrigues rotation formulas

**Matrix exponential.** Given a square matrix  $\mathbf{M}$  of dimension  $n$ , its exponential can be defined as:

$$e^{\mathbf{M}} = \mathbf{I}_n + \mathbf{M} + \frac{1}{2!} \mathbf{M}^2 + \frac{1}{3!} \mathbf{M}^3 + \dots = \sum_{i=0}^{\infty} \frac{1}{i!} \mathbf{M}^i$$

where  $\mathbf{I}_n$  is the identity matrix of dimension  $n$ . It is clear that  $e^{\mathbf{M}}$  is of the same dimension as  $\mathbf{M}$ . Here are some of the important properties concerning the exponentials of matrices. If  $\mathbf{0}_n$  is the zero matrix of  $n \times n$  and if  $\mathbf{M}$  and  $\mathbf{N}$  are two matrices  $n \times n$ , then:

$$\begin{aligned}e^{\mathbf{0}_n} &= \mathbf{I}_n \\ e^{\mathbf{M}} \cdot e^{\mathbf{N}} &= e^{\mathbf{M}+\mathbf{N}} \text{ (if the matrices commute)} \\ \frac{d}{dt} (e^{\mathbf{M}t}) &= \mathbf{M} \cdot e^{\mathbf{M}t}\end{aligned}$$

**Matrix logarithm.** Given a matrix  $\mathbf{M}$  of dimension  $n$ , the matrix  $\mathbf{L}$  is said to be a matrix logarithm of  $\mathbf{M}$  if  $e^{\mathbf{L}} = \mathbf{M}$ . As for complex numbers, the exponential function is not a one-to-one function and matrices may have more than one logarithm. Using power series, we define the logarithm of a square matrix as

$$\log \mathbf{M} = \sum_{i=1}^{\infty} \frac{(-1)^{i+1}}{i} (\mathbf{M} - \mathbf{I}_n)^i.$$

The sum is convergent if  $\mathbf{M}$  is close to identity.

**Rodrigues formulas.** A rotation matrix  $\mathbf{R}$  has an axis represented by a unit vector  $\mathbf{n}$  and an angle  $\alpha$  with respect to this axis. From  $\mathbf{n}$  and  $\alpha$  we can also generate the matrix  $\mathbf{R}$ . The link  $\mathbf{R} \leftrightarrow (\mathbf{n}, \alpha)$  is made by the following *Rodrigues formulas*

$$\begin{aligned} (i) \quad \mathbf{R} &= e^{\alpha \mathbf{n} \wedge} \\ (ii) \quad \alpha \mathbf{n} &= \text{Log}(\mathbf{R}) \end{aligned} \tag{1.5}$$

where

$$\text{Log}(\mathbf{R}) = \wedge^{-1}(\log \mathbf{R})$$

Equation (ii) is the reciprocal of (i). In these formulas,  $\alpha \mathbf{n} \wedge$  is a notation to represent the matrix  $\wedge(\alpha \mathbf{n})$ . As it will be shown in the exercise,

$$\mathbf{R} = e^{\alpha \mathbf{n} \wedge} = \exp(\alpha \mathbf{n} \wedge) = \text{Exp}(\alpha \mathbf{n})$$

is a rotation matrix and  $\mathbf{n}$  is an eigenvector associated with the eigenvalue 1 of  $\mathbf{R}$ . The capitalized version *Exp*, *Log* for *exp*, *log* can be seen as shortcuts to avoid using the  $\wedge$  operator [9][10]. This is illustrated by Figure 1.5.

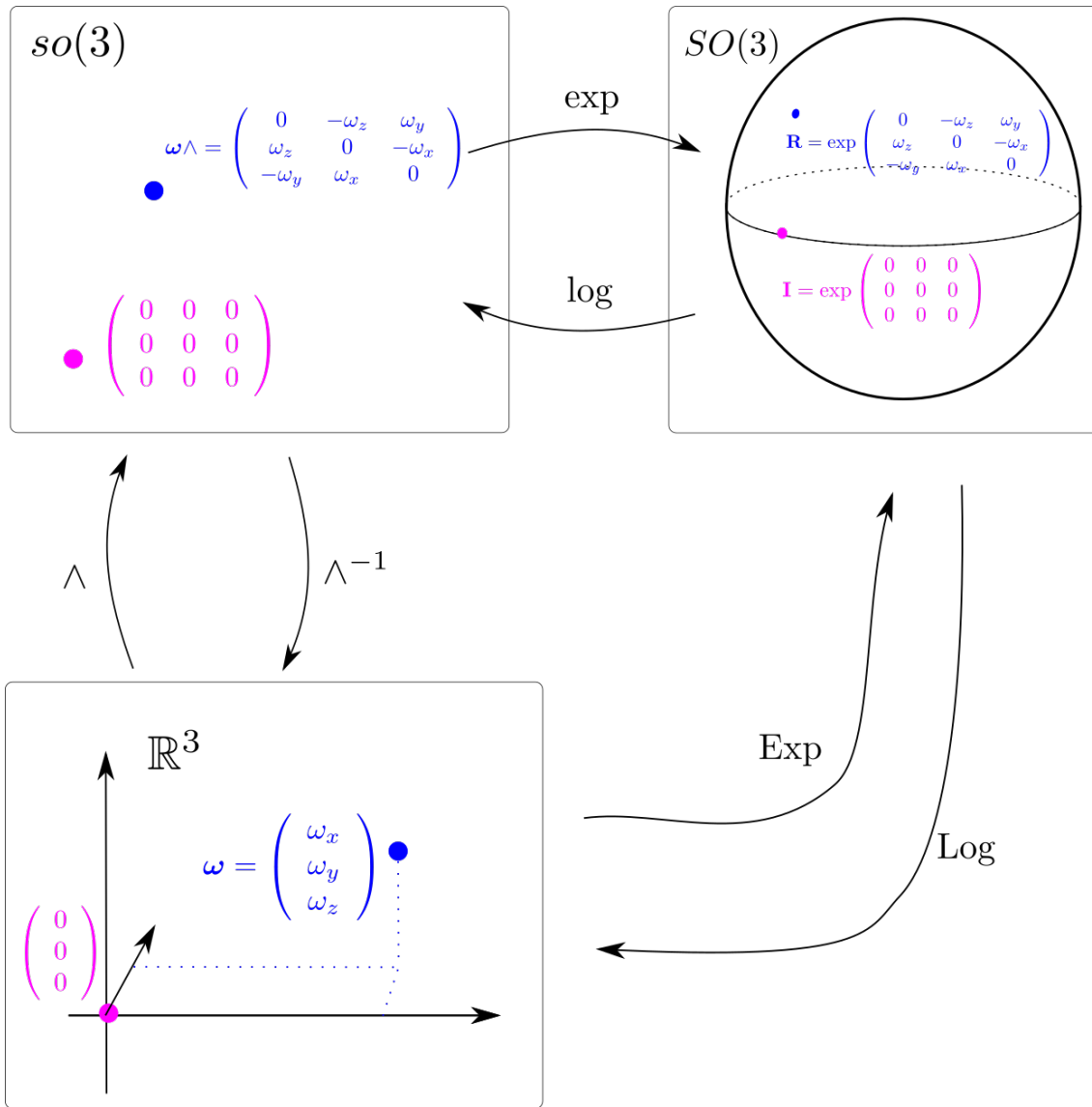


Figure 1.5: Correspondences between the Lie group  $SO(3)$  and the Lie algebra  $so(3)$

The following proposition provides an analytical expression for Log and Exp in the case of  $SO(3)$ .

**Proposition 10.** *Given a rotation matrix  $\mathbf{R}$  of  $SO(3)$ , its logarithm (which is a vector of  $so(3)$ ) is given by*

$$\begin{aligned} \text{Log}\mathbf{R} &= \frac{\alpha}{2 \sin \alpha} \cdot \wedge^{-1} (\mathbf{R} - \mathbf{R}^T) \\ \alpha &= \arccos\left(\frac{\text{tr}(\mathbf{R}) - 1}{2}\right) \end{aligned} \tag{1.6}$$

*Given a vector of  $\omega \in so(3)$ , its exponential (which is a rotation matrix) is given by*

$$\begin{aligned} \text{Exp}\omega &= \mathbf{I} + \frac{\sin \alpha}{\alpha} \cdot \omega \wedge + \frac{1 - \cos \alpha}{\alpha^2} \cdot (\omega \wedge)^2 \\ \alpha &= \|\omega\| \end{aligned}$$

*Proof.* We only prove (1.6). Take a rotation matrix  $\mathbf{R} = e^{\alpha \mathbf{n} \wedge}$  with eigen values  $1, \lambda_1, \lambda_2$  and eigen vectors  $\mathbf{n}, \mathbf{v}_1, \mathbf{v}_2$ . Consider the generalized polynomial  $f(x) = x - x^{-1}$ , where  $x$  is the indeterminate. From the correspondence theorem of eigen values/vectors, the eigen values of  $f(\mathbf{R}) = \mathbf{R} - \mathbf{R}^{-1} = \mathbf{R} - \mathbf{R}^T$  are  $f(1) = 0, f(\lambda_2), f(\lambda_3)$  and the eigen vectors are still  $\mathbf{n}, \mathbf{v}_1, \mathbf{v}_2$ . Now,

$$\begin{aligned} (i) \quad & f(\mathbf{R}) \text{ is skew symmetric} & (f^T(\mathbf{R}) + f(\mathbf{R}) = \mathbf{R}^T - \mathbf{R} + \mathbf{R} - \mathbf{R}^T = \mathbf{0}) \\ (ii) \quad & f(\mathbf{R}) \cdot \mathbf{n} = \mathbf{0} & (\mathbf{n} \text{ is the eigen vector of } f(\mathbf{R}) \text{ associated to } 0) \end{aligned} \quad (1.7)$$

Thus, the two matrices  $f(\mathbf{R})$  and  $(\mathbf{n} \wedge)$  are proportional. As a consequence  $\wedge^{-1}(f(\mathbf{R}))$  provides us the axis of  $\mathbf{R}$ .

To find the angle  $\alpha$ , we use the property that the trace of a matrix is *similarity-invariant*, which means that for any invertible matrix  $\mathbf{P}$ , we have  $\text{tr}(\mathbf{R}) = \text{tr}(\mathbf{P}^{-1} \cdot \mathbf{R} \cdot \mathbf{P})$ . Indeed

$$\text{tr}((\mathbf{P}^{-1} \cdot \mathbf{R}) \cdot \mathbf{P}) = \text{tr}(\mathbf{P} \cdot (\mathbf{P}^{-1} \cdot \mathbf{R})) = \text{tr}(\mathbf{R})$$

Take  $\mathbf{P}$  as the rotation matrix with transforms  $\mathbf{R}$  into a rotation along the first axis. We get

$$\text{tr}(\mathbf{R}) = \text{tr} \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & -\sin \alpha \\ 0 & \sin \alpha & \cos \alpha \end{pmatrix} = 1 + 2 \cos \alpha,$$

which provides us the angle of the rotation. □

## 1.7 Evolution of the rotation matrix

**Proposition 11.** *The evolution of the rotation matrix  $\mathbf{R}(t)$  of a rigid body follows the integration scheme:*

$$\mathbf{R}(t + dt) = \mathbf{R}(t) \cdot \text{Exp}(dt \cdot \boldsymbol{\omega}_r(t)). \quad (1.8)$$

where  $\boldsymbol{\omega}_r = \mathbf{R}^T \boldsymbol{\omega}$  is the rotation vector of the body expressed in its own frame.

*Proof.* Since  $\mathbf{R}^T(t) \dot{\mathbf{R}}(t) = (\boldsymbol{\omega}_r(t) \wedge)$  (see 1.4), we have

$$\dot{\mathbf{R}}(t) = \mathbf{R}(t) \cdot (\boldsymbol{\omega}_r(t) \wedge).$$

If  $dt$  is infinitely small, we get

$$\begin{aligned} \mathbf{R}(t + dt) &= \mathbf{R}(t) + dt \cdot \dot{\mathbf{R}}(t) + o(dt) \\ &= \mathbf{R}(t) + dt \cdot \mathbf{R}(t) \cdot (\boldsymbol{\omega}_r(t) \wedge) + o(dt) \\ &= \mathbf{R}(t) \cdot (\mathbf{I} + dt \cdot (\boldsymbol{\omega}_r(t) \wedge) + o(dt)) \\ &= \mathbf{R}(t) \cdot \exp(dt \cdot \boldsymbol{\omega}_r(t) \wedge) + o(dt) \\ &= \mathbf{R}(t) \cdot \text{Exp}(dt \cdot \boldsymbol{\omega}_r(t)) + o(dt) \end{aligned}$$

□

For numerical reasons, when we integrate for a long time, the matrix  $\mathbf{R}(t)$  may lose the property  $\mathbf{R} \cdot \mathbf{R}^T = \mathbf{I}$ . To avoid this, at each iteration, a normalization step is needed, *i.e.*, the current matrix  $\mathbf{R}$  should be projected on to  $SO(3)$ . For this, we can perform a QR factorization. A QR-decomposition of a square matrix  $\mathbf{M}$  provides two matrices  $\mathbf{Q}, \mathbf{R}$  such that  $\mathbf{M} = \mathbf{Q} \cdot \mathbf{R}$  where  $\mathbf{Q}$  is a rotation matrix and  $\mathbf{R}$  is triangular. When  $\mathbf{M}$  is almost a rotation matrix, then  $\mathbf{R}$  is almost diagonal and also almost a rotation, *i.e.*, on the diagonal of  $\mathbf{R}$  the entries are approximately  $\pm 1$  and outside the entries are almost zeros. The following PYTHON code performs the projection of  $\mathbf{M}$  on  $SO(3)$  and generates the rotation matrix  $\mathbf{M}_2$ .

```
Q,R = numpy.linalg.qr(M)
v=diag(sign(R))
M2=Q@diag(v)
```

## 1.8 Interpolation

Consider a Lie group  $G$  (for instance, the rotations), with two elements  $a, b \in G$ . We would like to interpolate between these elements, according to a parameter  $t \in [0, 1]$ . An interpolation from  $a$  to  $b$  is a function such that:

$$\begin{aligned} f : G \times G \times \mathbb{R} &\rightarrow G \\ f(a, b, 0) &= a \\ f(a, b, 1) &= b \end{aligned}$$

A possible interpolation is:

$$f(a, b, t) = \exp(t \cdot \log(b \cdot a^{-1})) \cdot a$$

or equivalently

$$f(a, b, t) = b^t \cdot a^{1-t}.$$

Note  $f(a, b, t)$  is always on  $G$ , due to the properties of the exponential map. The resulting path is along a geodesic of the manifold  $G$ .

## 1.9 Coordinate system change

Let  $\mathcal{R}_0 : (\mathbf{o}_0, \mathbf{i}_0, \mathbf{j}_0, \mathbf{k}_0)$  and  $\mathcal{R}_1 : (\mathbf{o}_1, \mathbf{i}_1, \mathbf{j}_1, \mathbf{k}_1)$  be two coordinate systems and let  $\mathbf{u}$  be a vector of  $\mathbb{R}^3$  (refer to Figure 1.6). We have the following relation:

$$\begin{aligned} \mathbf{u} &= x_0 \mathbf{i}_0 + y_0 \mathbf{j}_0 + z_0 \mathbf{k}_0 \\ &= x_1 \mathbf{i}_1 + y_1 \mathbf{j}_1 + z_1 \mathbf{k}_1 \end{aligned}$$

where  $(x_0, y_0, z_0)$  and  $(x_1, y_1, z_1)$  are the coordinates of  $\mathbf{u}$  in  $\mathcal{R}_0$  and  $\mathcal{R}_1$ , respectively.

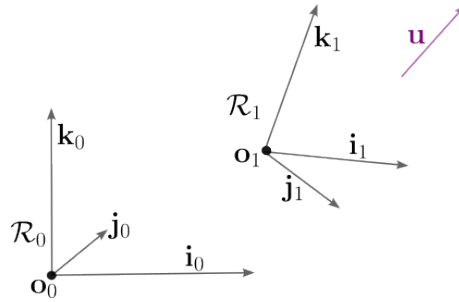


Figure 1.6: Changing the coordinate system  $\mathcal{R}_0$  to the system  $\mathcal{R}_1$

Thus, for any vector  $\mathbf{v}$  we have:

$$\langle x_0\mathbf{i}_0 + y_0\mathbf{j}_0 + z_0\mathbf{k}_0, \mathbf{v} \rangle = \langle x_1\mathbf{i}_1 + y_1\mathbf{j}_1 + z_1\mathbf{k}_1, \mathbf{v} \rangle.$$

By taking respectively  $\mathbf{v} = \mathbf{i}_0, \mathbf{j}_0, \mathbf{k}_0$ , we obtain the following three relations:

$$\begin{cases} \langle x_0\mathbf{i}_0 + y_0\mathbf{j}_0 + z_0\mathbf{k}_0, \mathbf{i}_0 \rangle &= \langle x_1\mathbf{i}_1 + y_1\mathbf{j}_1 + z_1\mathbf{k}_1, \mathbf{i}_0 \rangle \\ \langle x_0\mathbf{i}_0 + y_0\mathbf{j}_0 + z_0\mathbf{k}_0, \mathbf{j}_0 \rangle &= \langle x_1\mathbf{i}_1 + y_1\mathbf{j}_1 + z_1\mathbf{k}_1, \mathbf{j}_0 \rangle \\ \langle x_0\mathbf{i}_0 + y_0\mathbf{j}_0 + z_0\mathbf{k}_0, \mathbf{k}_0 \rangle &= \langle x_1\mathbf{i}_1 + y_1\mathbf{j}_1 + z_1\mathbf{k}_1, \mathbf{k}_0 \rangle \end{cases}$$

However, since the basis  $(\mathbf{i}_0, \mathbf{j}_0, \mathbf{k}_0)$  of  $\mathcal{R}_0$  is orthonormal,  $\langle \mathbf{i}_0, \mathbf{i}_0 \rangle = \langle \mathbf{j}_0, \mathbf{j}_0 \rangle = \langle \mathbf{k}_0, \mathbf{k}_0 \rangle = 1$  and  $\langle \mathbf{i}_0, \mathbf{j}_0 \rangle = \langle \mathbf{j}_0, \mathbf{k}_0 \rangle = \langle \mathbf{i}_0, \mathbf{k}_0 \rangle = 0$ . Thus, these three relations become:

$$\begin{cases} x_0 &= x_1 \cdot \langle \mathbf{i}_1, \mathbf{i}_0 \rangle + y_1 \cdot \langle \mathbf{j}_1, \mathbf{i}_0 \rangle + z_1 \cdot \langle \mathbf{k}_1, \mathbf{i}_0 \rangle \\ y_0 &= x_1 \cdot \langle \mathbf{i}_1, \mathbf{j}_0 \rangle + y_1 \cdot \langle \mathbf{j}_1, \mathbf{j}_0 \rangle + z_1 \cdot \langle \mathbf{k}_1, \mathbf{j}_0 \rangle \\ z_0 &= x_1 \cdot \langle \mathbf{i}_1, \mathbf{k}_0 \rangle + y_1 \cdot \langle \mathbf{j}_1, \mathbf{k}_0 \rangle + z_1 \cdot \langle \mathbf{k}_1, \mathbf{k}_0 \rangle \end{cases}$$

Or in matrix form:

$$\underbrace{\begin{pmatrix} x_0 \\ y_0 \\ z_0 \end{pmatrix}}_{=\mathbf{u}|_{\mathcal{R}_0}} = \underbrace{\begin{pmatrix} \langle \mathbf{i}_1, \mathbf{i}_0 \rangle & \langle \mathbf{j}_1, \mathbf{i}_0 \rangle & \langle \mathbf{k}_1, \mathbf{i}_0 \rangle \\ \langle \mathbf{i}_1, \mathbf{j}_0 \rangle & \langle \mathbf{j}_1, \mathbf{j}_0 \rangle & \langle \mathbf{k}_1, \mathbf{j}_0 \rangle \\ \langle \mathbf{i}_1, \mathbf{k}_0 \rangle & \langle \mathbf{j}_1, \mathbf{k}_0 \rangle & \langle \mathbf{k}_1, \mathbf{k}_0 \rangle \end{pmatrix}}_{=\mathbf{R}_{\mathcal{R}_0}^{\mathcal{R}_1}} \cdot \underbrace{\begin{pmatrix} x_1 \\ y_1 \\ z_1 \end{pmatrix}}_{=\mathbf{u}|_{\mathcal{R}_1}} \tag{1.9}$$

We can see a rotation matrix  $\mathbf{R}_{\mathcal{R}_0}^{\mathcal{R}_1}$  whose columns are the coordinates of  $\mathbf{i}_1, \mathbf{j}_1, \mathbf{k}_1$  expressed in the absolute system  $\mathcal{R}_0$ . Thus

$$\mathbf{R}_{\mathcal{R}_0}^{\mathcal{R}_1} = \left( \begin{array}{c|c|c} \mathbf{i}_1|_{\mathcal{R}_0} & \mathbf{j}_1|_{\mathcal{R}_0} & \mathbf{k}_1|_{\mathcal{R}_0} \end{array} \right)$$

This matrix depends on time and links the frame  $\mathcal{R}_1$  to  $\mathcal{R}_0$ . The matrix  $\mathbf{R}_{\mathcal{R}_0}^{\mathcal{R}_1}$  is often referred to as a *direction cosine matrix* since its components involve the direction cosines of the basis vectors of

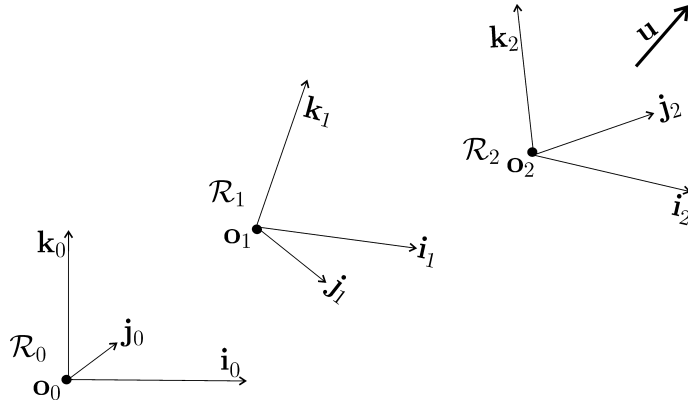


Figure 1.7: Composition of rotations

the two coordinate systems. Likewise, if we would have several systems  $\mathcal{R}_0, \dots, \mathcal{R}_n$  (see Figure 1.7), we would have:

$$\mathbf{u}|_{\mathcal{R}_0} = \mathbf{R}_{\mathcal{R}_0}^{\mathcal{R}_1} \cdot \mathbf{R}_{\mathcal{R}_1}^{\mathcal{R}_2} \cdot \dots \cdot \mathbf{R}_{\mathcal{R}_{n-1}}^{\mathcal{R}_n} \cdot \mathbf{u}|_{\mathcal{R}_n}.$$

**Dead Reckoning.** Let us consider the situation of a robot moving in a three-dimensional environment. Let us call  $\mathcal{R}_0 : (\mathbf{o}_0, \mathbf{i}_0, \mathbf{j}_0, \mathbf{k}_0)$  its reference frame (for example, the frame of the robot at an initial time). The position of the robot is represented by the vector  $\mathbf{p}(t)$  expressed in  $\mathcal{R}_0$  and its attitude (*i.e.*, its orientation) by the rotation matrix  $\mathbf{R}(t)$  which represents the coordinates of the vectors  $\mathbf{i}_1, \mathbf{j}_1, \mathbf{k}_1$  of the coordinate system  $\mathcal{R}_1$  of the robot expressed in the coordinate system  $\mathcal{R}_0$ , at time  $t$ . It follows that:

$$\mathbf{R}(t) = \left( \begin{array}{c|c|c} \mathbf{i}_1|_{\mathcal{R}_0} & \mathbf{j}_1|_{\mathcal{R}_0} & \mathbf{k}_1|_{\mathcal{R}_0} \\ \hline \hline \hline \end{array} \right) = \mathbf{R}_{\mathcal{R}_0}^{\mathcal{R}_1}(t).$$

This matrix can be returned by a precise attitude unit positioned on the robot. If the robot is also equipped with a *Doppler Velocity Log* (or DVL) which provides it with its speed vector  $\mathbf{v}_r$  relative to the ground or the seabed, expressed in the coordinate system  $\mathcal{R}_1$  of the robot, then the speed vector  $\mathbf{v}$  of the robot satisfies:

$$\underbrace{\mathbf{v}|_{\mathcal{R}_0}}_{\dot{\mathbf{p}}(t)} \stackrel{(1.9)}{=} \underbrace{\mathbf{R}_{\mathcal{R}_0}^{\mathcal{R}_1}}_{\mathbf{R}(t)} \cdot \underbrace{\mathbf{v}|_{\mathcal{R}_1}}_{\mathbf{v}_r(t)}$$

Or equivalently

$$\dot{\mathbf{p}}(t) = \mathbf{R}(t) \cdot \mathbf{v}_r(t). \quad (1.10)$$

*Dead reckoning* consists of integrating this state equation from the knowledge of  $\mathbf{R}(t)$  and  $\mathbf{v}_r(t)$ .

# Exercises

EXERCISE 1.– *Properties of the matrix  $\boldsymbol{\omega} \wedge$*

See the correction video at <https://youtu.be/D3DbfurFWXo>

Let us consider the vector  $\boldsymbol{\omega} = (\omega_x, \omega_y, \omega_z)$  and its associated skew symmetric matrix

$$\wedge(\boldsymbol{\omega}) = \begin{pmatrix} 0 & -\omega_z & \omega_y \\ \omega_z & 0 & -\omega_x \\ -\omega_y & \omega_x & 0 \end{pmatrix}$$

- 1) Show that the eigenvalues of  $\wedge(\boldsymbol{\omega})$  are  $\{0, \|\boldsymbol{\omega}\| \cdot i, -\|\boldsymbol{\omega}\| \cdot i\}$ . Give an eigenvector associated with 0. Discuss.
  - 2) Show that the vector  $\wedge(\boldsymbol{\omega}) \mathbf{x} = \boldsymbol{\omega} \wedge \mathbf{x}$  is a vector perpendicular to  $\boldsymbol{\omega}$  and  $\mathbf{x}$ , such that the trihedron  $(\boldsymbol{\omega}, \mathbf{x}, \boldsymbol{\omega} \wedge \mathbf{x})$  is direct.
  - 3) Show that the norm of  $\boldsymbol{\omega} \wedge \mathbf{x}$  is the surface of the parallelogram  $\mathcal{A}$  mediated by  $\boldsymbol{\omega}$  and  $\mathbf{x}$ .
- 

EXERCISE 2.– *Jacobi identity*

See the correction video at <https://youtu.be/gD04pW7iYBw>

The Jacobi identity for the vector product in  $\mathbb{R}^3$  is written as:

$$\mathbf{a} \wedge (\mathbf{b} \wedge \mathbf{c}) + \mathbf{c} \wedge (\mathbf{a} \wedge \mathbf{b}) + \mathbf{b} \wedge (\mathbf{c} \wedge \mathbf{a}) = \mathbf{0}.$$

- 1) Show that this identity is equivalent to:

$$\wedge(\mathbf{a} \wedge \mathbf{b}) = \wedge(\mathbf{a}) \cdot \wedge(\mathbf{b}) - \wedge(\mathbf{b}) \cdot \wedge(\mathbf{a})$$

where  $\wedge(\boldsymbol{\omega})$  is the skew symmetric matrix associated to the vector  $\boldsymbol{\omega} \in \mathbb{R}^3$ .

- 2) In the space of skew-symmetric matrices, the Lie bracket is defined as follows:

$$[\mathbf{A}, \mathbf{B}] = \mathbf{A} \cdot \mathbf{B} - \mathbf{B} \cdot \mathbf{A}.$$

Show that :

$$\wedge(\mathbf{a} \wedge \mathbf{b}) = [\wedge(\mathbf{a}), \wedge(\mathbf{b})].$$

3) Recall that the infinitesimal rotation following the skew-symmetric matrix  $\mathbf{A}$  is  $\exp(\mathbf{A} \cdot dt) \simeq \mathbf{I} + \mathbf{A} \cdot dt + \frac{\mathbf{A}^2}{2} \cdot dt^2 + o(dt^2)$ . Consider two skew-symmetric matrices  $\mathbf{A}, \mathbf{B}$  and define the two infinitesimal rotation matrices

$$\mathbf{R}_a = e^{\mathbf{A} \cdot dt} \text{ and } \mathbf{R}_b = e^{\mathbf{B} \cdot dt}.$$

Compute a Taylor expansion of second order associated to the rotation given by

$$\mathbf{R}_a^{-1} \mathbf{R}_b^{-1} \mathbf{R}_a \mathbf{R}_b.$$

From this result, give an interpretation of the Lie bracket  $[\mathbf{A}, \mathbf{B}] = \mathbf{AB} - \mathbf{BA}$ .

4) Verify that the set  $(\mathbb{R}^3, +, \wedge, \cdot)$  forms a *Lie algebra*.

---

### EXERCISE 3.— *Varignon's formula*

See the correction video at [https://youtu.be/dMhwZLTA\\_cA](https://youtu.be/dMhwZLTA_cA)

Let us consider a solid body whose center of gravity remains at the origin of a Galilean coordinate system and is rotating around an axis  $\Delta$  with a rotation vector of  $\boldsymbol{\omega}$ . Give the equation of the trajectory of a point  $\mathbf{x}$  of the body.

---

### EXERCISE 4.— *Quaternions*

See the correction video at [https://youtu.be/d5dRd\\_SjDTU](https://youtu.be/d5dRd_SjDTU)

Quaternions were discovered by W. R. Hamilton and are also used to represent a rotation in a 3-dimensional space (see, e.g., [11]). Quaternions can be seen as an extension of the complex numbers. A quaternion  $\hat{q}$  corresponds to a scalar  $s$  plus a vector  $\mathbf{v}$  of  $\mathbb{R}^3$ . We use the equivalent following notations:

$$\hat{q} = s + v_1 i + v_2 j + v_3 k = s + \langle v_1, v_2, v_3 \rangle = s + \langle \mathbf{v} \rangle,$$

where

$$i^2 = j^2 = k^2 = ijk = -1.$$

1) From these relations, fill the following multiplication table:

$\cdot$	1	$i$	$j$	$k$
1	1	$i$	$j$	$k$
$i$	$i$	?	?	?
$j$	$j$	?	?	?
$k$	$k$	?	?	?

2) A unit-quaternion  $\mathring{q}$  is a quaternion with a unit magnitude:

$$|\mathring{q}|^2 = s^2 + v_1^2 + v_2^2 + v_3^2 = 1.$$

Consider the rotation obtained by a rotation around the unit vector  $\mathbf{v}$  with an angle  $\theta$ . From the Rodrigues formula, we know that the corresponding rotation matrix is  $\exp(\theta \cdot \mathbf{v} \wedge)$ . To this rotation, we associate the quaternion:

$$\mathring{q} = \cos\left(\frac{\theta}{2}\right) + \sin\left(\frac{\theta}{2}\right) \cdot \langle \mathbf{v} \rangle .$$

Check that this quaternion has a unit magnitude.

3) Show that the quaternion  $\mathring{q}$  and its opposite  $-\mathring{q}$ , both correspond to the same rotation in  $\mathbb{R}^3$ .

4) Assuming that the composition of rotations corresponds to a multiplication of quaternions, show that

$$(s + \langle \mathbf{v} \rangle)^{-1} = s + \langle -\mathbf{v} \rangle .$$

5) We consider a rotation  $\mathcal{R}$  composed of 3 rotations. The first rotation is along  $\mathbf{i}$  with an angle  $\varphi = \frac{\pi}{2}$ , the second rotation is along  $\mathbf{j}$  with an angle  $\theta = \frac{\pi}{2}$ , the third rotation is along  $\mathbf{k}$  with an angle  $\psi = \frac{\pi}{2}$ . The transformation  $\mathcal{R}$  corresponds to a rotation around the unit vector  $\mathbf{v}$  with an angle  $\alpha$ . Give the values of  $\mathbf{v}$  and  $\alpha$  using 3 methods: the geometrical method (using hands only), the matrices and the quaternions.

EXERCISE 5.- (*Lie group  $SE(2)$* )

See the correction video at <https://youtu.be/vQWkOBpBdhk>

The group of 2D rigid transformations of the plane is named  $SE(2)$ . The classical matrix representation is the following:

$$\mathbf{P} = \begin{pmatrix} \cos p_3 & -\sin p_3 & p_1 \\ \sin p_3 & \cos p_3 & p_2 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} \mathbf{R} & \mathbf{t} \\ \mathbf{0} & 1 \end{pmatrix}$$

where  $\mathbf{t} = (p_1, p_2)$  is the translation and  $p_3$  is the angle for the rotation  $\mathbf{R}$ .

1) Show that  $SE(2)$  as defined above is a Lie group.

2) Using the derivatives at  $\mathbf{p} = \mathbf{0}$ , find the associated Lie algebra.

3) We consider the Dubins car

$$\begin{cases} \dot{x}_1 = u_1 \cdot \cos x_3 \\ \dot{x}_2 = u_1 \cdot \sin x_3 \\ \dot{x}_3 = u_2 \end{cases}$$

The state vector  $\mathbf{x} = (x_1, x_2, x_3)$  is a pose and can be represented by an element of  $\mathbf{P} \in SE(2)$ . For the integration, we should apply a formula similar to  $\mathbf{R}(t + dt) = \mathbf{R}(t) \cdot \text{Exp}(dt \cdot \boldsymbol{\omega}_r(t))$  (see equation 1.8) we had in  $SO(3)$ . Give the corresponding formula for  $SE(2)$ .

4) Assume that

$$\begin{aligned} u_1(t) &= 1 \\ u_2(t) &= 1 \\ \mathbf{x}(0) &= (1, 0, \frac{\pi}{2})^T \end{aligned}$$

Write a program which integrates the trajectory with three different approaches: exact, Euler and the exponential integration scheme and compare. We will take  $t \in [0, 6]$  and a sampling time equal to  $dt = 0.5$ .

5) Using the interpolation formula, find a path  $\mathbf{P}(t), t \in [0, 1]$  in  $SE(2)$  such that  $\mathbf{P}(0) = \mathbf{P}_0, \mathbf{P}(1) = \mathbf{P}_1$ , where  $\mathbf{P}_0$  and  $\mathbf{P}_1$  are the pose matrices associated to  $\mathbf{x}(0) = (-1, -1, -1)^T$  and  $\mathbf{x}(1) = (1, 1, 1)^T$ , respectively.

#### EXERCISE 6.– *Amphisbaena*

See the correction video at <https://youtu.be/DFAZUytmTk0>

The real projective space, denoted by  $\mathbb{RP}^n$ , is the topological space of lines passing through the origin  $\mathbf{0}$  in  $\mathbb{R}^{n+1}$ . The set  $\mathbb{RP}^1$  is thus the set space of lines of the plane passing through  $\mathbf{0}$  (see Figure 1.8). To reach line, we can associate the angle  $\theta$  with the positive  $x$ -axis.

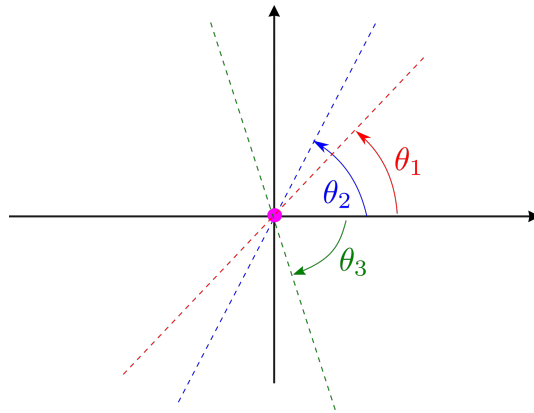


Figure 1.8: The product of the red line with the blue line corresponds to the green line

1) Show that a matrix representation for  $\mathbb{RP}^1$  is:

$$\mathbf{P}(\theta) = \begin{pmatrix} \cos 2\theta & -\sin 2\theta \\ \sin 2\theta & \cos 2\theta \end{pmatrix}$$

where the parameter  $\theta \in ]-\frac{\pi}{2}, \frac{\pi}{2}]$  is the angle of the line.

2) Show that  $\mathbb{RP}^1$  is a Lie group for the matrix multiplication.

3) Using the derivatives at  $\theta = 0$ , find the associated Lie algebra.

4) We consider the amphisbaena robot represented by Figure 1.9. It is fully symmetric, which means that if we rotate the robot by  $\pi$ , we get exactly the same robot. Its state  $\theta$  is thus a pose which can be represented by an element of  $\mathbf{P} \in \mathbb{RP}^1$ . The parameter  $\theta$  should be an element of the interval  $] -\frac{\pi}{2}, \frac{\pi}{2}]$ . We assume that the robot's dynamics is described by  $\dot{\theta} = u$ . Propose a simulation for the system. The integration has to be done in the Lie group.

5) Give a proportional controller so that our system points toward a specific direction  $\bar{\theta}$ . Illustrate the behavior of the controller by a simulation.

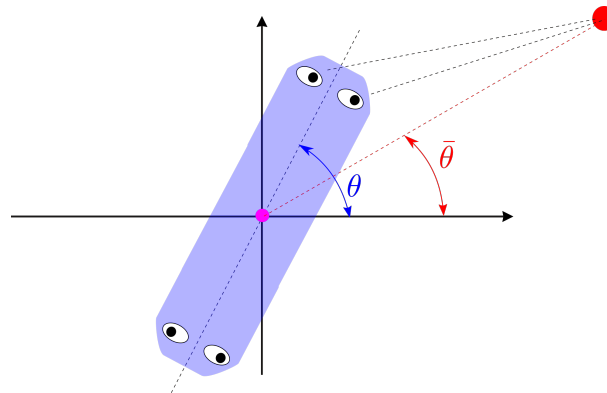


Figure 1.9: The amphisbaena robot has two heads. It has to look toward the target  $\bar{\theta}$  (red point) with its nearest head

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### EXERCISE 7.– *Car on the sphere*

See the correction video at <https://youtu.be/ipgs4DBQosk>

Consider a car moving in the plane described by the state equations:

$$\begin{cases} \dot{x}_1 = x_4 \cos x_3 \\ \dot{x}_2 = x_4 \sin x_3 \\ \dot{x}_3 = u_1 \\ \dot{x}_4 = u_2 \end{cases}$$

where  $(x_1, x_2, x_3)$  is the pose of the car,  $x_4$  is the speed,  $u_1$  is the rotation rate and  $u_2$  is the acceleration (see 1.10, left).

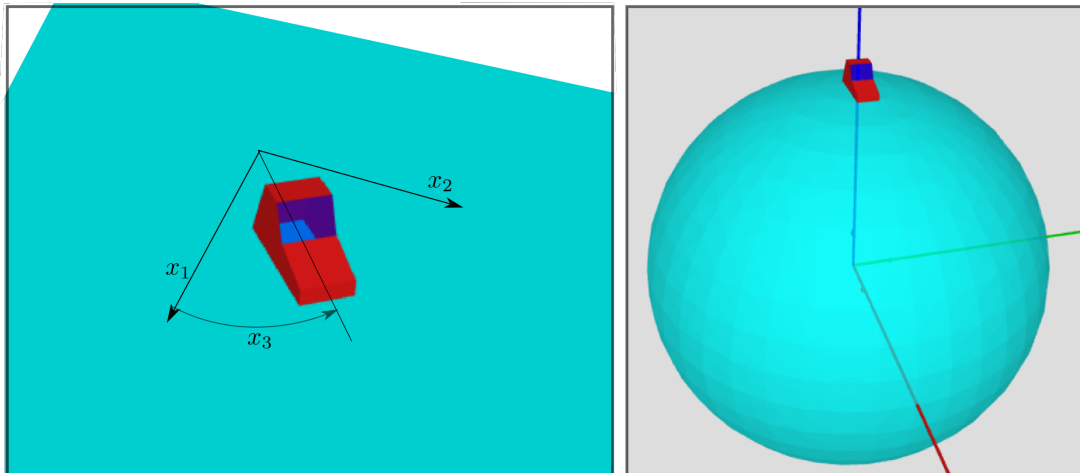


Figure 1.10: Left: car on the plane; Right: car on the sphere

For the input, we choose

$$\mathbf{u} = \begin{pmatrix} 0.1 \cdot \cos(0.1 \cdot t) + 0.02 \\ 1 - x_4 \end{pmatrix}$$

and for the initial state, we take  $\mathbf{x} = (0, 0, 0, 0)$ .

1) What is the shape of the manifold corresponding to the state space. Simulate the system for  $t \in [0, 200]$ .

2) We take the same car, with the same input  $\mathbf{u}$ . But now, the car moves on a sphere with radius  $r = 20$  as illustrated by Figure 1.10, right. What is the shape of the state-space?

3) Provide a simulation of the car moving on the sphere. The simulation should avoid the singularities.

# Chapter 2

## Euler angles

### 2.1 Definition

In related literature, the angles proposed by Euler in 1770 to represent the orientation of solid bodies in space are not uniquely defined. We mainly distinguish between the roll-yaw-roll, roll-pitch-roll and roll-pitch-yaw formulations. It is the latter that we will choose since it is imposed in the mobile robotics. Within this roll-pitch-yaw formulation, the Euler angles are sometimes referred to as *Cardan angles*. Any rotation matrix of  $\mathbb{R}^3$  can be expressed in the form of the product of three matrices as follows:

$$\begin{aligned}
 \mathbf{R}(\varphi, \theta, \psi) &= e^{\psi \mathbf{k}^\wedge} \cdot e^{\theta \mathbf{j}^\wedge} \cdot e^{\varphi \mathbf{i}^\wedge} \\
 &= \exp \begin{pmatrix} 0 & -\psi & 0 \\ \psi & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \cdot \exp \begin{pmatrix} 0 & 0 & \theta \\ 0 & 0 & 0 \\ -\theta & 0 & 0 \end{pmatrix} \cdot \exp \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & -\varphi \\ 0 & \varphi & 0 \end{pmatrix} \\
 &= \underbrace{\begin{pmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{pmatrix}}_{\mathbf{R}_\psi} \cdot \underbrace{\begin{pmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{pmatrix}}_{\mathbf{R}_\theta} \cdot \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \varphi & -\sin \varphi \\ 0 & \sin \varphi & \cos \varphi \end{pmatrix}}_{\mathbf{R}_\varphi}
 \end{aligned}$$

where  $\mathbf{i} = (1, 0, 0)^\top$ ,  $\mathbf{j} = (0, 1, 0)^\top$ ,  $\mathbf{k} = (0, 0, 1)^\top$ .

This formula makes appear the Lie algebra generators:

$$\mathbf{i}^\wedge = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{pmatrix}, \quad \mathbf{j}^\wedge = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ -1 & 0 & 0 \end{pmatrix}, \quad \mathbf{k}^\wedge = \begin{pmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

which is a basis a the space of skew-symmetric matrices of  $\mathbb{R}^3$ .

In developed form, we get that  $\mathbf{R}(\varphi, \theta, \psi)$  is

$$\begin{pmatrix} \underbrace{\cos \theta \cos \psi}_{\mathbf{i}_1|\mathcal{R}_0} & -\cos \varphi \sin \psi + \sin \theta \cos \psi \sin \varphi & \sin \psi \sin \varphi + \sin \theta \cos \psi \cos \varphi \\ \cos \theta \sin \psi & \underbrace{\cos \psi \cos \varphi + \sin \theta \sin \psi \sin \varphi}_{\mathbf{j}_1|\mathcal{R}_0} & -\cos \psi \sin \varphi + \sin \theta \cos \varphi \sin \psi \\ -\sin \theta & \cos \theta \sin \varphi & \underbrace{\cos \theta \cos \varphi}_{\mathbf{k}_1|\mathcal{R}_0} \end{pmatrix} \quad (2.1)$$

We obtained this expression using the following Sympy script:

```
from sympy import *
def Reuler(phi, theta, psi):
    Lphi = Matrix([[0, 0, 0], [0, 0, -1], [0, 1, 0]])
    Ltheta = Matrix([[0, 0, 1], [0, 0, 0], [-1, 0, 0]])
    Lpsi = Matrix([[0, -1, 0], [1, 0, 0], [0, 0, 0]])
    return exp(phi*Lphi)*exp(theta*Ltheta)*exp(psi*Lpsi)
phi, theta, psi = symbols('phi theta psi')
print(Reuler(phi, theta, psi))
```

The angles  $\varphi, \theta, \psi$  are the *Euler angles* and are called the *bank*, the *elevation* and the *heading*, respectively. The terms *roll*, *pitch*, *yaw* are often employed, although they correspond, respectively, to variations of bank, elevation and heading.

*Remark 12.* The same reasoning could have been done in  $\mathbb{R}^n$  for  $n \neq 3$ . For instance, for  $n = 2$ , we have

$$\mathbf{R}(\theta) = \exp \begin{pmatrix} 0 & -\theta \\ \theta & 0 \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

and for  $n = 4$ , we get

$$\begin{aligned} \mathbf{R}(\theta_1, \dots, \theta_6) &= \exp \begin{pmatrix} 0 & -\theta_1 & 0 & 0 \\ \theta_1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \cdot \exp \begin{pmatrix} 0 & 0 & -\theta_2 & 0 \\ 0 & 0 & 0 & 0 \\ \theta_2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \cdots \exp \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -\theta_6 \\ 0 & 0 & \theta_6 & 0 \end{pmatrix} \\ &= \begin{pmatrix} \cos \theta_1 & -\sin \theta_1 & 0 & 0 \\ \sin \theta_1 & \cos \theta_1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \cdots \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \cos \theta_6 & -\sin \theta_6 \\ 0 & 0 & \sin \theta_6 & \cos \theta_6 \end{pmatrix} \end{aligned}$$

## 2.2 Gimbal lock

When  $\theta = \frac{\pi}{2}$  (the same effect happens as soon as  $\cos\theta = 0$ ), we have

$$\begin{aligned}
\mathbf{R}(\varphi, \theta, \psi) &= \begin{pmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{pmatrix} \cdot \underbrace{\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & 0 & 0 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \varphi & -\sin \varphi \\ 0 & \sin \varphi & \cos \varphi \end{pmatrix}} \\
&= \begin{pmatrix} 0 & \sin \varphi & \cos \varphi \\ 0 & \cos \varphi & -\sin \varphi \\ -1 & 0 & 0 \end{pmatrix} \\
&= \begin{pmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{pmatrix} \cdot \underbrace{\begin{pmatrix} \cos \varphi & \sin \varphi & 0 \\ -\sin \varphi & \cos \varphi & 0 \\ 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & 0 & 0 \end{pmatrix}} \\
&= \begin{pmatrix} \cos(\psi - \varphi) & -\sin(\psi - \varphi) & 0 \\ \sin(\psi - \varphi) & \cos(\psi - \varphi) & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & 0 & 0 \end{pmatrix} \\
&= \begin{pmatrix} 0 & -\sin(\psi - \varphi) & \cos(\psi - \varphi) \\ 0 & \cos(\psi - \varphi) & \sin(\psi - \varphi) \\ -1 & 0 & 0 \end{pmatrix}
\end{aligned}$$

and thus,

$$\frac{d\mathbf{R}}{d\psi} = -\frac{d\mathbf{R}}{d\varphi}.$$

This corresponds to a singularity which tells us that when  $\theta = \frac{\pi}{2}$ , we cannot move on the manifold of rotation matrices  $SO(3)$ , in all directions using the Euler angles. Equivalently, this means that some trajectories  $\mathbf{R}(t)$  cannot be followed by Euler angles.

## 2.3 Rotation matrix to Euler angles

Given a rotation matrix  $\mathbf{R}$ , we can easily find the three Euler angles by solving, following Equation (2.1), the equations:

$$\begin{cases} -\sin \theta = r_{31} \\ \cos \theta \sin \varphi = r_{32} & \cos \theta \cos \varphi = r_{33} \\ \cos \theta \cos \psi = r_{11} & \cos \theta \sin \psi = r_{21} \end{cases}$$

By imposing  $\theta \in [-\frac{\pi}{2}, \frac{\pi}{2}]$ ,  $\varphi \in [-\pi, \pi]$ ,  $\psi \in [-\pi, \pi]$ , we find:

$$\begin{aligned}
\varphi &= \text{atan2}(r_{32}, r_{33}) \\
\theta &= -\arcsin r_{31} \\
\psi &= \text{atan2}(r_{21}, r_{11})
\end{aligned} \tag{2.2}$$

Here  $\text{atan2}$  is the two-argument arctangent function defined by

$$\alpha = \text{atan2}(y, x) \Leftrightarrow \alpha \in ]-\pi, \pi] \text{ and } \exists r > 0 \mid \begin{cases} x = r \cos \alpha \\ y = r \sin \alpha \end{cases} \quad (2.3)$$

## Application

Assume that we have a robot, for instance a boat, which is assumed to be static as represented by Figure 2.1. The world frame  $\mathcal{R}_0$  is such that  $Ox$  is oriented toward the North and  $Oz$  toward the sky. We want to estimate the Euler angles  $\varphi, \theta, \psi$  of the robot from the magnetic field  $\mathbf{y}_1$  and its acceleration  $\mathbf{a}_1$ , both measured in the robot frame  $\mathcal{R}_1$  (see Figure 2.1).

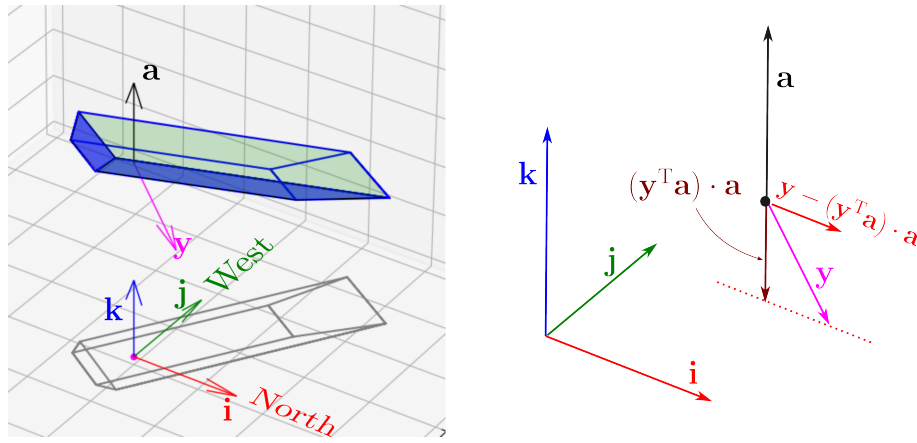


Figure 2.1: The boat has to find its Euler angles from the accelerometer and the magnetometer

For simplicity, we assume that  $\mathbf{y}$  and  $\mathbf{a}$  have been normalized, *i.e.*, their norm are both equal to 1. The frame  $\mathcal{R}_0$  is defined by the three vectors.  $\mathbf{i}, \mathbf{j}, \mathbf{k}$ . If we express these vector in  $\mathcal{R}_0$ , we get:

$$\mathbf{i}_0 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{j}_0 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \quad \mathbf{k}_0 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

In  $\mathcal{R}_0$ , we have

$$\mathbf{y}_0 = \begin{pmatrix} \cos I \\ 0 \\ -\sin I \end{pmatrix}, \quad \mathbf{a}_0 = \mathbf{k}_0 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

where  $I$  is the *magnetic dip* (or *inclination*), *i.e.*, the angle between the magnetic field and the vertical. From the knowledge of  $\mathbf{a}$  and  $\mathbf{y}$  we can build the vectors  $\mathbf{i}, \mathbf{j}, \mathbf{k}$ . Indeed

$$\begin{aligned} \mathbf{i} &= \frac{\mathbf{y} - (\mathbf{y}^T \mathbf{a}) \cdot \mathbf{a}}{\|\mathbf{y} - (\mathbf{y}^T \mathbf{a}) \cdot \mathbf{a}\|} \\ \mathbf{j} &= \mathbf{k} \wedge \mathbf{i} = \mathbf{a} \wedge \frac{\mathbf{y} - (\mathbf{y}^T \mathbf{a}) \cdot \mathbf{a}}{\|\mathbf{y} - (\mathbf{y}^T \mathbf{a}) \cdot \mathbf{a}\|} \\ \mathbf{k} &= \mathbf{a} \end{aligned}$$

Denote by  $\mathbf{R}$  the orientation matrix of the robot. We have

$$\begin{pmatrix} \mathbf{i}_0 & | & \mathbf{j}_0 & | & \mathbf{k}_0 \end{pmatrix} = \mathbf{R} \cdot \begin{pmatrix} \mathbf{i}_1 & | & \mathbf{j}_1 & | & \mathbf{k}_1 \end{pmatrix},$$

*i.e.*,

$$\mathbf{R} = \begin{pmatrix} \mathbf{i}_1 & | & \mathbf{j}_1 & | & \mathbf{k}_1 \end{pmatrix}^T.$$

The orientation matrix of the robot is thus

$$\mathbf{R} = \left( \frac{\mathbf{y}_1 - (\mathbf{y}_1^T \mathbf{a}_1) \cdot \mathbf{a}_1}{\|\mathbf{y}_1 - (\mathbf{y}_1^T \mathbf{a}_1) \cdot \mathbf{a}_1\|} \mid \mathbf{a}_1 \wedge \frac{\mathbf{y}_1 - (\mathbf{y}_1^T \mathbf{a}_1) \cdot \mathbf{a}_1}{\|\mathbf{y}_1 - (\mathbf{y}_1^T \mathbf{a}_1) \cdot \mathbf{a}_1\|} \mid \mathbf{a}_1 \right)^T$$

From (2.2), we get the Euler angles.

## 2.4 Rotation vector of a moving Euler matrix

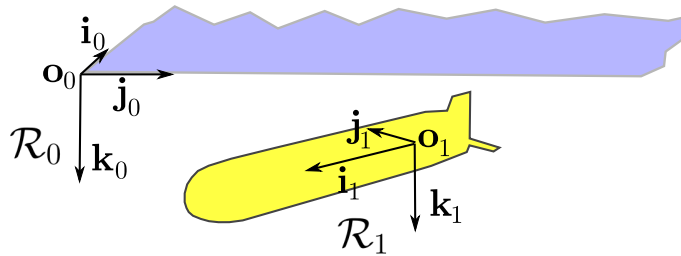


Figure 2.2: The coordinate system  $\mathcal{R}_1 : (\mathbf{o}_1, \mathbf{i}_1, \mathbf{j}_1, \mathbf{k}_1)$  attached to the robot

Let us consider a solid body moving in a coordinate system  $\mathcal{R}_0$  and a coordinate system  $\mathcal{R}_1$  attached to this body (refer to Figure 2.2). The conventions chosen here are those of the SNAME (*Society of Naval and Marine Engineers*). The two coordinate systems are assumed to be orthonormal. Let  $\mathbf{R}(t) = \mathbf{R}(\varphi(t), \theta(t), \psi(t))$  be the rotation matrix that links the two systems. We need to find the instantaneous rotation vector  $\boldsymbol{\omega}$  of the solid body relative to  $\mathcal{R}_0$  in function of  $\varphi, \theta, \psi, \dot{\varphi}, \dot{\theta}, \dot{\psi}$ .

We have

$$\boldsymbol{\omega}_r = \boldsymbol{\omega}_{|\mathcal{R}_1} = \wedge^{-1} \left( \mathbf{R}^T \cdot \dot{\mathbf{R}} \right)$$

where  $\mathbf{R} = \mathbf{R}_{\text{euler}}(\varphi, \theta, \psi)$  is given by (2.1). Set  $\mathbf{q} = (\varphi, \theta, \psi)$ . We have

$$\begin{aligned} \boldsymbol{\omega}_r(\mathbf{q}, \dot{\mathbf{q}}) &= \wedge^{-1} \left( \mathbf{R}(\mathbf{q})^T \cdot \frac{d}{dt} \mathbf{R}(\mathbf{q}) \right) \\ &= \wedge^{-1} \left( \mathbf{R}(\mathbf{q})^T \cdot \frac{\partial \mathbf{R}(\mathbf{q})}{\partial \mathbf{q}} \cdot \dot{\mathbf{q}} \right) \end{aligned}$$

Note that  $\boldsymbol{\omega}_r$  depends linearly of  $\dot{\mathbf{q}}$ , *i.e.*, it as the form

$$\boldsymbol{\omega}_r(\mathbf{q}, \dot{\mathbf{q}}) = \mathbf{A}(\mathbf{q}) \cdot \dot{\mathbf{q}}$$

By differentiating with respect to  $\dot{\mathbf{q}}$ , we get

$$\frac{\partial \boldsymbol{\omega}_r(\mathbf{q}, \dot{\mathbf{q}})}{\partial \dot{\mathbf{q}}} = \mathbf{A}(\mathbf{q}).$$

Thus we have

$$\boldsymbol{\omega}_r(\mathbf{q}, \dot{\mathbf{q}}) = \frac{\partial \boldsymbol{\omega}_r(\mathbf{q}, \dot{\mathbf{q}})}{\partial \dot{\mathbf{q}}} \cdot \dot{\mathbf{q}}$$

or equivalently,

$$\begin{aligned} \dot{\mathbf{q}} &= \left( \frac{\partial \boldsymbol{\omega}_r(\mathbf{q}, \dot{\mathbf{q}})}{\partial \dot{\mathbf{q}}} \right)^{-1} \cdot \boldsymbol{\omega}_r \\ &= \left( \frac{\partial}{\partial \dot{\mathbf{q}}} \left( \wedge^{-1} \left( \mathbf{R}(\mathbf{q})^T \cdot \frac{\partial \mathbf{R}(\mathbf{q})}{\partial \mathbf{q}} \cdot \dot{\mathbf{q}} \right) \right) \right)^{-1} \cdot \boldsymbol{\omega}_r. \end{aligned}$$

Taking into account the expression of  $\mathbf{R}(\mathbf{q}) = \mathbf{R}_{\text{euler}}(\varphi, \theta, \psi)$ , given by (2.1), we get

$$\underbrace{\begin{pmatrix} \dot{\varphi} \\ \dot{\theta} \\ \dot{\psi} \end{pmatrix}}_{\dot{\mathbf{q}}} = \underbrace{\begin{pmatrix} 1 & \tan \theta \sin \varphi & \tan \theta \cos \varphi \\ 0 & \cos \varphi & -\sin \varphi \\ 0 & \frac{\sin \varphi}{\cos \theta} & \frac{\cos \varphi}{\cos \theta} \end{pmatrix}}_{\left( \frac{\partial \boldsymbol{\omega}_r}{\partial \dot{\mathbf{q}}} \right)^{-1}} \cdot \boldsymbol{\omega}_r. \quad (2.4)$$

The expression for  $\left( \frac{\partial \boldsymbol{\omega}_r}{\partial \dot{\mathbf{q}}} \right)^{-1}$  in (2.4) could be obtained using Sympy by the following script:

```
t = symbols('t')
phi, theta, psi = Function('phi')(t), Function('theta')(t), Function('psi')(t)
dphi, dtheta, dpsi = Function('dphi')(t), Function('dtheta')(t), Function('dpsi')(t)
R=Reuler(phi, theta, psi)
W=Transpose(R)*diff(R, t)
wr=Matrix([[ -W[1,2]], [W[0,2]], [ -W[0,1]]])
wr=wr.subs({diff(phi, t):dphi, diff(theta, t):dtheta, diff(psi, t):dpsi})
wr=simplify(wr)
dq=[dphi, dtheta, dpsi]
print(simplify(wr.jacobian(dq).inv()))
```

We observe a singularity if  $\cos \theta = 0$ . We will therefore take care to never have an elevation  $\theta$  equal to  $\pm \frac{\pi}{2}$ , to avoid the gimbal lock.

# Exercises

---

## EXERCISE 8.– *Heading of a boat*

See the correction video at <https://youtu.be/riJZseoavFk>

We consider a boat moving in the waves. We want to estimate the Euler angles (and more particularly the heading) from a magnetometer and an accelerometer. We assume that the accelerometer returns the gravity in the robot frame, *i.e.*, the accelerations of the boat are small compare to the gravity  $9.81\text{ms}^{-2}$ . The *magnetic dip* (or *inclination*) is the angle  $I$  between the magnetic field and the vertical. The range of dip is from  $-\frac{\pi}{2}\text{rad}$  (at the South Magnetic Pole) and  $\frac{\pi}{2}\text{rad}$  (at the North Magnetic Pole). It is given approximately given by the formula

$$I = \arctan(2\tan\lambda)$$

where  $\lambda$  is the latitude. For instance, in Brest, we have  $\lambda = 48.39^\circ$ . Thus we get  $I = 66^\circ = 1.15\text{rad}$ . We consider a boat near Brest which is assumed to be static as represented by Figure 2.3. The world frame  $\mathcal{R}_0$  we consider is centered in Brest and is such that  $Ox$  is oriented toward the North and  $Oz$  toward the sky.

Assume that the boat has the orientation  $\varphi, \theta, \psi$  that are unknown. These Euler angles have to be estimated from the magnetic field  $\mathbf{y}_1$  and its acceleration  $\mathbf{a}_1$ , both given in the boat frame.

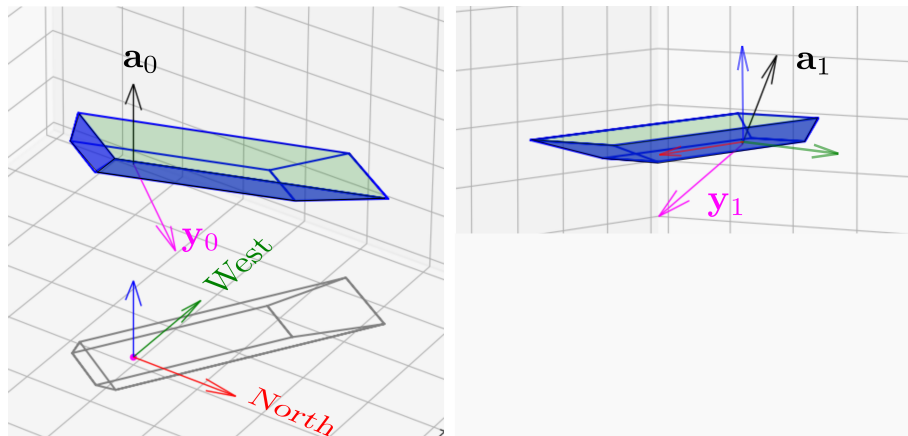


Figure 2.3: The boat has to find its heading from the accelerometer and the magnetometer

For simplicity, we assume that  $\mathbf{y}$  and  $\mathbf{a}$  have been normalized, *i.e.*, their norm are both equal to 1. In  $\mathcal{R}_0$ , we have

$$\mathbf{y}_0 = \begin{pmatrix} \cos I \\ 0 \\ -\sin I \end{pmatrix}, \quad \mathbf{a}_0 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

This means that when the orientation  $\mathbf{R}$  of the boat corresponds to identity, the magnetometer measures the field  $\mathbf{y}_0$  and the accelerometer returns  $\mathbf{a}_0$ .

- 1) From  $\mathbf{a}_1$ , give an estimation of the bank  $\theta$  (pitch angle) and the elevation  $\varphi$  (roll angle).
  - 2) Compute the rotation which leads our boat to the nearest horizontal orientation.
  - 3) Provide an estimation  $\hat{\psi}$  of the heading angle  $\psi$ .
  - 4) Using a simulation, check that the Euler angles are well estimated in case of small angles.
- 

#### EXERCISE 9.– *Immersion*

See the correction video at <https://youtu.be/E-XX52wbBQ8>

Consider the pendulum described by the state equations:

$$\begin{cases} \dot{\theta} \\ \dot{\omega} \end{cases} = \begin{pmatrix} \omega \\ -\sin \theta - \frac{1}{2}\omega \end{pmatrix}$$

1) Considering that the two states  $(\theta, \omega)$  and  $(\theta + 2k\pi, \omega), k \in \mathbb{Z}$ , are identical, show that the state space corresponds to a cylinder.

2) We want to represent the cylinder which is a two-dimensional manifold in the larger dimensional Cartesian space  $\mathbb{R}^3$ . We call this representation an *immersion*. Set  $x_1 = \cos \theta$ ,  $x_2 = \sin \theta$ ,  $x_3 = \omega$ . Give the state equations in the  $\mathbf{x}$ -space. Which constraint should satisfy  $\mathbf{x}$ ?

3) Provide a Runge-Kutta simulation on the  $\mathbf{x}$ -space of the system. The initial vector is chosen as  $(-1, 0, 6)$ . Draw the corresponding trajectory and interpret.

---

#### EXERCISE 10.– *Car on the torus*

See the correction video at <https://youtu.be/TcZ0fliVB-A>

Consider a car moving on the torus, parameterized by the two radius  $r_1, r_2$ , as illustrated by Figure 2.4. The pose of the car is represented by 3 angles  $x_1, x_2, x_3$ . The speed of the car is always equal to 1. The car has a single input  $u$  corresponding to its rotation rate with respect to the ground.

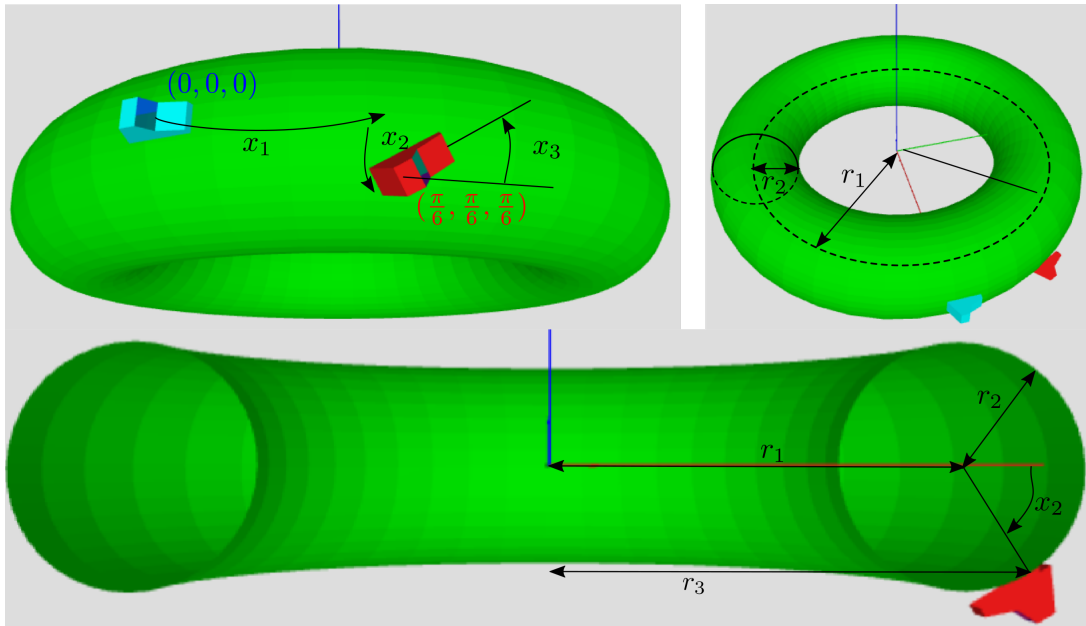


Figure 2.4: Representation of the pose of the car on the torus

- 1) What is the shape of the manifold corresponding to the state space. Give the state equations of the system.
- 2) Simulate the system for  $r_1 = 10, r_2 = 6$  and  $t \in [0, 100]$ . The initial state is taken as  $x_1 = x_2 = x_3 = 0$ . For the control, we choose  $u = 0.1$ .

#### EXERCISE 11.– *Manipulator robot*

See the correction video at <http://youtu.be/TPIHx0PE6P4>

A manipulator robot, such as *Staubli* represented on Figure 2.5, is composed of several rigid arms. We retrieve the coordinates of the end effector, at the extremity of the robot, using a series of geometric transformations. We can show that a parametrization with four degrees of freedom allows to represent these transformations. There are several possible parametrizations, each with its own advantages and disadvantages. The most widely used one is probably the *Denavit-Hartenberg* parametrization. In the case where the articulations are rotational joints (as is the case of the Staubli robot where the joins can turn), the parametrization represented by the figure might prove to be practical since it makes drawing the robot easier. This transformation is the composition of four elementary transformations: (i) a translation of length  $r$  following  $z$ ; (ii) a translation of length  $d$  following  $x$ ; (iii) a rotation of  $\alpha$  around  $y$  and (iv) a rotation of  $\theta$  (the variable activated around  $z$ ). Using the figure for drawing the arms and the photo for the robot, perform a realistic simulation of the robot's movement.

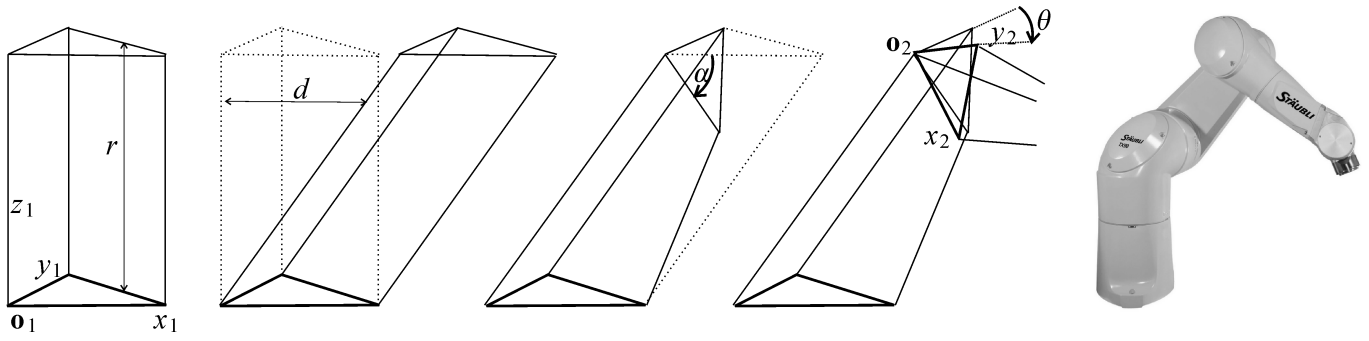


Figure 2.5: Parametrization for the direct geometric model of a manipulator robot

# Chapter 3

## Inertial unit

Assume that we are enclosed inside a box near to the Earth without any possibility to see the environment outside. Inside the box we may perform some inertial experiments such as observing bodies translating or rotating. These experiments allow us to measure indirectly the accelerations and the angular speed of the box in its own frame. The principle of an inertial unit is to estimate the pose (position and orientation) of the box from these inertial measurements only (called *proprioceptive measurements*), assuming that the initial pose is known.

### 3.1 Mechanization equations

For this purpose, we will describe the motion of our box by a kinematic model the inputs of which are the accelerations and the angular speeds these are directly measurable in the coordinate system of the box. The state vector is composed of

- the vector  $\mathbf{p} = (p_x, p_y, p_z)$  that gives the coordinates of the center of the box expressed in the absolute inertial coordinate system  $\mathcal{R}_0$ ,
- the orientation (which can either be given by the three Euler angles  $(\varphi, \theta, \psi)$  or a rotation matrix  $\mathbf{R}$ ) and
- the speed vector  $\mathbf{v}_r$  of the box expressed in its own coordinate system  $\mathcal{R}_1$ .

The inputs of the system are

- the acceleration  $\mathbf{a}_r = \mathbf{a}_{\mathcal{R}_1}$  of the center of the box also given in  $\mathcal{R}_1$  and
- the vector  $\boldsymbol{\omega}_r = \boldsymbol{\omega}_{\mathcal{R}_1/\mathcal{R}_0|\mathcal{R}_1} = (\omega_x, \omega_y, \omega_z)$  corresponding to the rotation vector of the box relative to  $\mathcal{R}_0$  expressed in  $\mathcal{R}_1$ .

We have indeed to express  $\mathbf{a}$ ,  $\boldsymbol{\omega}$  in the coordinate system of the box since these quantities are generally measured *via* the sensors attached on it. The first state equation is:

$$\dot{\mathbf{p}} \stackrel{(1,10)}{=} \mathbf{R}(\varphi, \theta, \psi) \cdot \mathbf{v}_r.$$

Let us differentiate this equation. We obtain:

$$\dot{\mathbf{p}} = \dot{\mathbf{R}} \cdot \mathbf{v}_r + \mathbf{R} \cdot \dot{\mathbf{v}}_r$$

with  $\mathbf{R} = \mathbf{R}(\varphi, \theta, \psi)$ . From this equation, we isolate  $\dot{\mathbf{v}}_r$  to get

$$\dot{\mathbf{v}}_r = \underbrace{\mathbf{R}^T \cdot \dot{\mathbf{p}}}_{\mathbf{a}_r} - \mathbf{R}^T \dot{\mathbf{R}} \cdot \mathbf{v}_r \stackrel{(1.4)}{=} \mathbf{a}_r - \boldsymbol{\omega}_r \wedge \mathbf{v}_r.$$

which constitutes the second state equation. Figure 3.1 shows a situation where a robot has a constant speed and follows a circle. The speed vector  $\mathbf{v}_r$  is a constant whereas  $\mathbf{a}_r$  is different from zero.

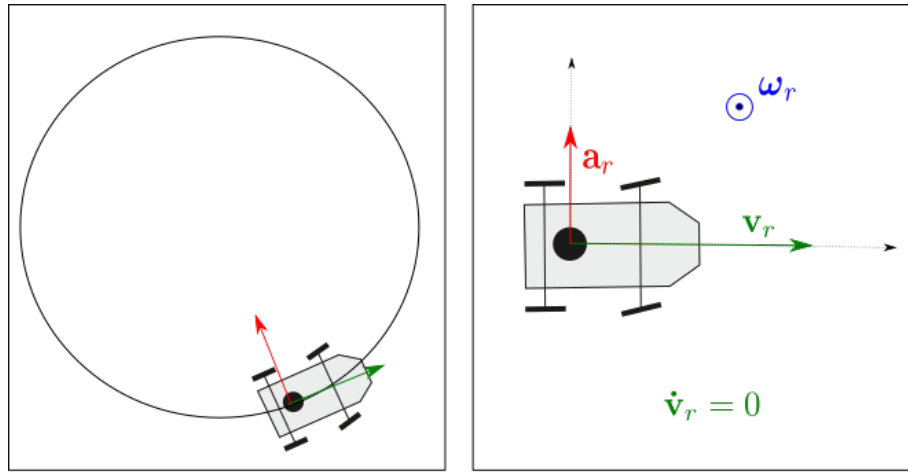


Figure 3.1: Illustration of the formula  $\dot{\mathbf{v}}_r = \mathbf{a}_r - \boldsymbol{\omega}_r \wedge \mathbf{v}_r$  when  $\dot{\mathbf{v}}_r = \mathbf{0}$ . Left: the robot follows a circle; right: representation of the vectors in the robot frame.

We obtain the *navigation mechanization equations* [12]:

$$\begin{cases} \dot{\mathbf{p}} &= \mathbf{R}(\varphi, \theta, \psi) \cdot \mathbf{v}_r \\ \begin{pmatrix} \dot{\varphi} \\ \dot{\theta} \\ \dot{\psi} \end{pmatrix} &\stackrel{(2.4)}{=} \begin{pmatrix} 1 & \tan \theta \sin \varphi & \tan \theta \cos \varphi \\ 0 & \cos \varphi & -\sin \varphi \\ 0 & \frac{\sin \varphi}{\cos \theta} & \frac{\cos \varphi}{\cos \theta} \end{pmatrix} \cdot \boldsymbol{\omega}_r \\ \dot{\mathbf{v}}_r &= \mathbf{a}_r - \boldsymbol{\omega}_r \wedge \mathbf{v}_r \end{cases} \quad (3.1)$$

**Using rotation matrix instead of Euler angles.** From Equation (1.4), we have  $\mathbf{R}^T \dot{\mathbf{R}} = \boldsymbol{\Lambda}(\mathbf{R}^T \boldsymbol{\omega})$ . Thus, the navigation mechanization equations can be written without any Euler angles as

$$\begin{cases} \dot{\mathbf{p}} &= \mathbf{R} \cdot \mathbf{v}_r \\ \dot{\mathbf{R}} &= \mathbf{R} \cdot (\boldsymbol{\omega}_r \wedge) \\ \dot{\mathbf{v}}_r &= \mathbf{a}_r - \boldsymbol{\omega}_r \wedge \mathbf{v}_r \end{cases} \quad (3.2)$$

The main advantage of this representation is that we do not have the singularity that exists in (2.4) for  $\cos \theta = 0$ . But instead, we have redundancies, since  $\mathbf{R}$  of dimension 9 replaces the 3 Euler angles.

### 3.2 With gravity

In case where there exists a gravity  $\mathbf{g}(\mathbf{p})$  which depends on  $\mathbf{p}$ , the actual acceleration  $\mathbf{a}_r$  corresponds to the gravity plus the measured acceleration  $\mathbf{a}_r^{mes}$ . In this case, the inertial unit can be described by the following state equations

$$\begin{cases} \dot{\mathbf{p}} &= \mathbf{R} \cdot \mathbf{v}_r \\ \dot{\mathbf{R}} &= \mathbf{R} \cdot (\boldsymbol{\omega}_r \wedge) \\ \dot{\mathbf{v}}_r &= \mathbf{R}^T \cdot \mathbf{g}(\mathbf{p}) + \mathbf{a}_r^{mes} - \boldsymbol{\omega}_r \wedge \mathbf{v}_r \end{cases} \quad (3.3)$$

In this case, the input of the system are  $\mathbf{a}_r^{mes}$  and  $\boldsymbol{\omega}_r$  which are obtained by the accelerometers and the gyroscope [13]. To integrate this state equation for localization purpose, we need to know the initial state and a *gravity map*  $\mathbf{g}(\mathbf{p})$ .

### 3.3 Integration scheme

A pure inertial unit (without hybridization and without taking into account Earth's gravity) represents the robot by the kinematic model of Figure 3.2, which itself uses mechanization equations. The input vector  $\mathbf{u} = (\mathbf{a}_r, \boldsymbol{\omega}_r)$  corresponds to the measured inertial inputs (accelerations and rotation speeds viewed by an observer on the ground, but expressed in the frame of the robot).

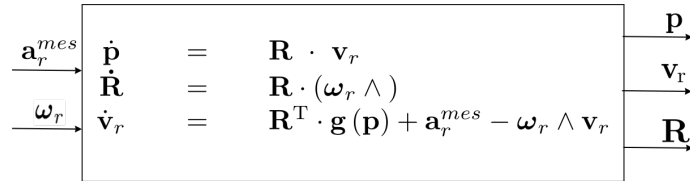


Figure 3.2: Mechanization equations

We may use a numerical integration method such the Euler method:

$$\begin{aligned} \mathbf{x}(t + dt) &:= \mathbf{x}(t) + dt \cdot \mathbf{f}(\mathbf{x}(t), \mathbf{R}(t), \mathbf{u}(t)) \\ \mathbf{R}(t + dt) &:= \mathbf{R}(t) \cdot e^{dt \cdot \boldsymbol{\omega}_r(t)} \end{aligned}$$

with

$$\begin{aligned} \mathbf{x} &= (\mathbf{p}, \mathbf{v}_r) \\ \mathbf{f}(\mathbf{x}, \mathbf{R}, \mathbf{u}) &= \begin{pmatrix} \dot{\mathbf{p}} \\ \dot{\mathbf{R}} \\ \dot{\mathbf{v}}_r \end{pmatrix} = \begin{pmatrix} \mathbf{R} \cdot \mathbf{v}_r \\ \mathbf{R} \cdot (\boldsymbol{\omega}_r \wedge) \\ \mathbf{R}^T \cdot \mathbf{g}(\mathbf{p}) + \mathbf{a}_r^{mes} - \boldsymbol{\omega}_r \wedge \mathbf{v}_r \end{pmatrix} \end{aligned}$$

If we use a Runge-Kutta midpoint integration scheme instead, we get

$$\begin{aligned} \mathbf{x}(t + dt) &:= \mathbf{x}(t) + dt \cdot \left( \mathbf{f} \left( \mathbf{x}(t) + \frac{dt}{2} \mathbf{f}(\mathbf{x}(t), \mathbf{R}(t), \mathbf{u}(t)), \mathbf{R}(t) \cdot e^{\frac{dt}{2} \cdot \boldsymbol{\omega}_r(t)}, \mathbf{u}(t) \right) \right) \\ \mathbf{R}(t + dt) &:= \mathbf{R}(t) \cdot e^{dt \cdot \boldsymbol{\omega}_r(t)} \end{aligned}$$

as illustrated by Figure 3.3.

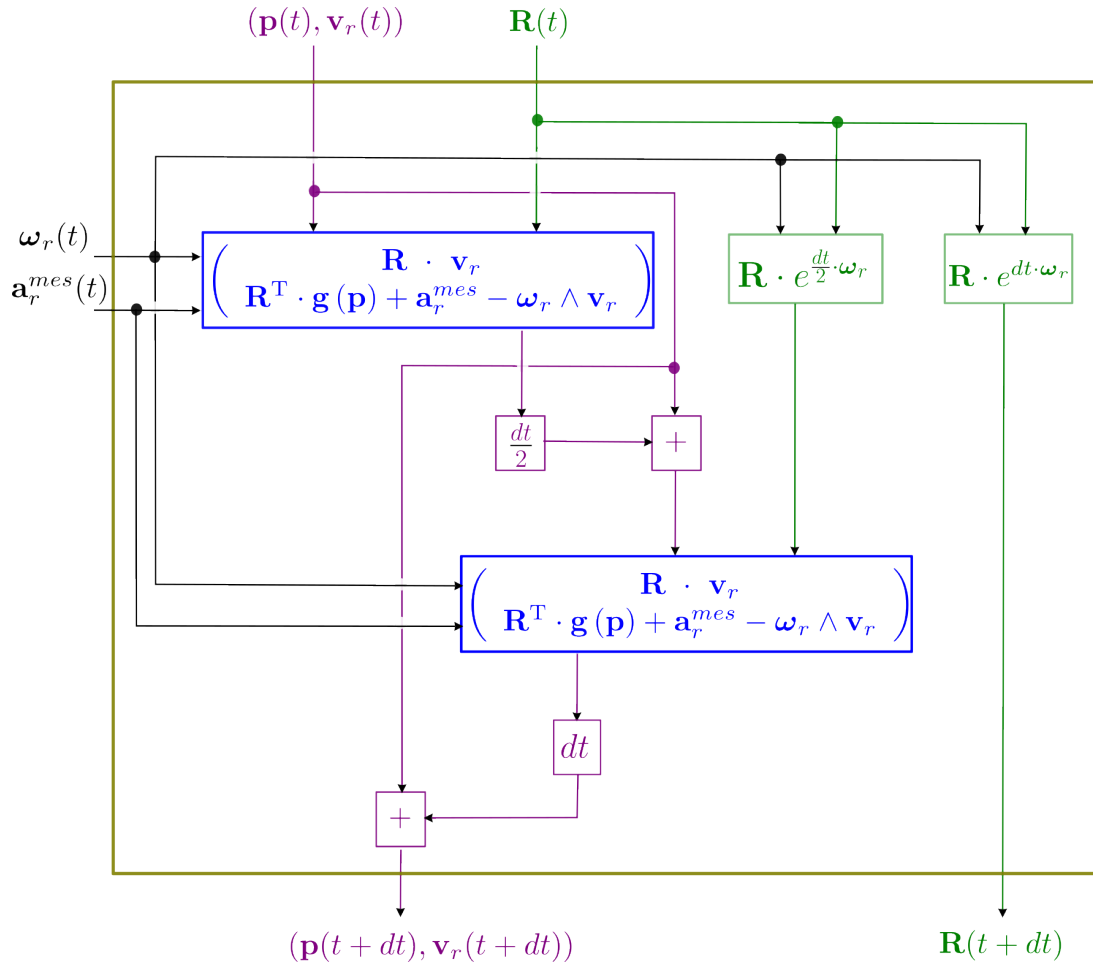


Figure 3.3: Runge Kutta midpoint method for integrating the mechanization equations

### 3.4 Dead reckoning

For dead reckoning (*i.e.*, localization without external sensors) there are generally extremely precise laser gyrometers (around  $0.001 \text{ deg/s}$ ). These make use of the Sagnac effect (in a circular optical fiber turning around itself, the time taken by light to travel an entire round-trip depends on the path direction). Using three fibers, these gyrometers generate the vector  $\boldsymbol{\omega}_r = (\omega_x, \omega_y, \omega_z)$ . There are also accelerometers capable of measuring the acceleration  $\mathbf{a}_r$  with a very high degree of precision. In pure inertial mode, we determine our position by integration of Equations (3.3) only using the acceleration  $\mathbf{a}_r$  and the rotation speed  $\boldsymbol{\omega}_r$ , both expressed in the coordinate system of the box. In the case where we are measuring the quantity  $\mathbf{v}_r$  (also expressed in the frame of the inertial unit attached to the robot) with a *Doppler velocity log*, we only need to integrate the first and the last of these three equations. Finally, when the robot is a correctly ballasted submarine or a terrestrial robot moving on a relatively plane ground, we know *a priori* that on average the bank and

the elevation are equal to zero. We may therefore incorporate this information through a Kalman filter in order to limit the drift in positioning. An efficient inertial unit integrates an amalgamation of all the available information.



# Exercises

---

## EXERCISE 12.– *Foucault pendulum*

See the correction video at <https://youtu.be/eRb-fi0DqX8>

The Foucault pendulum is a simple device, introduced in 1851 by the French physicist Léon Foucault, to provide an evidence of the Earth's rotation. It consists of a long and heavy pendulum suspended from the high roof above a circular area. Over an extended time period, we observe that the plane of oscillation rotates. The goal of the exercise is to compute the associated oscillation period.

1) A car drives with a speed  $v$  on a flat circular road of radius  $r$ , as illustrated by Figure 3.4. Attached to this car, a pendulum oscillates without friction in a plane, called the *oscillation plane*. The ratio  $\frac{r}{v}$  is assumed to be large enough to have an oscillation plane continuously vertical. Give an expression of the period  $T_p$  to which the oscillation plane rotates.

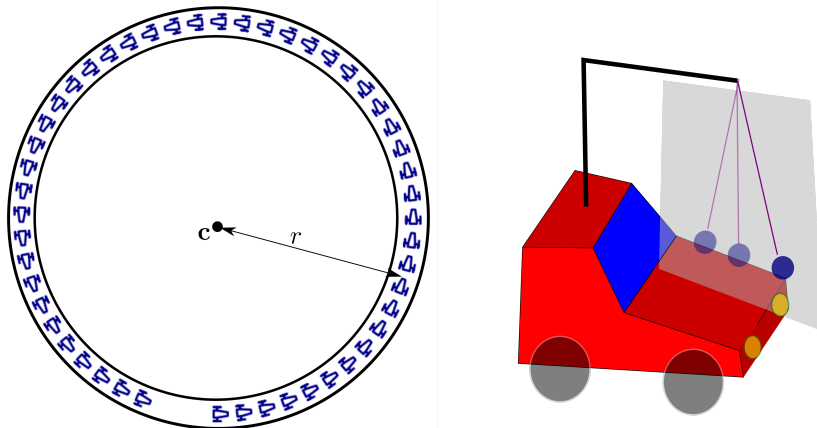


Figure 3.4: A car driving on a circular road with center  $\mathbf{c}$ . The oscillation plane (gray) remains vertical

2) Consider a non-rotating Earth of radius  $r_0$  with our car at the surface going at speed  $v$ . The pendulum is still oscillating in a vertical oscillation plane in the car's frame. If the car goes straight, it will follow a geodesic, as represented by the green trajectory of Figure 3.5. For the sphere, these geodesics are circles with center  $\mathbf{o}$  and the oscillation plane remains static in the car frame. Consider

now the situation where the car drives along a parallel toward East. If the car is in the North hemisphere, it has to turn left to stay on the parallel, as illustrated by the red trajectory. Similarly, if the car is in the South hemisphere, it has to turn right to follow the parallel. Compute the rotation rate of pendulum oscillation plane with respect to  $v$  and its the latitude  $\theta$  of the parallel.

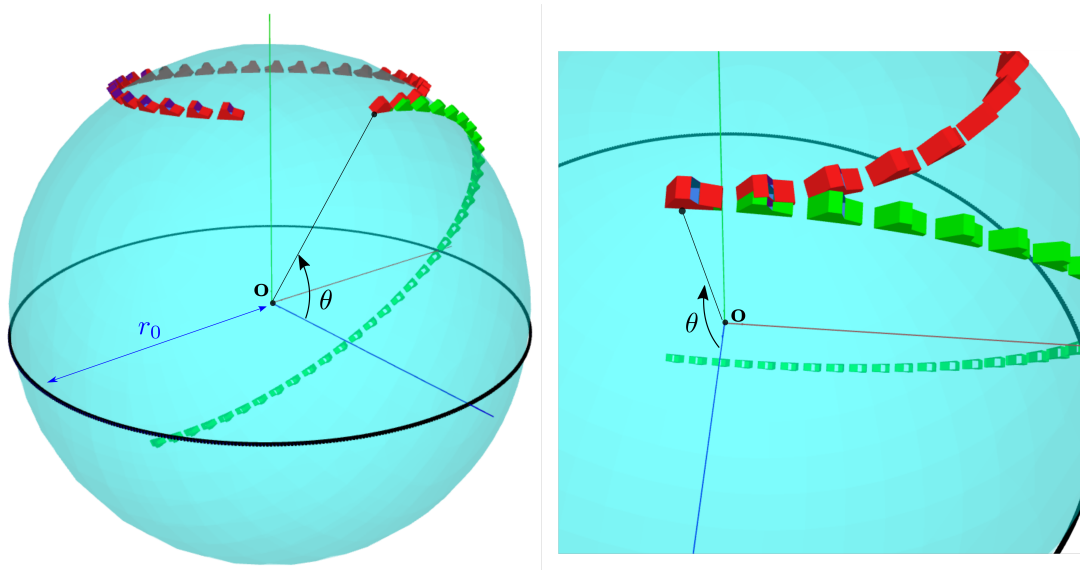


Figure 3.5: The car turns around a static Earth

3) Consider the Earth of radius  $r_0$  rotating at a period of  $T_E = 24h$ . The car is now static with respect to the Earth and always encloses the pendulum inside. Compute the period  $T_p$  of the rotation of the oscillation plane.

---

EXERCISE 13.– *Schuler oscillations in an inertial unit*

See the correction video at <https://youtu.be/9vsZCnPXCPo>

Consider the Earth which is considered as static, without rotation, with a gravity field directed toward the center and with a magnitude  $g = 9.81ms^{-2}$ . An inertial unit is fixed in a robot  $\mathcal{R}_1$  at the surface of the Earth of radius  $r$ . Its position is with  $\mathbf{p} = (r, 0, 0)$  and its orientation  $\mathbf{R}$  corresponds to the identity matrix, as in Figure 3.6. The robot  $\mathcal{R}_1$  will always be static in the exercise.

1) Give the values for  $\mathbf{a}_r^{mes}$  and  $\boldsymbol{\omega}_r$  collected by the inertial unit of  $\mathcal{R}_1$ .

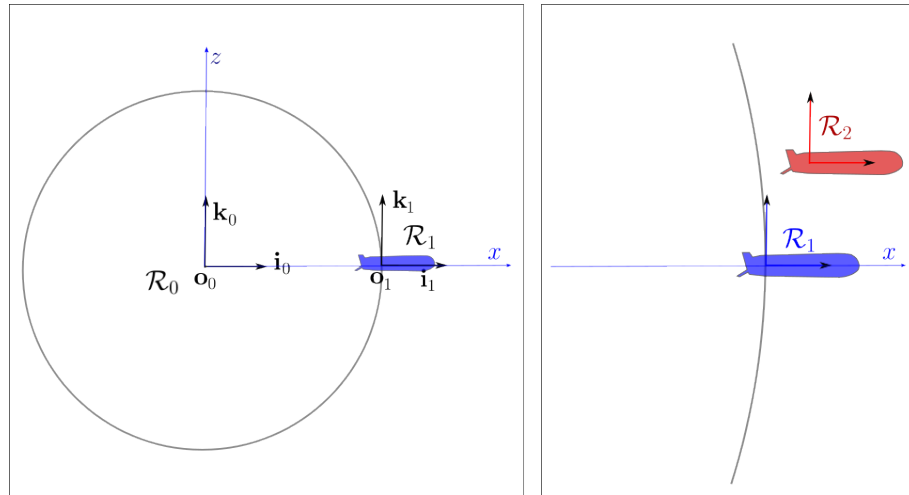


Figure 3.6: Blue the actual robot which is static. Red the fake position though by the inertial unit

2) We now want to lure the inertial unit by initializing the robot at some other place, say  $\mathcal{R}_2$  at the neighborhood of  $\mathcal{R}_1$ , and finding a trajectory so that the unit believes that it is fixed in  $\mathcal{R}_1$ . More precisely, we want to find a motion  $\mathbf{p}(t) = (x(t), y(t), z(t))$  for  $\mathcal{R}_2$  so that the inertial unit senses exactly the inputs  $\mathbf{a}_r^{mes}, \boldsymbol{\omega}_r$  of  $\mathcal{R}_1$ . For this, we assume that at time  $t = 0$ , the orientation  $\mathbf{R}$  for  $\mathcal{R}_2$  is the same as for  $\mathcal{R}_1$ , *i.e.*, it is the identity matrix. For simplicity, we also assume that  $\mathcal{R}_2$  remains in the plane  $y = 0$ . Simplify the state equations for  $\mathcal{R}_2$  in our specific case.

3) Provide a simulation of the trajectory of  $\mathcal{R}_2$ , initialized at  $\mathbf{p}(0) = (r, 0, 1)^T$ . Interpret.

4) Linearize the state equation for  $\mathcal{R}_2$  at the state vector corresponding to  $\mathcal{R}_1$ . Compute the eigen values and give an interpretation.

5) Assuming that the inertial unit is not perfectly initialized, what can you conclude for the corresponding error propagation.

EXERCISE 14.- *3D robot graphics*

See the correction video at <https://youtu.be/t4ay2KENg-k>

Drawing two- or three-dimensional robots or objects on the screen is widely used for simulation in robotics. The classic method (used by OPENGL) relies on modeling the posture of objects using a series of affine transformations (rotations, translations, homotheties) of the form:

$$\mathbf{f}_i : \begin{matrix} \mathbb{R}^n & \rightarrow & \mathbb{R}^n \\ \mathbf{x} & \mapsto & \mathbf{A}_i \mathbf{x} + \mathbf{b}_i \end{matrix}$$

with  $n = 2$  or  $3$ . However, the manipulation of compositions of affine functions is less simple than that of linear applications. The idea of the transformation in *homogeneous coordinates* is to transform a system of affine equations into a system of linear equations. Notice first of all that an affine equation of the type  $\mathbf{y} = \mathbf{A}\mathbf{x} + \mathbf{b}$  can be written as:

$$\begin{pmatrix} \mathbf{y} \\ 1 \end{pmatrix} = \begin{pmatrix} \mathbf{A} & \mathbf{b} \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \mathbf{x} \\ 1 \end{pmatrix}.$$

We will thus define the *homogeneous transformation* of a vector as follows:

$$\mathbf{x} \mapsto \mathbf{x}_h = \begin{pmatrix} \mathbf{x} \\ 1 \end{pmatrix}.$$

Thus, an equation of the type

$$\mathbf{y} = \mathbf{A}_3 (\mathbf{A}_2 (\mathbf{A}_1 \mathbf{x} + \mathbf{b}_1) + \mathbf{b}_2) + \mathbf{b}_3$$

involving the composition of three affine transformations, can be written as:

$$\mathbf{y}_h = \begin{pmatrix} \mathbf{A}_3 & \mathbf{b}_3 \\ \mathbf{0} & 1 \end{pmatrix} \begin{pmatrix} \mathbf{A}_2 & \mathbf{b}_2 \\ \mathbf{0} & 1 \end{pmatrix} \begin{pmatrix} \mathbf{A}_1 & \mathbf{b}_1 \\ \mathbf{0} & 1 \end{pmatrix} \mathbf{x}_h.$$

A *sketch* is a matrix with two or three rows (depending on the object being in the plane or in space) and  $n$  columns representing the  $n$  vertices of a rigid polygon embodying the object. It is important that the union of all the segments formed by two consecutive points of the sketch forms all the vertices of the polygon that we wish to represent.

1) Let us consider the underwater robot (or AUV for *Autonomous Underwater Vehicle*) whose sketch matrix in homogeneous coordinates is :

$$\mathbf{M} = \begin{pmatrix} 0 & 0 & 10 & 0 & 0 & 10 & 0 & 0 \\ -1 & 1 & 0 & -1 & -0.2 & 0 & 0.2 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{pmatrix}.$$

Draw this sketch in perspective view on a piece of paper.

2) The state equations of the robot are the following:

$$\begin{cases} \dot{\mathbf{p}} & = \begin{pmatrix} v \cos \theta \cos \psi \\ v \cos \theta \sin \psi \\ -v \sin \theta \end{pmatrix} \\ \dot{v} & = u_1 \\ \begin{pmatrix} \dot{\varphi} \\ \dot{\theta} \\ \dot{\psi} \end{pmatrix} & = \begin{pmatrix} 1 & \tan \theta \sin \varphi & \tan \theta \cos \varphi \\ 0 & \cos \varphi & -\sin \varphi \\ 0 & \frac{\sin \varphi}{\cos \theta} & \frac{\cos \varphi}{\cos \theta} \end{pmatrix} \cdot \begin{pmatrix} -0.1 \sin \varphi \cdot \cos \theta \\ v \cdot u_2 \\ v \cdot u_3 \end{pmatrix} \end{cases}$$

where  $(\varphi, \theta, \psi)$  are the three Euler angles. The inputs of the system are the tangential acceleration  $u_1$ , the pitch rudder  $u_2$  and the yaw rudder  $u_3$ . The state corresponds to  $(\mathbf{p}, v, \varphi, \theta, \psi)$ . Program a function able to draw the robot in 3D together with its shadow, in the plane  $x-y$ . Verify that the drawing is correct by moving the six degrees of freedom of the robot one by one. Obtain a three-dimensional representation such as the one illustrated in Figure 3.7.

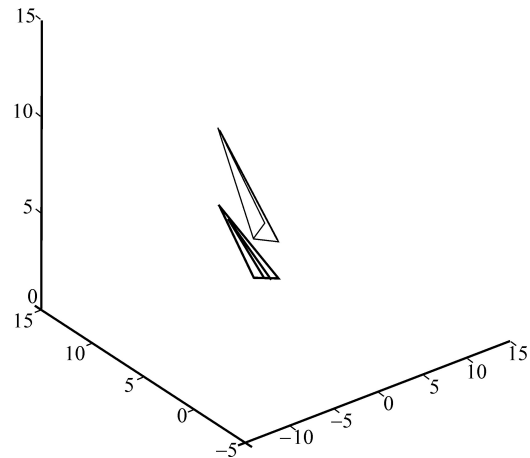


Figure 3.7: 3D representation of the robot together with its shadow in the horizontal plane

3) Simulate the robot in various conditions. In the simulation, draw the instantaneous rotation vector of the robot.

4) Provide a model and a simulation which uses a rotation matrix for the orientation instead of the Euler angles. Compare with the simulation made previously.



# Chapter 4

## Dynamic modeling

### 4.1 Principle

A robot (airplane, submarine, boat) can often be considered as a solid body whose inputs are the (tangential and angular) accelerations. These quantities are analytic functions of the forces that are at the origin of the robot's movement. For the dynamic modeling of a submarine the reference work is the book of Fossen [14], but the related notions can also be used for other types of solid robots such as planes, boats or quadrotors. In order to obtain a dynamic model, it is sufficient to take the kinematic equations and to consider that the angular and tangential accelerations are caused by forces and torques which become the new inputs of our system. The link between the accelerations and the forces is done by Newton's second law (or the *fundamental principle of dynamics*). Thus, for instance if  $\mathbf{f}$  is the external force expressed in the inertial frame and  $m$  is the mass of the robot, we have

$$m\ddot{\mathbf{p}} = \mathbf{f}.$$

Since the speed and accelerations are generally measured by sensors embedded in the system, we generally prefer to express the speed and accelerations in the frame of the robot:

$$m\mathbf{a}_r = \mathbf{f}_r,$$

where  $\mathbf{f}_r$  is the applied forces vector expressed in the robot frame. The same type of relation, known as the *Euler's rotation equation*, exists for rotations. It is given by

$$\mathbf{I}\dot{\boldsymbol{\omega}}_r + \boldsymbol{\omega}_r \wedge (\mathbf{I}\boldsymbol{\omega}_r) = \boldsymbol{\tau}_r \tag{4.1}$$

where  $\boldsymbol{\tau}_r$  is the applied torque and  $\boldsymbol{\omega}_r$  is the rotation vector both expressed in the robot frame. The inertia matrix  $\mathbf{I}$  is attached to the robot (*i.e.*, computed in the robot frame) and we generally choose the robot frame to make  $\mathbf{I}$  diagonal. Recall that the inertia matrix of a solid body occupying the volume  $V$  is

$$\mathbf{I} = \int_V \rho(x, y, z) \begin{pmatrix} y^2 + z^2 & -xy & -xz \\ -xy & x^2 + z^2 & -yz \\ -xz & -yz & x^2 + y^2 \end{pmatrix} dx dy dz$$

where  $\rho(x, y, z)$  is the density.

Relation 4.1 is a consequence of the *Euler's second law* which states that in an inertial frame, the time derivative of the angular momentum  $\mathcal{L} = \mathbf{R} \cdot \mathbf{I} \cdot \mathbf{R}^T \cdot \boldsymbol{\omega}$  equals the applied torque  $\boldsymbol{\tau}$ . In the inertial frame, this can be expressed as

$$\begin{aligned} & \frac{d}{dt} \mathcal{L} = \boldsymbol{\tau} \\ \Leftrightarrow & \frac{d}{dt} (\mathbf{R} \cdot \mathbf{I} \cdot \boldsymbol{\omega}_r) = \mathbf{R} \cdot \boldsymbol{\tau}_r \\ \Leftrightarrow & \dot{\mathbf{R}} \mathbf{I} \cdot \boldsymbol{\omega}_r + \mathbf{R} \cdot \mathbf{I} \cdot \dot{\boldsymbol{\omega}}_r = \mathbf{R} \cdot \boldsymbol{\tau}_r \\ \Leftrightarrow & \mathbf{R}^T \dot{\mathbf{R}} \mathbf{I} \cdot \boldsymbol{\omega}_r + \mathbf{I} \cdot \dot{\boldsymbol{\omega}}_r = \boldsymbol{\tau}_r \\ \Leftrightarrow & \boldsymbol{\omega}_r \wedge (\mathbf{I} \boldsymbol{\omega}_r) + \mathbf{I} \dot{\boldsymbol{\omega}}_r = \boldsymbol{\tau}_r. \end{aligned}$$

## 4.2 Equation of a 3D robot

The state equations for a 3-dimensional rigid robot are given by

$$\left\{ \begin{array}{ll} \dot{\mathbf{p}} & = \mathbf{R} \cdot \mathbf{v}_r & (i) \\ \dot{\mathbf{R}} & = \mathbf{R} \cdot (\boldsymbol{\omega}_r \wedge) & (ii) \\ \dot{\mathbf{v}}_r & = \mathbf{R}^T \cdot \mathbf{g} + \frac{1}{m} \cdot \mathbf{f}_r - \boldsymbol{\omega}_r \wedge \mathbf{v}_r & (iii) \\ \dot{\boldsymbol{\omega}}_r & = \mathbf{I}^{-1} \cdot (\boldsymbol{\tau}_r - \boldsymbol{\omega}_r \wedge \mathbf{I} \cdot \boldsymbol{\omega}_r) & (iv) \end{array} \right. \quad (4.2)$$

where  $\mathbf{p}$  is the position,  $\mathbf{R}$  is the orientation matrix of the robot,  $\mathbf{v}_r$  the speed in the robot frame,  $\boldsymbol{\omega}_r$  the rotation vector in the robot frame and  $\mathbf{g} = (0, 0, g)$  is the gravity. The three first equations (i), (ii), (iii) correspond to the kinematic equations and have already been derived (see Equation 3.3). Note that in Equation (iii), the acceleration  $\mathbf{a}_r$  (expressed in the robot frame) comes from the Newton's second law where  $\mathbf{f}_r$  is the *resultant*. Equation (iv) comes from the Euler's rotation equation (4.1) where  $\boldsymbol{\tau}_r$  is the torque vector.

The motion of the robot is thus mainly governed by forces which create the resultant  $\mathbf{f}_r$  and the torque, the  $\boldsymbol{\tau}_r$ . Assume for simplicity that we have only one force with origin  $\mathbf{q}$  which produces a force  $f$  in the direction given by the unit vector  $\mathbf{d}$ . Both  $\mathbf{q}$  and  $\mathbf{d}$  are given in the body frame. The force contributes to the resultant on the robot as  $\mathbf{f}_r = \mathbf{d} \cdot f$  and to the torque as

$$\boldsymbol{\tau}_r = \mathbf{q} \wedge \mathbf{d} \cdot f. \quad (4.3)$$

The forces can be actuated or not. For example:

1. For an underwater robot, the Archimedes force is an upward buoyant force that is exerted on a body which is equal to the weight of the water that the body displaces.
2. The propeller of torpedo generates a controlled force oriented forward.

The torque  $\boldsymbol{\tau}_r$  can be actuated or not. Let us give few examples.

1. *Static restoring torque.* Consider an underwater vehicle and denote by  $\mathbf{q}$  the vector between the center of gravity of and the center of buoyancy then the water yields a torque given (in the robot frame) by (see 4.3)

$$\boldsymbol{\tau}_r = \mathbf{q} \wedge \mathbf{R}^T \cdot \begin{pmatrix} 0 \\ 0 \\ -g \end{pmatrix} \cdot \rho V$$

where  $V$  is the immersed volume and  $\rho$  is the density of the water. It is called the *restoring torque*.

2. *Dynamic restoring torque.* A robot, such as an aircraft or a torpedo, may rotate about its center of mass. The resultant of the aerodynamic forces acts on a point, the center of pressure, which is not the center of gravity. In this case a small yaw or pitch motions, aerodynamic forces occur. The stability exists when the corresponding torque tends to restore the system to its initial orientation. The stability is obtained when the center of gravity is in front of the center of pressure.
3. *Damping torque.* When robot rotates, it will create a friction that slows down the robot. A linear approximation is

$$\boldsymbol{\tau}_r = -\mathbf{D} \cdot \boldsymbol{\omega}_r$$

where  $\mathbf{D}$  is the damping matrix.

### 4.3 Modeling a quadrotor

As an illustration, we consider a quadrotor (see Figure 4.2) for which we want a dynamic model. The robot has four propellers that can be tuned independently. This will allow us to control the attitude and position of the robot by changing the speeds of the motors. The drone matrice 600 represented on Figure 4.1 has 6 propellers but can also be considered has a quadrotor since all its propellers have a vertical orientation.



Figure 4.1: Matrice 600 of ENSTA-Bretagne

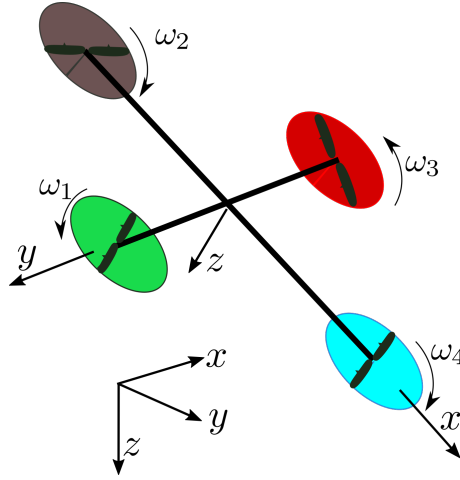


Figure 4.2: Quadrotor

We distinguish the front/rear propellers (blue and black), rotating clockwise and right/left propellers (red and green), rotating counterclockwise. The value of the force generated by the  $i$ th propeller is proportional to the squared speeds of the rotors, *i.e.*, is equal to  $\beta \cdot \omega_i \cdot |\omega_i|$ , where  $\beta$  is the thrust factor. Denote by  $\delta$  the drag factor and  $\ell$  the distance between any rotor and the center of the robot. The forces and torques generated on the robot are

$$\begin{pmatrix} \tau_0 \\ \tau_1 \\ \tau_2 \\ \tau_3 \end{pmatrix} = \begin{pmatrix} \beta & \beta & \beta & \beta \\ -\beta\ell & 0 & \beta\ell & 0 \\ 0 & -\beta\ell & 0 & \beta\ell \\ -\delta & \delta & -\delta & \delta \end{pmatrix} \cdot \begin{pmatrix} \omega_1 \cdot |\omega_1| \\ \omega_2 \cdot |\omega_2| \\ \omega_3 \cdot |\omega_3| \\ \omega_4 \cdot |\omega_4| \end{pmatrix}$$

where  $\tau_0$  is the total thrust generated by the rotors and  $\tau_1, \tau_2, \tau_3$  are the torques generated by differences in the rotor speeds.

Since the resultant and the torque are given by

$$\mathbf{f}_r = \begin{pmatrix} 0 \\ 0 \\ -\tau_0 \end{pmatrix} \text{ and } \boldsymbol{\tau}_r = \begin{pmatrix} \tau_1 \\ \tau_2 \\ \tau_3 \end{pmatrix},$$

from (4.2), we get that the state equations for the quadrotor are given by

$$\begin{cases} \dot{\mathbf{p}} &= \mathbf{R} \cdot \mathbf{v}_r \\ \dot{\mathbf{R}} &= \mathbf{R} \cdot (\boldsymbol{\omega}_r \wedge) \\ \dot{\mathbf{v}}_r &= \mathbf{R}^T \cdot \mathbf{g} + \begin{pmatrix} 0 \\ 0 \\ -\frac{\tau_0}{m} \end{pmatrix} - \boldsymbol{\omega}_r \wedge \mathbf{v}_r \\ \dot{\boldsymbol{\omega}}_r &= \mathbf{I}^{-1} \cdot (\boldsymbol{\tau}_r - \boldsymbol{\omega}_r \wedge (\mathbf{I} \cdot \boldsymbol{\omega}_r)) \end{cases}$$

where  $\mathbf{p}$  is the position and  $\mathbf{R}$  is the orientation matrix of the robot.

## 4.4 Euler-Lagrange method

To find the state equations of a more general complex mechanical system with articulations or constraints with the environment, a Lagrangian method is often used. When dealing with Lagrangian mechanics, we generally spot the mechanical system at each time by the vector of the  $n$  degrees of freedom  $\mathbf{q} = (q_1, \dots, q_n)^T$ . The  $q_i$ 's are called the generalized coordinates. The vector of the generalized speeds is  $\dot{\mathbf{q}} = (\dot{q}_1, \dots, \dot{q}_n)^T$ . We define the *Lagrangian* as

$$L(\mathbf{q}, \dot{\mathbf{q}}) = T(\mathbf{q}, \dot{\mathbf{q}}) - V(\mathbf{q}),$$

where  $T(\mathbf{q}, \dot{\mathbf{q}})$  is the kinetic energy and  $V(\mathbf{q})$  is the potential energy. The quantity  $T(\mathbf{q}, \dot{\mathbf{q}})$  can generally be written as :

$$T(\mathbf{q}, \dot{\mathbf{q}}) = \frac{1}{2} \dot{\mathbf{q}}^T \mathbf{M}(\mathbf{q}) \dot{\mathbf{q}},$$

where  $\mathbf{M}(\mathbf{q})$  is called the *inertia matrix*. It is a  $n \times n$  symmetric and positive definite. The Euler-Lagrange equations are given by

$$\frac{d}{dt} \frac{\partial L}{\partial \dot{q}_i} - \frac{\partial L}{\partial q_i} = u_i, \quad i = 1, \dots, n. \quad (4.4)$$

where  $u_i$  are called the generalized forces associated to  $q_i$ . It corresponds to a force (if  $q_i$  is a length) or to a torque (if  $q_i$  is an angle). The generalized forces are external to the system and do not derive from a potential. The Euler-Lagrange equations form a system of  $n$  differential equations of second order:

$$\mathbf{M}(\mathbf{q}) \ddot{\mathbf{q}} + \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) \dot{\mathbf{q}} + \mathbf{g}(\mathbf{q}) = \mathbf{u},$$

where  $\mathbf{g}(\mathbf{q})$  is the gravity,  $\mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) \dot{\mathbf{q}}$  is the vector of Coriolis and centrifugal forces. We thus get the state equations:

$$\frac{d}{dt} \begin{pmatrix} \mathbf{q} \\ \dot{\mathbf{q}} \end{pmatrix} = \begin{pmatrix} \dot{\mathbf{q}} \\ \mathbf{M}^{-1}(\mathbf{q}) \cdot (\mathbf{u} - \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) \dot{\mathbf{q}} - \mathbf{g}(\mathbf{q})) \end{pmatrix}.$$

## 4.5 Proof of the Euler-Lagrange-equations

We will now give a simple proof for the Euler-Lagrange equations (4.4) which is based on the second Newton law. What is surprising is that only the kinetic and potential energies of the system are involved in these equations. Expressions for internal forces have completely disappeared.

For simplicity's sake, we will assume that the mechanical system under consideration is made up of a set of elementary particles (the system's atoms, for example) of mass  $m_k$ . Define

- $\mathbf{r}_k(\mathbf{q})$  : the position vector for each particle. It is a function of the generalized coordinates  $\mathbf{q}$  of the system.

- $\mathbf{v}_k(\mathbf{q}, \dot{\mathbf{q}})$  : is the velocity vector for each particle.

The kinetic energy of the system is

$$T(\mathbf{q}, \dot{\mathbf{q}}) = \frac{1}{2} \sum_k m_k \mathbf{v}_k^T \mathbf{v}_k.$$

Assume move  $q_i$  maintaining all other generalized coordinates constant. Let us compute  $\frac{d}{dt} \frac{\partial T}{\partial \dot{q}_i} - \frac{\partial T}{\partial q_i}$ . First,

$$\begin{aligned} \frac{\partial}{\partial q_i} T(\mathbf{q}, \dot{\mathbf{q}}) &= \frac{1}{2} \sum_k m_k \frac{\partial}{\partial q_i} (\mathbf{v}_k^T \mathbf{v}_k) \\ &= \frac{1}{2} \sum_k m_k \left( \mathbf{v}_k^T \frac{\partial \mathbf{v}_k}{\partial q_i} + \frac{\partial \mathbf{v}_k^T}{\partial q_i} \mathbf{v}_k \right) \\ &= \sum_k m_k \mathbf{v}_k^T \frac{\partial \mathbf{v}_k}{\partial q_i}. \end{aligned} \quad (4.5)$$

and

$$\begin{aligned} \frac{d}{dt} \frac{\partial}{\partial \dot{q}_i} T(\mathbf{q}, \dot{\mathbf{q}}) &= \frac{d}{dt} \frac{\partial}{\partial \dot{q}_i} \left( \frac{1}{2} \sum_k m_k \mathbf{v}_k^T \mathbf{v}_k \right) \\ &= \sum_k \frac{1}{2} m_k \frac{d}{dt} \left( \frac{\partial}{\partial \dot{q}_i} (\mathbf{v}_k^T \mathbf{v}_k) \right) \\ &= \sum_k \frac{1}{2} m_k \frac{d}{dt} \left( \frac{\partial \mathbf{v}_k^T}{\partial \dot{q}_i} \mathbf{v}_k + \mathbf{v}_k^T \frac{\partial \mathbf{v}_k}{\partial \dot{q}_i} \right) \\ &= \sum_k m_k \frac{d}{dt} \left( \mathbf{v}_k^T \frac{\partial \mathbf{v}_k}{\partial \dot{q}_i} \right) \\ &= \sum_k m_k \left( \frac{d \mathbf{v}_k^T}{dt} \cdot \frac{\partial \mathbf{v}_k}{\partial \dot{q}_i} + \mathbf{v}_k^T \cdot \frac{d}{dt} \left( \frac{\partial \mathbf{v}_k}{\partial \dot{q}_i} \right) \right). \end{aligned} \quad (4.6)$$

Note,  $\frac{d \mathbf{v}_k}{dt}$  represents the acceleration vector  $\mathbf{a}_k$  of the  $k$ th particle. Now,

$$\frac{\partial \mathbf{v}_k}{\partial \dot{q}_i} = \frac{\partial}{\partial \dot{q}_i} \left( \frac{d \mathbf{r}_k}{dt} \right) = \frac{\partial}{\partial \dot{q}_i} \left( \frac{\partial \mathbf{r}_k}{\partial q_i} \dot{q}_i \right) = \frac{\partial \mathbf{r}_k}{\partial q_i},$$

since  $\mathbf{r}_k$  does not depend on  $\dot{q}_i$ . Thus, (4.6) rewrites

$$\begin{aligned} \frac{d}{dt} \frac{\partial}{\partial \dot{q}_i} T(\mathbf{q}, \dot{\mathbf{q}}) &= \sum_k m_k \left( \mathbf{a}_k^T \cdot \frac{\partial \mathbf{r}_k}{\partial q_i} + \mathbf{v}_k^T \cdot \frac{d}{dt} \left( \frac{\partial \mathbf{r}_k}{\partial q_i} \right) \right), \\ &= \sum_k m_k \mathbf{a}_k^T \cdot \frac{\partial \mathbf{r}_k}{\partial q_i} + \sum_k m_k \mathbf{v}_k^T \cdot \frac{\partial \mathbf{v}_k}{\partial q_i}. \end{aligned}$$

Subtracting this equation with (4.5) yields

$$\frac{d}{dt} \frac{\partial}{\partial \dot{q}_i} T(\mathbf{q}, \dot{\mathbf{q}}) - \frac{\partial}{\partial q_i} T(\mathbf{q}, \dot{\mathbf{q}}) = \sum_k m_k \mathbf{a}_k^T \cdot \frac{\partial \mathbf{r}_k}{\partial q_i}.$$

From the second Newton law  $m_k \mathbf{a}_k$  is equal to the sum of forces applies to the  $k$ th particle. Note by  $\mathbf{f}_k(\mathbf{q})$  the resultant of all the forces derivating from a potential  $V$  (*i.e.*,  $\mathbf{f}_k$  is a function of  $\mathbf{q}$ ) and  $\mathbf{g}_k(t)$  the resultant of all other forces. We have  $\mathbf{f}_k + \mathbf{g}_k = m_k \mathbf{a}_k$ . Thus,

$$\frac{d}{dt} \frac{\partial}{\partial \dot{q}_i} T(\mathbf{q}, \dot{\mathbf{q}}) - \frac{\partial}{\partial q_i} T(\mathbf{q}, \dot{\mathbf{q}}) = \sum_k (\mathbf{f}_k^T(\mathbf{q}) + \mathbf{g}_k^T) \cdot \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i} \quad (4.7)$$

Now, the force  $\mathbf{f}_k(\mathbf{q})$  satisfies.

$$\mathbf{f}_k(\mathbf{q}) = -m_k \frac{\partial V_k(\mathbf{q})}{\partial \mathbf{r}_k}$$

where  $V_k(\mathbf{q})$  is the potential of the  $k$ th particle. Thus Equation 4.7 becomes

$$\begin{aligned} \frac{d}{dt} \frac{\partial}{\partial \dot{q}_i} T(\mathbf{q}, \dot{\mathbf{q}}) - \frac{\partial}{\partial q_i} T(\mathbf{q}, \dot{\mathbf{q}}) &= \sum_k \left( -m_k \frac{\partial V_k(\mathbf{q})}{\partial \mathbf{r}_k} \right) \cdot \left( \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i} \right) + \sum_k \mathbf{g}_k^T \cdot \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i}, \\ &= \sum_k \left( -m_k \frac{\partial V_k(\mathbf{q})}{\partial q_i} \right) + \sum_k \mathbf{g}_k^T \cdot \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i}, \\ &= -\frac{\partial}{\partial q_i} \sum_k m_k V_k(\mathbf{q}) + \sum_k \mathbf{g}_k^T \cdot \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i}. \end{aligned} \quad (4.8)$$

The quantity  $\sum_k m_k V_k(\mathbf{q})$  represents the potential energy  $U(\mathbf{q})$  of the system. Equation 4.8 rewrites into

$$\frac{d}{dt} \frac{\partial}{\partial \dot{q}_i} T(\mathbf{q}, \dot{\mathbf{q}}) - \frac{\partial}{\partial q_i} T(\mathbf{q}, \dot{\mathbf{q}}) = -\frac{\partial U(\mathbf{q})}{\partial q_i} + \sum_k \mathbf{g}_k^T \cdot \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i},$$

or equivalently

$$\frac{d}{dt} \frac{\partial}{\partial \dot{q}_i} (T(\mathbf{q}, \dot{\mathbf{q}}) - U(\mathbf{q})) - \frac{\partial}{\partial q_i} (T(\mathbf{q}, \dot{\mathbf{q}}) - U(\mathbf{q})) = \sum_k \mathbf{g}_k^T \cdot \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i},$$

due to the fact that  $\frac{\partial}{\partial \dot{q}_i} (U(\mathbf{q})) = 0$ . Set  $L(\mathbf{q}, \dot{\mathbf{q}}) = T(\mathbf{q}, \dot{\mathbf{q}}) - U(\mathbf{q})$ , we get the Euler-Lagrange equation :

$$\frac{d}{dt} \frac{\partial}{\partial \dot{q}_i} (L(\mathbf{q}, \dot{\mathbf{q}})) - \frac{\partial}{\partial q_i} (L(\mathbf{q}, \dot{\mathbf{q}})) = \sum_k \mathbf{g}_k^T \cdot \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i}.$$

The quantity  $u_i = \sum_k \mathbf{g}_k^T \cdot \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i}$  is called *generalized force*. It can be interpreted as a force since a small modification  $dq_i$  of  $q_i$  requires a work equal to

$$d\tau = \sum_k \mathbf{g}_k^T \cdot d\mathbf{r}_k = \sum_k \mathbf{g}_k^T \cdot \frac{\partial \mathbf{r}_k(\mathbf{q})}{\partial q_i} dq_i = u_i \cdot dq_i.$$

This equation expresses that the work input is equal to the product of force and displacement. Thus, the generalized force  $u_i$  can be defined as the ratio  $\frac{d\tau}{dq_i}$  where  $d\tau$  is the amount of work required to generate an increase of  $dq_i$  on the joint  $q_i$ . So  $u_i$  represents a force if  $q_i$  corresponds to a length, and a torque if  $q_i$  corresponds to an angle.

## 4.6 Modeling a simple pendulum using the Euler-Lagrange method

Let us now consider a simple example to illustrate the principle of the approach where the system is a simple pendulum.

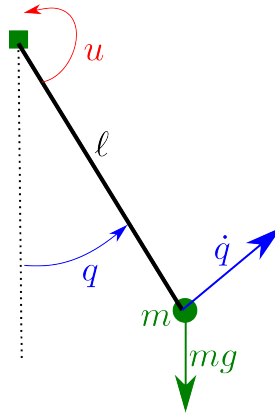


Figure 4.3: A simple pendulum

The variable  $q$  is the angle of the pendulum and  $\dot{q}$  is the angular velocity. The Lagrangian is given by

$$L(q, \dot{q}) = T(q, \dot{q}) - V(q) = \frac{1}{2}m (\ell\dot{q})^2 - \ell (1 - \cos q) mg.$$

The Euler-Lagrange equation is

$$\frac{d}{dt} \left( \frac{\partial L}{\partial \dot{q}} \right) - \left( \frac{\partial L}{\partial q} \right) = u,$$

where  $u$  is an external torque. Weight, for example, cannot be considered as external, as its influence is already taken into account in the Lagrangian via the potential energy  $V(q)$ . Since

$$\begin{aligned} \frac{\partial L}{\partial \dot{q}} &= m\ell^2 \dot{q} \\ \frac{\partial L}{\partial q} &= -\ell mg \sin q. \end{aligned}$$

The Euler-Lagrange equation rewrites as

$$\frac{d}{dt} (m\ell^2 \dot{q}) + \ell mg \sin q = u,$$

*i.e.*,

$$m\ell^2 \ddot{q} + \ell mg \sin q = u.$$

The state equation of the pendulum are thus

$$\begin{cases} \dot{q} &= \dot{q} \\ \ddot{q} &= \frac{-\ell mg \sin q + u}{m\ell^2} \end{cases}$$

# Exercises

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EXERCISE 15.– *Modeling an underwater robot*

See the correction video at <https://youtu.be/Ox8zavW9WYw>

The robot we will be modeling is the Redermor (*greyhound of the sea* in the Breton language). It is represented on Figure 4.4. It is an entirely autonomous underwater robot. This robot, developed by GESMA (Groupe d'Etude Sous-Marine de l'Atlantique - *Atlantic underwater research group*), has a length of 6 m, a diameter of 1 m and a weight of 3 800 kg. It has an efficient propulsion and control system with the aim of finding mines on the seabed.

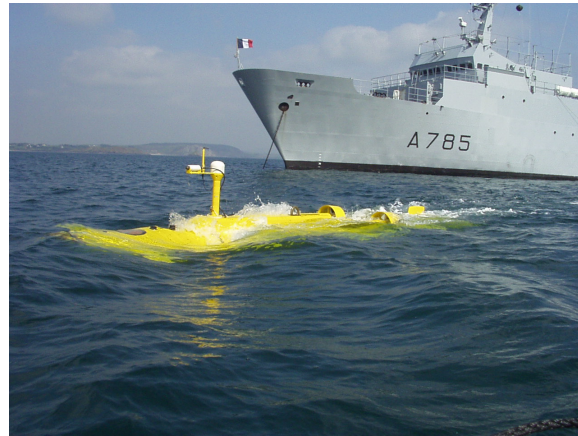


Figure 4.4: *Redermor* built by GESMA (Groupe d'Etude Sous-Marine de l'Atlantique - *Atlantic underwater research group*), on the water surface, still close to the boat it was launched from

Let us build a local coordinate system  $\mathcal{R}_0 : (\mathbf{o}_0, \mathbf{i}_0, \mathbf{j}_0, \mathbf{k}_0)$  over the area traveled by the robot. The point  $\mathbf{o}_0$  is placed on the surface of the ocean. The vector  $\mathbf{i}_0$  indicates north,  $\mathbf{j}_0$  points to the east and  $\mathbf{k}_0$  is oriented towards the center of the Earth. Let  $\mathbf{p} = (p_x, p_y, p_z)$  be the coordinates of the center of the robot expressed in the coordinate system  $\mathcal{R}_0$ . The state variables of the underwater robot are its position  $\mathbf{p}$  in the coordinate system  $\mathcal{R}_0$ , its tangential  $v$  and its three Euler angles  $\varphi, \theta, \psi$ . Its inputs are the tangential acceleration  $\dot{v}$  as well as three control rudders which act respectively on

$\omega_x, \omega_y, \omega_z$ . More formally, we have:

$$\begin{cases} u_1 &= \dot{v} \\ vu_2 &= \omega_y \\ vu_3 &= \omega_z \\ u_4 &= \omega_x \end{cases}$$

where the factor  $v$  preceding  $u_2, u_3$  indicates that the robot is only able to turn left/right (through  $u_3$ ) or up/down (through  $u_2$ ) when it is advancing. Give the kinematic state model for this system.

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EXERCISE 16.– *Dzhanibekov effect*

See the correction video at <https://youtu.be/EA1Rh5MgGKI>

The Dzhanibekov effect (discovered by Poinot [15]) corresponds to the instability of the rotation of a rigid body around its second principal axes. We consider a parallelepiped of mass  $m$  given in Figure 4.5. We assume that the density is uniform.

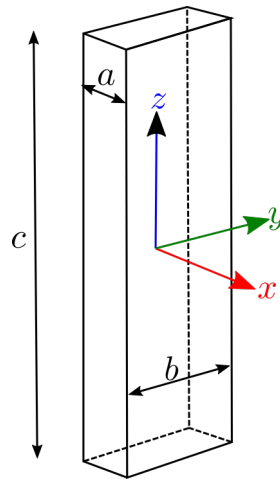


Figure 4.5: Parallelepiped body used to illustrate the Dzhanibekov effect

1) Recall that the inertia matrix of a solid body occupying the volume  $V$  is

$$\mathbf{I} = \int_V \rho(x, y, z) \begin{pmatrix} y^2 + z^2 & -xy & -xz \\ -xy & x^2 + z^2 & -yz \\ -xz & -yz & x^2 + y^2 \end{pmatrix} dx dy dz$$

where  $\rho(x, y, z)$  is the density. Compute the inertia matrix of the parallelepiped.

2) The Euler equation of a free rigid body is given by

$$\dot{\boldsymbol{\omega}}_r = -\mathbf{I}^{-1} \cdot (\boldsymbol{\omega}_r \wedge (\mathbf{I} \cdot \boldsymbol{\omega}_r))$$

where  $\boldsymbol{\omega}_r = (\omega_1, \omega_2, \omega_3)$  is the rotation vector expressed in the body frame. Give the state equation governing the rotation of our parallelepiped. The state vector is taken as  $(\omega_1, \omega_2, \omega_3)$ .

3) Give the equilibrium points of the system in the state-space. Conclude about the rotation of any solid asteroid in the rotating space.

4) Study the stability of the rotation along the steady states.

5) Take  $a = 0.4, b = 1, c = 3, m = 1$ . Illustrate by a simulation, an unstable situation with a rotation of the parallelepiped rotating along one of its principle axis.

### EXERCISE 17.– *Euler field*

See the correction video at <https://youtu.be/in5wrQkVv4A>

The goal of this exercise is to provide a geometrical interpretation of the behavior of a solid body in rotation. We consider the Euler equation given by

$$\dot{\boldsymbol{\omega}}_r = -\mathbf{I}^{-1} \cdot (\boldsymbol{\omega}_r \wedge (\mathbf{I} \cdot \boldsymbol{\omega}_r))$$

where  $\boldsymbol{\omega}_r$  is the rotation vector expressed in the body frame and  $\mathbf{I}$  is the inertia matrix given by

$$\mathbf{I} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix}$$

1) Show that the kinetic energy given by

$$E_K = \frac{1}{2} \boldsymbol{\omega}_r^T \mathbf{I} \boldsymbol{\omega}_r$$

is constant.

2) Draw in three dimensional figure, the ellipsoid  $\mathcal{E}$  corresponding to the set of all  $\boldsymbol{\omega}_r$  such that energy equal to  $E_0 = 10\text{J}$ . Draw the vector field describing the evolution of  $\boldsymbol{\omega}_r$  on this ellipsoid.

3) Simulate with a Runge-Kutta method the evolution of the rotation vector for an initial condition  $\boldsymbol{\omega}_r(0) = (4, \sqrt{2}, 0)$ . Draw the corresponding trajectory on the ellipsoid  $\mathcal{E}$ . Interpret the result.

4) Simulate an evolution for  $\boldsymbol{\omega}_r(0) = (0.01, \sqrt{10}, 0)$ . Conclude.

### EXERCISE 18.– *Flat disk*

See the correction video at <https://youtu.be/Ra918LJ08WQ>

Consider a disk spinning in the space without any gravity. The disk is assumed to be flat, *i.e.*, its inertia matrix

$$\mathbf{I} = \begin{pmatrix} I_1 & 0 & 0 \\ 0 & I_2 & 0 \\ 0 & 0 & I_3 \end{pmatrix}$$

should satisfy (i) the positivity condition:  $I_1 > 0, I_2 > 0, I_3 > 0$  and (ii) the flatness condition:  $I_1 = I_2 + I_3$ .

We assume that we can control some internal forces in a strap down manner. It means that we have no inertial wheels to control the rotation of the disk. Instead, we could slightly modify  $\mathbf{I}$  by dilatation or compression of some parts of the disk. We also consider that we are not able to change the center of gravity of the disk.

At rest, the inertial matrix is assumed to be

$$\bar{\mathbf{I}} = \begin{pmatrix} \bar{I}_1 & 0 & 0 \\ 0 & \bar{I}_2 & 0 \\ 0 & 0 & \bar{I}_3 \end{pmatrix} = \begin{pmatrix} \frac{1}{2}mr^2 & 0 & 0 \\ 0 & \frac{1}{4}mr^2 & 0 \\ 0 & 0 & \frac{1}{4}mr^2 \end{pmatrix}$$

where  $r$  is the radius of the disk and  $m$  is its mass. We assume that, we have two symmetric pairs of masses which can move along the  $y$  and  $z$  axis as in Figure 4.6.

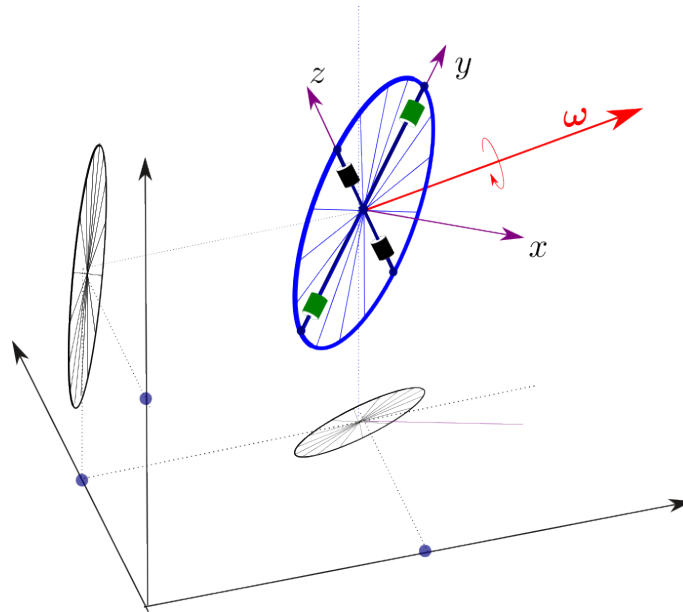


Figure 4.6: The disk spins and the two pairs of masses (black and green) can slightly move in order to control the motion of the rotation vector  $\boldsymbol{\omega}$

The inertia of the disk is now

$$\mathbf{I} = \begin{pmatrix} \frac{1}{2}mr^2 + \frac{1}{2}\ell_2^2 + \frac{1}{2}\ell_3^2 & 0 & 0 \\ 0 & \frac{1}{4}mr^2 + \frac{1}{2}\ell_3^2 & 0 \\ 0 & 0 & \frac{1}{4}mr^2 + \frac{1}{2}\ell_2^2 \end{pmatrix}$$

where  $\ell_2$  corresponds to the distance of the  $y$ -masses (green) to the center and  $\ell_3$  is the distance of the  $z$ -masses (black) to the center. Our objective is to propose a controller to modify the rotation of the disk.

1) Denote by  $\mathbf{R}$  the orientation of the flat disk and by  $\boldsymbol{\omega}_r = (\omega_1, \omega_2, \omega_3)$  the spin vector in the body frame. Taking into account the time invariance of the angular momentum, show that state equation to which obeys the disk can be given by,

$$\begin{cases} \dot{\omega}_1 = -\frac{\omega_1}{I_2+I_3}(u_1 + u_2) - \frac{I_3-I_2}{I_2+I_3}\omega_2\omega_3 \\ \dot{\omega}_2 = -\frac{\omega_2}{I_2}u_1 - \omega_3\omega_1 \\ \dot{\omega}_3 = -\frac{\omega_3}{I_3}u_2 + \omega_1\omega_2 \\ \dot{I}_2 = u_1 \\ \dot{I}_3 = u_2 \end{cases}$$

where

$$\begin{aligned} u_1 &= \ell_2 \dot{\ell}_2 \\ u_2 &= \ell_3 \dot{\ell}_3 \end{aligned}$$

2) To find a controller for our system, we follow a Lyapunov control approach [16]. For this, we choose a positive function  $V(\mathbf{x})$  such that  $V(\mathbf{x}) = 0$  when the objective is reached. Give an expression for a controller which tends to decrease  $V(\mathbf{x})$ .

3) *Alignment control.* We want the disk spins around one principal axis of the body. For instance its first axis, *i.e.*, the  $x$ -axis. We define the objective function

$$V(\mathbf{x}) = \frac{1}{2}(\omega_2^2 + \omega_3^2 + (I_2 - \bar{I}_2)^2 + (I_3 - \bar{I}_3)^2).$$

The quantity  $\omega_2^2 + \omega_3^2$  corresponds to the *alignment error*. Note that when  $V(\mathbf{x}) = 0$ , we have  $\omega_2 = \omega_3 = 0$  (the alignment is performed) and  $\mathbf{I} = \bar{\mathbf{I}}$  (the inertia matrix is at its nominal position). Propose a controller which tends to decrease  $V(\mathbf{x})$ . Illustrate by a simulation.

4) *Passivity control.* We now want the disk losses energy with the intuition that it may limit the precession. We define the *mechanical energy* as

$$V(\mathbf{x}) = \frac{1}{2}I_1\omega_1^2 + \frac{1}{2}I_2\omega_2^2 + \frac{1}{2}I_3\omega_3^2 + \frac{1}{2}(I_2 - \bar{I}_2)^2 + \frac{1}{2}(I_3 - \bar{I}_3)^2.$$

The quantity  $\frac{1}{2}\boldsymbol{\omega}_r^T \mathbf{I} \boldsymbol{\omega}_r$  corresponds to kinetic energy and  $\frac{1}{2}(I_2 - \bar{I}_2)^2 + \frac{1}{2}(I_3 - \bar{I}_3)^2$  is the artificial potential energy. Propose a controller which tends to decrease  $V(\mathbf{x})$ . Illustrate by a simulation.

5) *Precession controller.* To cancel it, we propose to take the objective function

$$V(\mathbf{x}) = \frac{1}{2}((I_3 - I_2)\omega_2\omega_3)^2 + \frac{1}{2}((I_1 - I_3)\omega_3\omega_1)^2 + \frac{1}{2}((I_2 - I_1)\omega_1\omega_2)^2 + \frac{1}{2}(I_2 - \bar{I}_2)^2 + \frac{1}{2}(I_3 - \bar{I}_3)^2.$$

The quantity

$$((I_3 - I_2)\omega_2\omega_3)^2 + ((I_1 - I_3)\omega_3\omega_1)^2 + ((I_2 - I_1)\omega_1\omega_2)^2 = \|\boldsymbol{\omega}_r \wedge \mathbf{I} \cdot \boldsymbol{\omega}_r\|^2$$

can be interpreted precession energy, *i.e.*, a part of the kinetic energy that creates the precession and that can be recovered. Indeed, when  $\boldsymbol{\omega}_r \wedge \mathbf{I} \cdot \boldsymbol{\omega}_r = \mathbf{0}$ , we see from the Euler equation that  $\dot{\boldsymbol{\omega}}_r = \mathbf{0}$

when  $\mathbf{u} = \mathbf{0}$ , which means that we have no more precession. Implement the corresponding controller and illustrate by a simulation.

6) Assume that at time  $t = 0$ , the disk spins around the second axis:

$$\boldsymbol{\omega}_r(0) = (10^{-5}, 10, 0)$$

and that we want it spins around the first axis. Combine the previous controllers to achieve this goal. Validate on a simulation.

### EXERCISE 19.– *Inverted rod pendulum*

See the correction video at <https://youtu.be/bmwtuAcF6iY>

Consider the inverted rod pendulum represented on Figure 4.7, composed of a pendulum placed on a carriage.

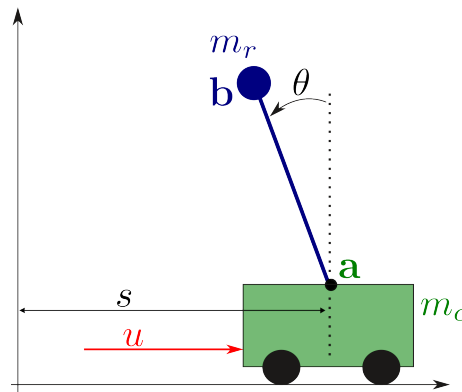


Figure 4.7: Inverted pendulum

The value  $u$  is the force exerted on the carriage of mass  $m_c = 5 \text{ kg}$ ,  $s$  indicates the position of the carriage,  $\theta$  is the angle between the pendulum and the vertical. At the tip  $\mathbf{b}$  of the pendulum of length  $\ell = 1 \text{ m}$  is a fixed mass  $m_r = 1 \text{ Kg}$ . Finally,  $\mathbf{a}$  is the point of articulation between the rod and the carriage.

1) Define  $\mathbf{q} = (s, \theta)$  the vector of generalized coordinates. Show that the Lagrangian is

$$L(\mathbf{q}, \dot{\mathbf{q}}) = \frac{m_c + m_r}{2} \dot{s}^2 - m_r \ell \dot{s} \dot{\theta} \cos \theta + \frac{m_r \ell^2}{2} \dot{\theta}^2 - m_r g \ell \cos \theta$$

2) Using `Sympy`, write the Euler-Lagrange equations

3) Derive the state equations of the system using `Sympy` as much as possible.

4) Propose a simulation of the inverted rod pendulum using a Runge-Kutta method.

EXERCISE 20.– *Rolling disk*

See the correction video at <https://youtu.be/VAKjnEcIjXg>

Consider a disk rolling on a plane without friction nor sliding, as shown on Figure 4.8.

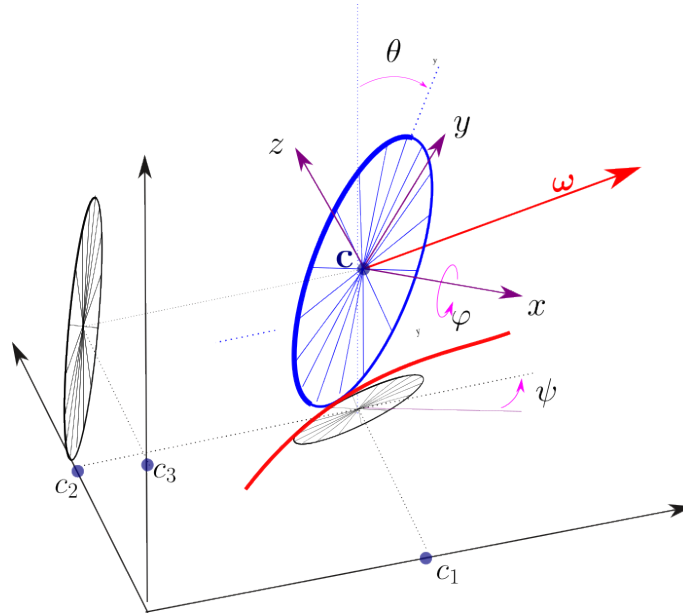


Figure 4.8: Disk (blue) rolling on a plane. The vertical and horizontal projections are painted black

We assume that the disk mass is  $m = 5\text{kg}$  and its radius is  $r = 1\text{m}$ . The gravity is taken as  $g = 9.81\text{ms}^{-2}$ .

In this exercise, we want to find the state equations describing the motion of the disk. The Lagrangian approach, often applied to model robots, will be chosen.

The state vector is chosen as  $\mathbf{x} = (c_1, c_2, \varphi, \theta, \psi, \dot{\varphi}, \dot{\theta}, \dot{\psi})$  where  $(c_1, c_2)$  is the vertical projection of center  $\mathbf{c} = (c_1, c_2, c_3)$  of the disk and  $\varphi, \theta, \psi$  are the three Euler angle. As illustrated by Figure 4.8,

- $\varphi$  is the spin angle
- $\theta$  is the stand angle, *i.e.*, when  $\theta = 0$ , the disk is vertical
- $\psi$  is the heading, *i.e.*, the horizontal orientation of the disk.

1) Give an expression of the Lagrangian  $\mathcal{L}$  with respect to the state variables.

2) If the ground is a flat frozen lake where the disk can slide in both direction (horizontally and laterally), then the state vector is

$$(\mathbf{q}, \dot{\mathbf{q}}) = (c_1, c_2, \varphi, \theta, \psi, \dot{c}_1, \dot{c}_2, \dot{\varphi}, \dot{\theta}, \dot{\psi}),$$

Using the Euler Lagrange equations, given by

$$\underbrace{\frac{d}{dt} \left( \frac{\partial \mathcal{L}}{\partial \dot{\mathbf{q}}} \right) - \frac{\partial \mathcal{L}}{\partial \mathbf{q}}}_{\mathcal{Q}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}})} = \boldsymbol{\tau}$$

where  $\boldsymbol{\tau}$  are the external constraint forces, find a differential equation describing the motions of the disk sliding on the ice.

3) Assume now that, no sliding is possible and the disk can only roll. Show that these rolling constraint on  $(\mathbf{q}, \dot{\mathbf{q}})$  are linked by the differential equations:

$$\begin{aligned} \dot{c}_1 &= r \sin \psi \cdot \dot{\varphi} + r \cos \psi \cos \theta \cdot \dot{\theta} - r \sin \psi \sin \theta \cdot \dot{\psi} \\ \dot{c}_2 &= -r \cos \psi \cdot \dot{\varphi} + r \sin \psi \cos \theta \cdot \dot{\theta} + r \cos \psi \sin \theta \cdot \dot{\psi} \end{aligned}$$

4) We have no friction no external forces to thrust or slow down the disk. The only external forces that apply to the disk are the reaction of the ground on the disk. The principle of d'Alembert states that the rolling constraints do not work. It can be used to find an expression for  $\boldsymbol{\tau}$  which occurs in the right hand side of the Euler-Lagrange equation. Show that

$$\boldsymbol{\tau} = \lambda_1 \cdot \begin{pmatrix} 1 \\ 0 \\ -r \sin \psi \\ -r \cos \psi \cos \theta \\ r \sin \psi \sin \theta \end{pmatrix} + \lambda_2 \cdot \begin{pmatrix} 0 \\ 1 \\ r \cos \psi \\ -r \sin \psi \cos \theta \\ -r \cos \psi \sin \theta \end{pmatrix}$$

where  $\lambda_1$  and  $\lambda_2$  are the Lagrange parameters which depend on  $\mathbf{q}$ .

5) Deduce the state equations of the rolling disk with a state vector given by  $\mathbf{x} = (c_1, c_2, \varphi, \theta, \psi, \dot{\varphi}, \dot{\theta}, \dot{\psi})$ .

6) Simulate the rolling disk with the following initial state

$$(c_1, c_2, \varphi, \theta, \psi, \dot{\varphi}, \dot{\theta}, \dot{\psi}) = (2, 0, 0, 0.1, 0, 2.5, 0, 0)$$

for  $t \in [0, 10]$ . The simulation should behave as illustrated by Figure 4.9 where the blue disk corresponds to the initial state.

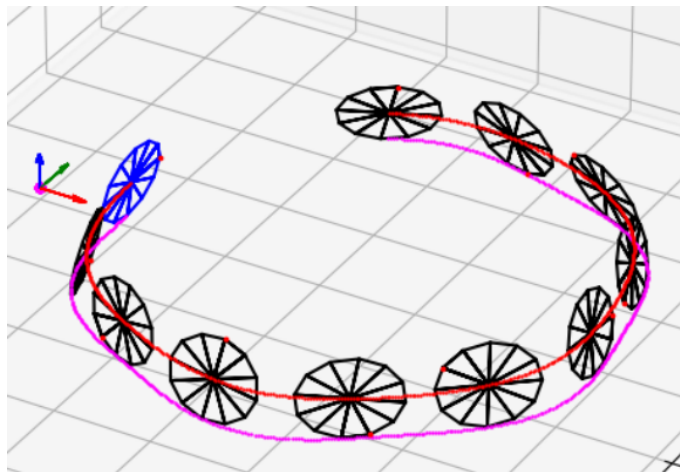


Figure 4.9: Disk rolling on a plane with precession

7) Find some initial conditions for the disk so that the simulation generates a perfect a circular trajectory.



# Chapter 5

## Inertial control

In this chapter, we consider the equations (4.2) given by

$$\begin{cases} \dot{\mathbf{p}} &= \mathbf{R} \cdot \mathbf{v}_r & (i) \\ \dot{\mathbf{R}} &= \mathbf{R} \cdot (\boldsymbol{\omega}_r \wedge) & (ii) \\ \dot{\mathbf{v}}_r &= \mathbf{R}^T \cdot \mathbf{g} + \frac{1}{m} \cdot \mathbf{f}_r - \boldsymbol{\omega}_r \wedge \mathbf{v}_r & (iii) \\ \dot{\boldsymbol{\omega}}_r &= \mathbf{I}^{-1} \cdot (\boldsymbol{\tau}_r - \boldsymbol{\omega}_r \wedge \mathbf{I} \cdot \boldsymbol{\omega}_r) & (iv) \end{cases} \quad (5.1)$$

This model can be qualified as *inertial* since two inertial parameters are involved : the mass  $m$  and the inertia matrix  $\mathbf{I}$ . A control loop on such a system is also qualified as *inertial* since it requires the knowledge of these two inertial parameters only.

In this chapter, we show how to choose the forces  $\mathbf{f}_r$  and the torque  $\boldsymbol{\tau}_r$  to track a virtual pose represented by a desired position  $\mathbf{p}_d(t)$  and a desired orientation  $\mathbf{R}_d(t)$ . We assume these quantities have a closed form and thus their two first derivatives  $\dot{\mathbf{p}}_d, \ddot{\mathbf{p}}_d, \dot{\mathbf{R}}_d, \ddot{\mathbf{R}}_d$ , are available.

### 5.1 Control the accelerations

We first note that if we choose

$$\begin{cases} \mathbf{f}_r &= m\dot{\mathbf{v}}_r^d - m\mathbf{R}^T \cdot \mathbf{g} + m\boldsymbol{\omega}_r \wedge \mathbf{v}_r \\ \boldsymbol{\tau}_r &= \mathbf{I} \cdot \dot{\boldsymbol{\omega}}_r^d + \boldsymbol{\omega}_r \wedge \mathbf{I} \cdot \boldsymbol{\omega}_r \end{cases}$$

where the new inputs are  $\dot{\mathbf{v}}_r^d$  and  $\dot{\boldsymbol{\omega}}_r^d$  the values we want for  $\dot{\mathbf{v}}_r$  and  $\dot{\boldsymbol{\omega}}_r$ , the system (5.1) simplifies into

$$\begin{cases} \dot{\mathbf{p}} &= \mathbf{R} \cdot \mathbf{v}_r \\ \dot{\mathbf{R}} &= \mathbf{R} \cdot (\boldsymbol{\omega}_r \wedge) \\ \dot{\mathbf{v}}_r &= \dot{\mathbf{v}}_r^d \\ \dot{\boldsymbol{\omega}}_r &= \dot{\boldsymbol{\omega}}_r^d \end{cases}$$

We now have a model which is purely kinematic. The parameters  $m$  and  $\mathbf{I}$  do not occur anymore.

## 5.2 Positioner

The *positioner* is a controller which computes the acceleration  $\dot{\mathbf{v}}_r^d$  to apply to the robot to reach a desired target  $\mathbf{p}_d(t)$ . More precisely, we want an expression of  $\dot{\mathbf{v}}_r^d$  with respect to  $\mathbf{p}_d(t)$  and the state variables  $\mathbf{p}, \mathbf{R}, \mathbf{v}_r, \boldsymbol{\omega}_r$ . We define the error

$$\mathbf{e} = \mathbf{p}_d - \mathbf{p}.$$

To cancel this error, by choosing the right acceleration, we take here a proportional and derivative controller. For this, we want the error to satisfy the error equation

$$\ddot{\mathbf{e}} + \alpha_1 \dot{\mathbf{e}} + \alpha_0 \mathbf{e} = \mathbf{0}$$

where  $\alpha_0, \alpha_1$  are chosen to generate a stable characteristic polynomial.

In the robot frame, the error equation becomes  $\ddot{\mathbf{e}}_r + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r = \mathbf{0}$ , where

$$\begin{aligned} \mathbf{e}_r &= \mathbf{R}^T \mathbf{e} = \mathbf{R}^T (\mathbf{p}_d - \mathbf{p}) \\ \dot{\mathbf{e}}_r &= \mathbf{R}^T \dot{\mathbf{e}} = \mathbf{R}^T (\dot{\mathbf{p}}_d - \dot{\mathbf{p}}) = \mathbf{R}^T \dot{\mathbf{p}}_d - \mathbf{v}_r \\ \ddot{\mathbf{e}}_r &= \mathbf{R}^T \ddot{\mathbf{e}} = \mathbf{R}^T (\ddot{\mathbf{p}}_d - \ddot{\mathbf{p}}) \\ &= \mathbf{R}^T \ddot{\mathbf{p}}_d - \mathbf{R}^T (\dot{\mathbf{R}} \mathbf{v}_r + \mathbf{R} \dot{\mathbf{v}}_r) \\ &= \mathbf{R}^T \ddot{\mathbf{p}}_d - \boldsymbol{\omega}_r \wedge \mathbf{v}_r - \dot{\mathbf{v}}_r \end{aligned}$$

Thus

$$\begin{aligned} \ddot{\mathbf{e}}_r + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r &= \mathbf{0} \\ \Leftrightarrow \mathbf{R}^T \ddot{\mathbf{p}}_d - \boldsymbol{\omega}_r \wedge \mathbf{v}_r - \dot{\mathbf{v}}_r + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r &= \mathbf{0} \\ \Leftrightarrow \dot{\mathbf{v}}_r = \mathbf{R}^T \ddot{\mathbf{p}}_d - \boldsymbol{\omega}_r \wedge \mathbf{v}_r + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r \end{aligned}$$

Since  $\dot{\mathbf{v}}_r = \dot{\mathbf{v}}_r^d$  the expression of the positioner is thus

$$\dot{\mathbf{v}}_r^d = \mathbf{R}^T \ddot{\mathbf{p}}_d - \boldsymbol{\omega}_r \wedge \mathbf{v}_r + \alpha_1 (\mathbf{R}^T \dot{\mathbf{p}}_d - \mathbf{v}_r) + \alpha_0 \mathbf{R}^T (\mathbf{p}_d - \mathbf{p}).$$

This means that if we are able to produce the forces and torques that generate the acceleration  $\dot{\mathbf{v}}_r$  then, the error  $\mathbf{e} = \mathbf{p}_d - \mathbf{p}$  will converge to zero.

## 5.3 Orientator

The *orientator* is a controller which tells us how to change the rotation vector in order to follow a desired orientation  $\mathbf{R}_d(t)$ , *i.e.*, we want an expression of  $\dot{\boldsymbol{\omega}}_r^d$  with respect to  $\mathbf{R}_d$  and the state variables  $\mathbf{p}, \mathbf{R}, \mathbf{v}_r, \boldsymbol{\omega}_r$ .

We need to recall that

$$\frac{d}{dt} \log(\mathbf{R}(t)) = \mathbf{R}^T(t) \cdot \dot{\mathbf{R}}(t). \quad (5.2)$$

Indeed,

$$\begin{aligned}
& \frac{d}{dt} \exp(\log(\mathbf{R}(t))) = \frac{d}{dt} \mathbf{R}(t) \\
\Rightarrow & \exp(\log(\mathbf{R}(t))) \cdot \frac{d}{dt} \log(\mathbf{R}(t)) = \dot{\mathbf{R}}(t) \\
\Rightarrow & \frac{d}{dt} \log(\mathbf{R}(t)) = \mathbf{R}^T(t) \cdot \dot{\mathbf{R}}(t)
\end{aligned}$$

In  $SO(3)$ , the orientation error can be represented by  $\mathbf{R}_d \cdot \mathbf{R}^T$ . Now, this quantity cannot be qualified as an error since  $SO(3)$  has no addition operator. In the corresponding Lie algebra, we define the error as

$$\mathbf{e} = \text{Log}(\mathbf{R}_d \mathbf{R}^T)$$

where  $\mathbf{e}$  corresponds to the rotation vector we have to follow for 1sec to go from  $\mathbf{R}$  to  $\mathbf{R}_d$ . To cancel thus error, by choosing the right angular acceleration  $\dot{\boldsymbol{\omega}}_r^d$ , we can consider a proportional and derivative controller. For this, we want the error to satisfy

$$\ddot{\mathbf{e}} + \alpha_1 \dot{\mathbf{e}} + \alpha_0 \mathbf{e} = \mathbf{0},$$

where  $\alpha_0, \alpha_1$  are chosen to generate a stable characteristic polynomial. Now,

$$\begin{aligned}
\mathbf{e} \wedge &= \log(\mathbf{R}_d \mathbf{R}^T) \\
\dot{\mathbf{e}} \wedge &= \frac{d}{dt} \log(\mathbf{R}_d \mathbf{R}^T) \\
&\stackrel{(5.2)}{=} \mathbf{R} \mathbf{R}_d^T \cdot \frac{d}{dt} (\mathbf{R}_d \mathbf{R}^T) \\
&= \mathbf{R} \mathbf{R}_d^T \cdot (\mathbf{R}_d \dot{\mathbf{R}}^T + \dot{\mathbf{R}}_d \mathbf{R}^T) \\
&= \underbrace{\mathbf{R} \dot{\mathbf{R}}^T}_{=-(\boldsymbol{\omega} \wedge)} + \mathbf{R} \underbrace{\mathbf{R}_d^T \dot{\mathbf{R}}_d}_{=(\mathbf{R}_d^T \boldsymbol{\omega}_d) \wedge} \mathbf{R}^T \\
&\stackrel{(1.3)}{=} -(\boldsymbol{\omega} \wedge) + ((\mathbf{R} \cdot \mathbf{R}_d^T \boldsymbol{\omega}_d) \wedge)
\end{aligned}$$

Thus

$$\begin{aligned}
\mathbf{e} &= \text{Log}(\mathbf{R}_d \mathbf{R}^T) \\
\dot{\mathbf{e}} &= -\boldsymbol{\omega} + \mathbf{R} \mathbf{R}_d^T \boldsymbol{\omega}_d \\
\ddot{\mathbf{e}} &= -\dot{\boldsymbol{\omega}} + \dot{\mathbf{R}} \mathbf{R}_d^T \boldsymbol{\omega}_d + \mathbf{R} \dot{\mathbf{R}}_d^T \boldsymbol{\omega}_d + \mathbf{R} \mathbf{R}_d^T \dot{\boldsymbol{\omega}}_d
\end{aligned}$$

In the robot frame, the error equation  $\ddot{\mathbf{e}} + \alpha_1 \dot{\mathbf{e}} + \alpha_0 \mathbf{e} = \mathbf{0}$ , becomes  $\ddot{\mathbf{e}}_r + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r = \mathbf{0}$ , where

$$\begin{aligned}
\mathbf{e}_r &= \mathbf{R}^T \mathbf{e} = \mathbf{R}^T \text{Log}(\mathbf{R}_d \mathbf{R}^T) \\
\dot{\mathbf{e}}_r &= \mathbf{R}^T \dot{\mathbf{e}} = -\mathbf{R}^T \boldsymbol{\omega} + \mathbf{R}_d^T \boldsymbol{\omega}_d \\
\ddot{\mathbf{e}}_r &= \mathbf{R}^T \ddot{\mathbf{e}} = -\underbrace{\mathbf{R}^T \dot{\boldsymbol{\omega}}}_{=\dot{\boldsymbol{\omega}}_r} + \underbrace{\mathbf{R}^T \dot{\mathbf{R}} \mathbf{R}_d^T \boldsymbol{\omega}_d}_{=\boldsymbol{\omega}_r \wedge} + \dot{\mathbf{R}}_d^T \boldsymbol{\omega}_d + \mathbf{R}_d^T \dot{\boldsymbol{\omega}}_d
\end{aligned} \tag{5.3}$$

The differential equation for the error is

$$\begin{aligned}
\mathbf{0} &= \ddot{\mathbf{e}}_r + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r \\
\Leftrightarrow \mathbf{0} &\stackrel{(5.3)}{=} -\dot{\boldsymbol{\omega}}_r + (\boldsymbol{\omega}_r \wedge) \cdot \mathbf{R}_d^T \boldsymbol{\omega}_d + \dot{\mathbf{R}}_d^T \boldsymbol{\omega}_d + \mathbf{R}_d^T \dot{\boldsymbol{\omega}}_d + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r \\
\Leftrightarrow \dot{\boldsymbol{\omega}}_r &= (\boldsymbol{\omega}_r \wedge) \mathbf{R}_d^T \boldsymbol{\omega}_d + \dot{\mathbf{R}}_d^T \boldsymbol{\omega}_d + \mathbf{R}_d^T \dot{\boldsymbol{\omega}}_d + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r
\end{aligned}$$

Since  $\dot{\boldsymbol{\omega}}_r^d = \dot{\boldsymbol{\omega}}_r$ , we get

$$\dot{\boldsymbol{\omega}}_r^d = (\boldsymbol{\omega}_r \wedge) \mathbf{R}_d^T \boldsymbol{\omega}_d + \dot{\mathbf{R}}_d^T \boldsymbol{\omega}_d + \mathbf{R}_d^T \dot{\boldsymbol{\omega}}_d + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r$$

As a consequence, to provide the robot an orientation  $\mathbf{R}_d(t)$  we can take the following controller

Orientator (in : $\mathbf{R}_d, \dot{\mathbf{R}}_d, \ddot{\mathbf{R}}_d, \mathbf{R}, \boldsymbol{\omega}_r$ ; out: $\dot{\boldsymbol{\omega}}_r^d$ )	
1	$\boldsymbol{\omega}_d = \wedge^{-1}(\dot{\mathbf{R}}_d \cdot \mathbf{R}_d^T)$
2	$\dot{\boldsymbol{\omega}}_d = \wedge^{-1}(\dot{\mathbf{R}}_d \cdot \mathbf{R}_d^T + \ddot{\mathbf{R}}_d \cdot \mathbf{R}_d^T)$
3	$\mathbf{e}_r = \mathbf{R}^T \cdot \text{Log}(\mathbf{R}_d \mathbf{R}^T)$
4	$\dot{\mathbf{e}}_r = -\boldsymbol{\omega}_r + \mathbf{R}_d^T \boldsymbol{\omega}_d$
5	$\dot{\boldsymbol{\omega}}_r^d = \left( (\boldsymbol{\omega}_r \wedge) \mathbf{R}_d^T + \dot{\mathbf{R}}_d^T \right) \cdot \boldsymbol{\omega}_d + \mathbf{R}_d^T \dot{\boldsymbol{\omega}}_d + \alpha_1 \dot{\mathbf{e}}_r + \alpha_0 \mathbf{e}_r$

We can take a characteristic polynomial for the error which is stable as for instance:  $(s + 1) = s^2 + 2s + 1$ . In this case,  $\alpha_0 = 1$  and  $\alpha_1 = 2$ .

## 5.4 Controller

As illustrated by Figure 5.1, we are now able to generate the right resultant  $\mathbf{f}_r$  and the right torque  $\boldsymbol{\tau}_r$  in order to track a time dependent position  $\mathbf{p}_d(t)$  with the right time dependent orientation  $\mathbf{R}_d(t)$ .

To get generate the torque  $\boldsymbol{\tau}_r$ , we can either use the external forces such as propellers or use inertial disks inside the robot.

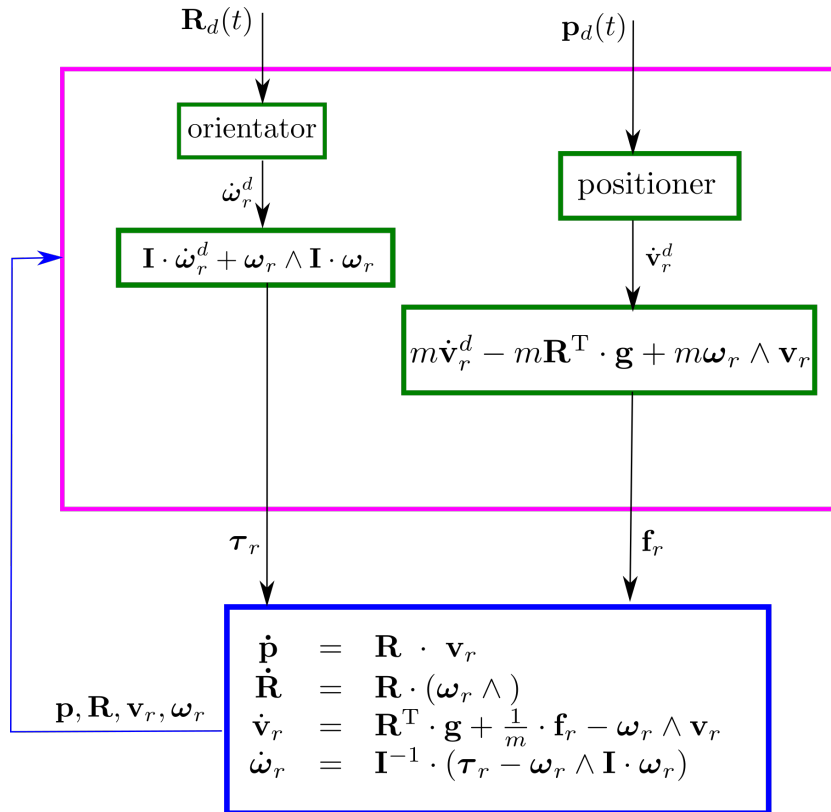


Figure 5.1: Orientator and positioner

## 5.5 Two-dimensional robots

If the robot moves in the plane, then, we can still apply Equation 5.1 which now becomes simpler. This is illustrated by Figure 5.2. For a two dimensional robot, we get

$$\mathbf{R} = \begin{pmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{pmatrix}, \mathbf{p} = \begin{pmatrix} p_1 \\ p_2 \\ 0 \end{pmatrix}, \boldsymbol{\omega}_r = \begin{pmatrix} 0 \\ 0 \\ \omega \end{pmatrix},$$

$$\boldsymbol{\omega}_r \wedge = \begin{pmatrix} 0 & -\omega & 0 \\ \omega & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \dot{\boldsymbol{\omega}}_r = \begin{pmatrix} 0 \\ 0 \\ \dot{\omega} \end{pmatrix}, \mathbf{f}_r = \begin{pmatrix} f_{r1} \\ f_{r2} \\ 0 \end{pmatrix}$$

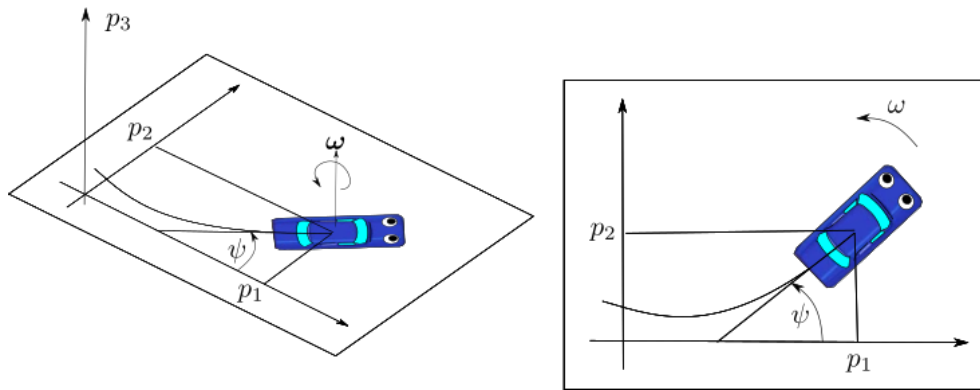


Figure 5.2: Robot in the plane

If we select only the  $x, y$  components as represented in Figure 5.2, right, we get

$$\begin{cases} \dot{\mathbf{p}} &= \mathbf{R} \cdot \mathbf{v}_r \\ \dot{\mathbf{R}} &= \mathbf{R} \cdot \boldsymbol{\omega} \wedge \\ \dot{\mathbf{v}}_r &= \frac{1}{m} \cdot \mathbf{f}_r - (\boldsymbol{\omega} \wedge) \mathbf{v}_r \\ \dot{\omega} &= \frac{\tau}{I_{33}} \end{cases} \quad (5.4)$$

In this equation,  $\omega$  is a scalar and we have

$$\boldsymbol{\omega} \wedge = \begin{pmatrix} 0 & -\omega \\ \omega & 0 \end{pmatrix}.$$

There is no need to specify if the relation is expressed in the robot frame or in the world frame: they are the same. Moreover,

$$\mathbf{R} = \begin{pmatrix} \cos \psi & -\sin \psi \\ \sin \psi & \cos \psi \end{pmatrix}$$

and the vectors  $\mathbf{v}_r, \mathbf{f}_r$  are both two dimensional. The main difference between Equations (4.2) and (5.4) is the Euler equation which collapses into  $\dot{\omega} = \frac{\tau}{I_{33}}$ . This is due to the fact that  $\omega, \tau$  are now both scalar instead of three dimensional vectors.

The equations for the orientator and the positioner remain valid.



# Exercises

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EXERCISE 21.– *Lie bracket for control*

See the correction video at <https://youtu.be/PfitQ7ZuNZs>

The motivation of Lie bracket for control is illustrated by Figure 5.3 on the car parking problem. We would like to move the blue car sideways between the two red cars already parked. Moving sideways is not possible, but we can approach this direction using many small forward-backward maneuvers. Building this new sideways direction corresponds to the Lie bracket operator between two dynamics. The goal of this exercise is to illustrate Lie brackets on a simple Dubins car.

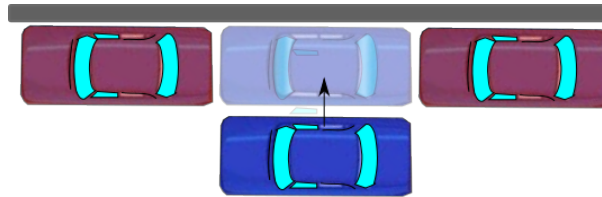


Figure 5.3: How to maneuver in order to park the blue car

Consider two vector fields  $\mathbf{f}$  and  $\mathbf{g}$  of  $\mathbb{R}^n$  corresponding to two state equations  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x})$  and  $\dot{\mathbf{x}} = \mathbf{g}(\mathbf{x})$ . We define the Lie bracket between these two vector fields as

$$[\mathbf{f}, \mathbf{g}] = \frac{d\mathbf{g}}{d\mathbf{x}} \cdot \mathbf{f} - \frac{d\mathbf{f}}{d\mathbf{x}} \cdot \mathbf{g}.$$

We can check that the set of vector fields equipped with the Lie bracket is a Lie algebra. The proof is not difficult but tedious and will not be asked in this exercise.

- 1) Consider the linear vector fields  $\mathbf{f}(\mathbf{x}) = \mathbf{A} \cdot \mathbf{x}$ ,  $\mathbf{g}(\mathbf{x}) = \mathbf{B} \cdot \mathbf{x}$ . Compute  $[\mathbf{f}, \mathbf{g}](\mathbf{x})$ .
- 2) Consider the system

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}) \cdot u_1 + \mathbf{g}(\mathbf{x}) \cdot u_2.$$

We propose to apply the following cyclic sequence:

$t \in [0, \delta]$	$t \in [\delta, 2\delta]$	$t \in [2\delta, 3\delta]$	$t \in [3\delta, 4\delta]$	$t \in [4\delta, 5\delta]$	...
$\mathbf{u} = (1, 0)$	$\mathbf{u} = (0, 1)$	$\mathbf{u} = (-1, 0)$	$\mathbf{u} = (0, -1)$	$\mathbf{u} = (1, 0)$	...

where  $\delta$  is an infinitesimal time period  $\delta$ . This periodic sequence will be denoted by

$$\{(1, 0), (0, 1), (-1, 0), (0, -1)\}$$

Show that

$$\mathbf{x}(t + 2\delta) = \mathbf{x}(t - 2\delta) + [\mathbf{f}, \mathbf{g}](\mathbf{x}(t)) \delta^2 + o(\delta^2).$$

3) Which periodic sequence should we take in order to follow the field  $\nu \cdot [\mathbf{f}, \mathbf{g}]$ . We will first consider the case  $\nu \geq 0$  and then the case  $\nu \leq 0$ .

4) Propose a controller such that the system becomes

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}) \cdot a_1 + \mathbf{g}(\mathbf{x}) \cdot a_2 + [\mathbf{f}, \mathbf{g}](\mathbf{x}) \cdot a_3$$

where  $\mathbf{a} = (a_1, a_2, a_3)$  is the new input vector.

5) Consider a Dubins car described by the following state equations:

$$\begin{cases} \dot{x} &= u_1 \cos \theta \\ \dot{y} &= u_1 \sin \theta \\ \dot{\theta} &= u_2 \end{cases}$$

where  $u_1$  is the speed of the cart,  $\theta$  its orientation and  $(x, y)$  the coordinates of its center. Using Lie Brackets, add a new input to the system, *i.e.*, a new direction of control.

6) Propose a simulation to check the good behavior of your controller. To do this, take  $\delta$  small enough to be consistent with the second order Taylor approximation but large with respect to the sampling time  $dt$ . For instance, we may take  $\delta = \sqrt{dt}$ . The initial state vector is taken as  $\mathbf{x}(0) = (0, 0, 1)$ .

7) Propose a controller so that the car goes up. Consider the same questions, where the car goes down, right and left, respectively.

EXERCISE 22.– *Follow the equator*

See the correction video at <https://youtu.be/8brHzqFHHsG>

We consider a mobile body moving around a planet, the Earth, with a radius  $\rho_E$  rotating along the vertical axis with a rotation rate  $\dot{\psi}_E$ .

1) Draw the Earth rotating with a static body at position  $\mathbf{p}$  with the following frames (see Figure 5.4):

1.  $\mathcal{R}_0$  is the Earth-centered inertial frame, fixed with respect to the stars
2.  $\mathcal{R}_1$  is the Earth frame and turning with the Earth with a rotational speed  $\dot{\psi}_E$
3.  $\mathcal{R}_2$  is the navigation frame, ego-centered and pointing toward the sky
4.  $\mathcal{R}_3$  is the ego-centered frame of the robot.

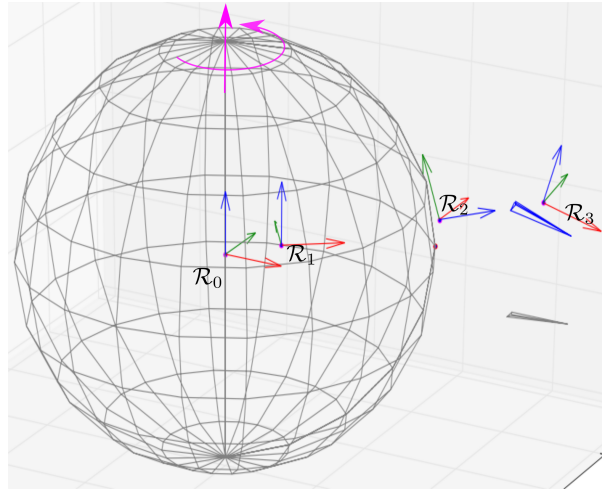


Figure 5.4: Earth with the different frames. The origin of the frames have been shifted for more visibility

For the simulation, take  $\dot{\psi}_E = 0.2 \text{ rad/s}$ ,  $\rho_E = 10\text{m}$ .

2) Provide the kinematic state equations for the body. The state variables are taken as  $(\mathbf{p}, \mathbf{R}_3, \mathbf{v}_3)$ , where  $\mathbf{p}$  is the position of the body in  $\mathcal{R}_0$ ,  $\mathbf{R}_3$  is the orientation of the body, and  $\mathbf{v}_3$  is the speed of the body seen from  $\mathcal{R}_0$  and expressed in  $\mathcal{R}_3$ . The inputs are chosen as  $\boldsymbol{\omega}_3$ , the rotation vector of the body and  $\mathbf{a}_3$ , the measured acceleration, both expressed in  $\mathcal{R}_3$ . Propose a simulation with the Earth rotating and the body moving in the space with a gravity given by  $\mathbf{g}(\mathbf{p}) = -9.81 \cdot \rho_T^2 \frac{\mathbf{p}}{\|\mathbf{p}\|^3}$ .

3) The planet is now assumed to be made of water and the body is an underwater robot equipped with actuators (such as propellers) to move inside the water. The gravity is taken as  $\mathbf{g}(\mathbf{p}) = -\frac{9.81}{\rho_T} \mathbf{p}$ . The water of the rotating planet drags the robot through friction forces (tangential and rotational). Moreover, the robot which has the same density as the surrounding fluid receives an Archimedes upward buoyant force. Propose a dynamical model describing the motion of the robot. The inputs will be the rotational acceleration  $\mathbf{u}_{\omega_3}$  et the tangential acceleration  $\mathbf{u}_{a_3}$  expressed in  $\mathcal{R}_3$ . These acceleration result from the action of the actuators creating forces and torques on the robot. Propose a simulation illustrating all these effects.

4) The robot behaves like a 3D Dubins-like car model. More precisely, it is equipped with a single propeller which push the robot forward without any possibility to control the speed. We take  $\mathbf{u}_{a_3} = (1, 0, 0)^T$ . We assume that the robot can chose its rotation vector using rudders, jet pumps or inertial wheels. Find a controller such that the robot goes to East along the equator. Validate by a simulation.

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EXERCISE 23.– *Modeling and control of a torpedo*

See the correction video at <https://youtu.be/4kLr7KnyNi8>

We consider the autonomous underwater vehicle (AUV), a *Riptide*, represented by Figure 5.5. For its locomotion, the robot uses three fins and one propeller. The inputs are  $u_0$ , the rotation speed

of the propeller and the angles  $u_1, u_2, u_3$  of the fins. We assume that we have no *side slip* effect. This means that the velocity vector  $\mathbf{v}_r$  has always the direction of the robot, or equivalently, the robot has no lateral speed.

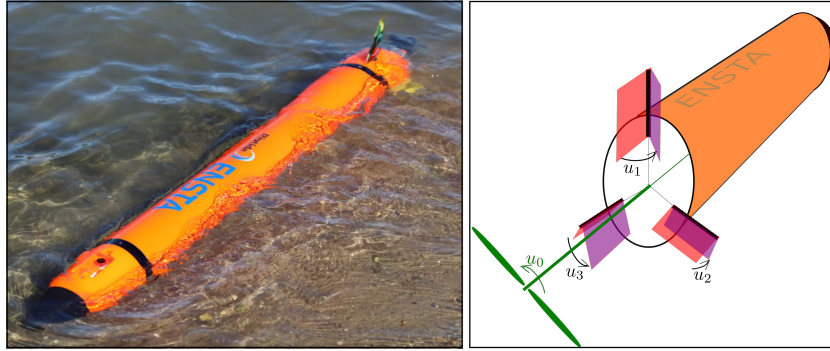


Figure 5.5: The Riptide robot has one propeller and three fins

1) Recall that the mechanization kinematic equations of a solid body are given by

$$\begin{cases} \dot{\mathbf{p}} &= \mathbf{R} \cdot \mathbf{v}_r \\ \dot{\mathbf{R}} &= \mathbf{R} \cdot (\boldsymbol{\omega}_r \wedge) \\ \dot{\mathbf{v}}_r &= \mathbf{a}_r - \boldsymbol{\omega}_r \wedge \mathbf{v}_r \end{cases}$$

where  $\mathbf{R}$  is the orientation matrix,  $\boldsymbol{\omega}_r$  is the rotation vector expressed in the robot frame,  $\mathbf{a}_r$  is the acceleration vector in the robot frame, and  $\mathbf{p}$  is the position of the robot in the world frame. Find a kinematic state space model for the robot, where the inputs are  $\boldsymbol{\omega}_r$  and the longitudinal acceleration  $a_{rx}$ .

2) Considering that the robot has no side slip, propose a dynamic model for the robot. We will assume that the robot has a cylinder shape, that its buoyancy is neutral, and the added mass is equal to the mass of the robot. The inputs are now,  $u_0, u_1, u_2, u_3$ .

3) Propose a controller for the dynamical model. More precisely, we want that the closed loop system has the desired rotation vector  $\bar{\boldsymbol{\omega}}_r$  and the desired acceleration  $\bar{a}_{rx}$ .

4) Illustrate the behavior of the system using a simulation. Try to get a trajectory similar to that illustrated by Figure 5.5.

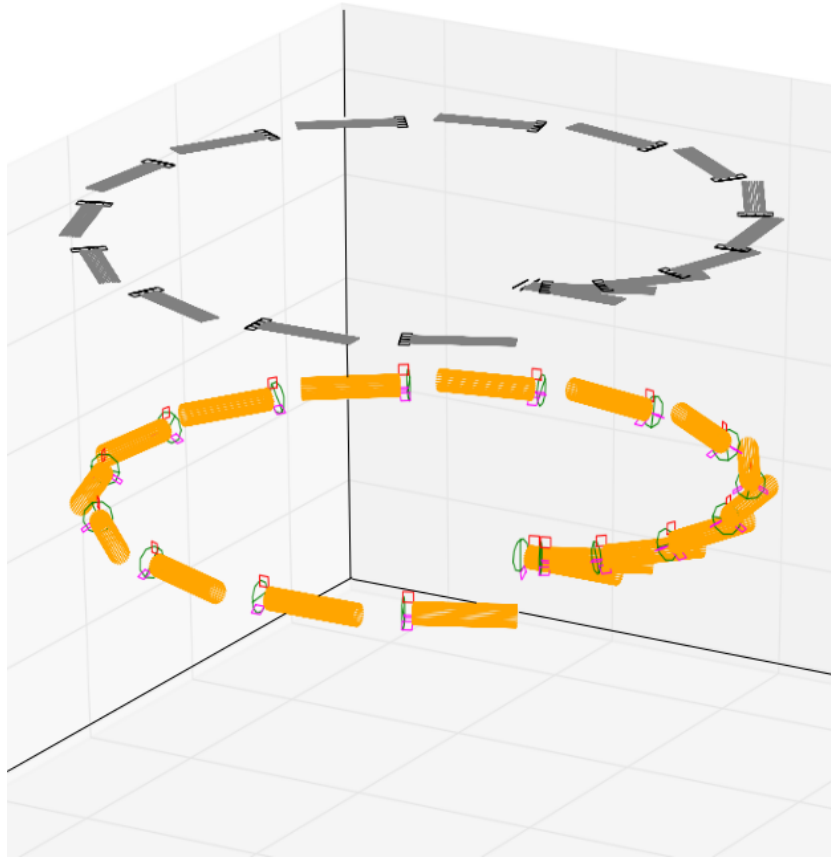


Figure 5.6: The Riptide (orange) is controlled to turn left with a desired speed of  $1\text{ms}^{-1}$ . The surface shadow at the surface is painted gray.

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EXERCISE 24.– *Geodesic*

See the correction video at <https://youtu.be/S1-j554Szac>

Consider a  $n$ -dimensional manifold embedded into  $\mathbb{R}^m$ ,  $m > n$  with the parametric equations

$$\mathbf{x} = \mathbf{r}(\mathbf{q}),$$

where  $\mathbf{q} \in \mathbb{R}^n$  represents the configuration vector (or the vector of the degrees of freedom) of a robot and  $\mathbf{x} \in \mathbb{R}^m$  is coordinate vector in the workspace. We assume that we can control the acceleration  $\ddot{\mathbf{q}}$  of  $\mathbf{q}$  via a control  $\mathbf{u}$ . In this exercise, we want to find a controller (see Figure 5.7) which computes  $\mathbf{u}$  such that  $\ddot{\mathbf{x}} - \mathbf{w}$  is as small as possible. We will then make the correspondence with the notion of geodesic used to characterize the shortest path between two points of a manifold [17].

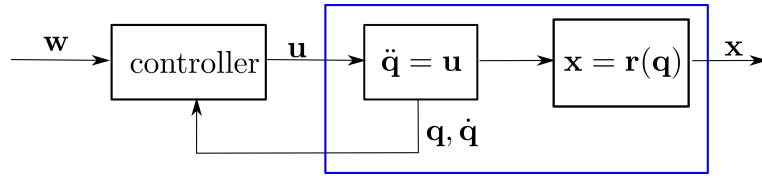


Figure 5.7: The controller generates  $\mathbf{u}$  in order to minimize  $\|\ddot{\mathbf{x}} - \mathbf{w}\|$

1) Consider a curve  $\mathbf{x}(t) = \mathbf{r}(\mathbf{q}(t))$ . Give an expression of  $\ddot{\mathbf{x}}$  as a function of  $\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}$ .

2) We want  $\ddot{\mathbf{x}} = \mathbf{w}$  where  $\mathbf{w}$  is the wanted acceleration. Now, due to the fact that  $\mathbf{x}(t)$  should stay on the manifold  $\mathbf{r}(\mathbb{R}^n)$ , all accelerations  $\ddot{\mathbf{x}}$  are not allowed. Only the projection of  $\ddot{\mathbf{x}}$  on the tangent plane can be chosen as desired. This is illustrated by Figure 5.8 where the car can turn left and right but should stay on the surface. This means that we cannot have any impact on the component of  $\ddot{\mathbf{x}}$  which is orthogonal to the manifold (red arrow in the figure).

Find the controller (see Figure 5.8 ) which computes  $\mathbf{u}$  such that  $\|\ddot{\mathbf{x}} - \mathbf{w}\|$  is as small as possible.

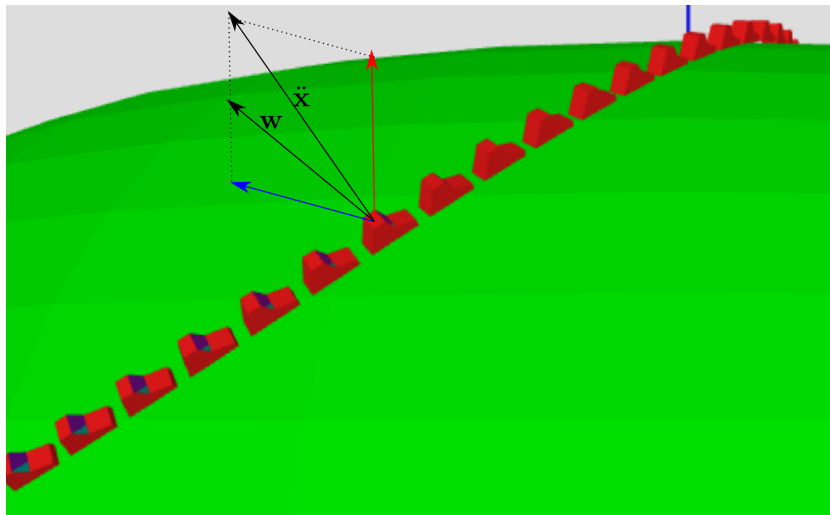


Figure 5.8: When moving on the manifold, we can only have an action the acceleration along the manifold (blue) but not on the transverse component (red)

3) When we do not want any acceleration along the manifold, we chose  $\mathbf{w} = \mathbf{0}$  and the corresponding trajectory is called a *geodesic*. What is the equation of a geodesic starting  $\mathbf{q}(0) = \mathbf{q}_0$  and  $\dot{\mathbf{q}}(0) = \dot{\mathbf{q}}_0$ .

4) We consider the torus, given by

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \mathbf{r}(q_1, q_2) = \begin{pmatrix} (a_1 + a_2 \cos q_2) \cos q_1 \\ (a_1 + a_2 \cos q_2) \sin q_1 \\ -a_2 \sin q_2 \end{pmatrix}$$

with  $a_1 = 10, a_2 = 6$ . Compute and draw a geodesic on the torus starting from  $\mathbf{q}(0) = (0, 0)$  and  $\dot{\mathbf{q}}(0) = (0.1, 0.2), t \in [0, 50]$ . An illustration of what you should obtain is given by Figure 5.9.

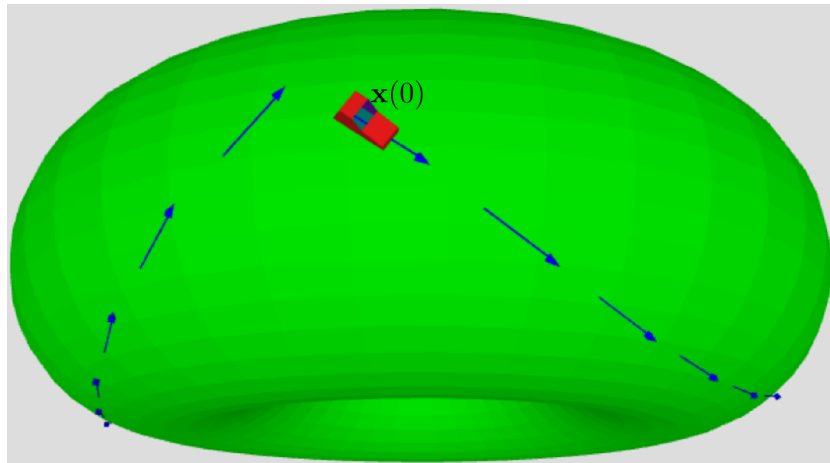


Figure 5.9: The geodesic oscillates around the external equator

5) We consider the ellipsoid, described by

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \mathbf{r}(q_1, q_2) = \begin{pmatrix} a_1 \cos q_2 \cos q_1 \\ a_2 \cos q_2 \sin q_1 \\ a_3 \sin q_2 \end{pmatrix}$$

where  $a_1 = 8, a_2 = 4, a_3 = 2$ . Write a program which computes a geodesic starting from  $\mathbf{q}(0) = (0, 0)$  and such that  $\mathbf{q}(10) = (2, 1)$ .

#### EXERCISE 25.– *Helicopter looping*

See the correction video at <https://youtu.be/aCI-Tgydct8>

Consider the helicopter robot described by Figure 5.5. It has a main rotor on top and a small tail rotor. In order to control the helicopter, one can tilt the angle of each blade on both rotors. The main rotor is used to move the helicopter up and down, and to make the helicopter tilt forward, backward, left or right. By tilting a blade to increase the blade's angle of attack, the pilot can increase the force of lift that is pushing up on that blade [18].

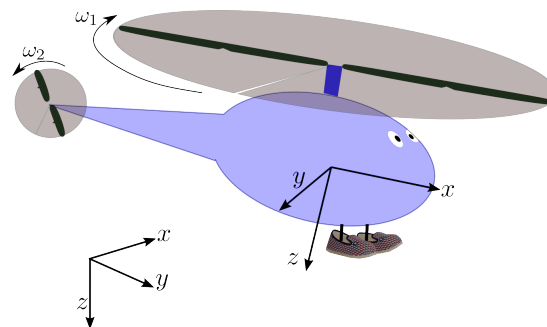


Figure 5.10: Helicopter

A state space model is given by

$$\left\{ \begin{array}{lcl} \dot{\mathbf{p}} & = & \mathbf{R} \cdot \mathbf{v}_r & (i) \\ \dot{\mathbf{R}} & = & \mathbf{R} \cdot (\boldsymbol{\omega}_r \wedge) & (ii) \\ \dot{\mathbf{v}}_r & = & \mathbf{R}^T \cdot \begin{pmatrix} 0 \\ 0 \\ g \end{pmatrix} + \frac{1}{m} \cdot \begin{pmatrix} 0 \\ 0 \\ -\tau_0 \end{pmatrix} - \boldsymbol{\omega}_r \wedge \mathbf{v}_r & (iii) \\ \dot{\boldsymbol{\omega}}_r & = & \mathbf{I}^{-1} \cdot \left( \begin{pmatrix} \tau_1 \\ \tau_2 \\ \tau_3 \end{pmatrix} - \boldsymbol{\omega}_r \wedge (\mathbf{I} \cdot \boldsymbol{\omega}_r) \right) & (iv) \end{array} \right.$$

where

$$\begin{pmatrix} \tau_0 \\ \tau_1 \\ \tau_2 \\ \tau_3 \end{pmatrix} = \begin{pmatrix} \beta_1 \omega_1^2 & 0 & 0 & 0 \\ 0 & \beta_2 \omega_1^2 & 0 & 0 \\ 0 & 0 & \beta_3 \omega_1^2 & 0 \\ -\delta_1 \omega_1^2 & 0 & 0 & -\beta_4 l \omega_2^2 \end{pmatrix} \cdot \begin{pmatrix} u_1 \\ u_2 \\ u_3 \\ u_4 \end{pmatrix}.$$

In this model,  $\mathbf{R}$  corresponds to the orientation,  $\mathbf{p}$  the position,  $\mathbf{v}_r$  the speed vector expressed in the robot frame,  $\boldsymbol{\omega}_r$  the rotation vector also expressed in the robot frame,  $\tau_0$  is the total thrust generated by the main rotor,  $\tau_1$  is the roll torque,  $\tau_2$  is the pitch torque,  $\tau_3$  is head torque generated by the tail rotor and  $\mathbf{u}$  the input vector. We assume that  $\omega_1 = \omega_2 = 100$ ,  $\beta_1 = 0.02$ ,  $\beta_2 = \beta_3 = \frac{\beta_1}{10}$ ,  $\beta_4 = 0.002$ ,  $\delta_1 = \frac{\beta_1}{5}$  all expressed in the international unit system.

1) Provide a controller to that the helicopter has the desired orientation  $\mathbf{R}^d$  and the desired total thrust  $\tau_0^d$ .

2) Give a controller for the system, so that the helicopter performs a looping, as illustrated by Figure 5.5.

3) Validate using a simulation.

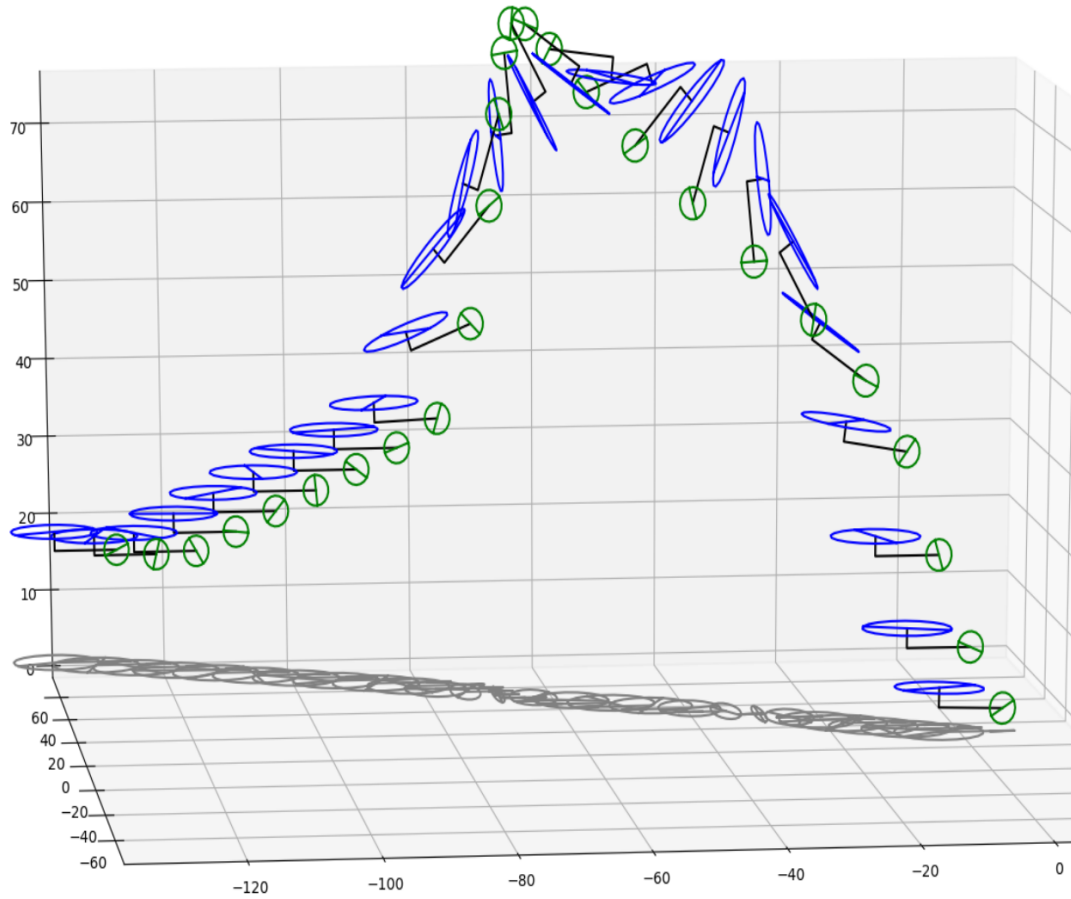


Figure 5.11: Helicopter performing a looping

EXERCISE 26.– *Hexarotor*

See the correction video at <https://youtu.be/3HVC9IMDB4o>

A hexarotor is composed of a body and 6 rotors. Each rotor is at position  $\mathbf{q}(i)$  and produces a force  $f_i$  in the direction  $\mathbf{d}(i)$ . In the body frame, the coordinates of  $\mathbf{q}(i)$ ,  $\mathbf{d}(i)$  are given by the following table:

$i$	1	2	3	4	5	6
$\mathbf{q}(i)$	$\begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$	$\begin{pmatrix} -1 \\ 0 \\ 0 \end{pmatrix}$	$\begin{pmatrix} 0 \\ -1 \\ 0 \end{pmatrix}$	$\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$	$\begin{pmatrix} \frac{1}{2} \\ 0 \\ 0 \end{pmatrix}$	$\begin{pmatrix} 0 \\ \frac{1}{2} \\ 0 \end{pmatrix}$
$\mathbf{d}(i)$	$\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$	$\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$	$\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$	$\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$	$\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$	$\begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$

This is illustrated by Figure 5.12. Each force  $f_i$  contributes to the resultant on the robot as  $\mathbf{d}(i) \cdot f_i$  and to the torque as  $\mathbf{q}(i) \wedge \mathbf{d}(i) \cdot f_i$ .

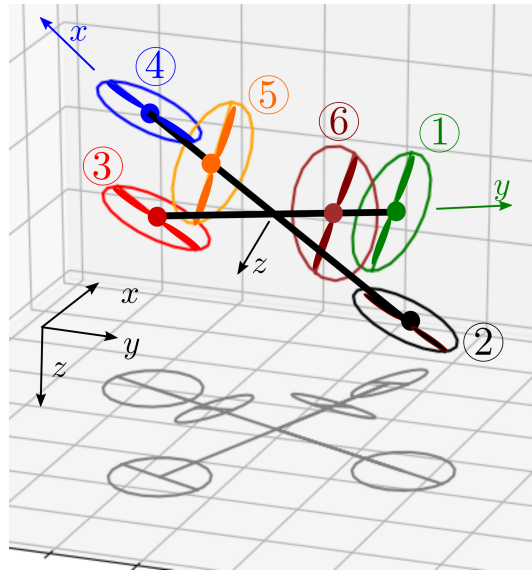


Figure 5.12: Hexarotor with its six propellers

1) Find the state equations of the hexarotor. The state is composed of the position  $\mathbf{p}$ , the orientation  $\mathbf{R}$ , the speed  $\mathbf{v}_r$  in the robot frame and the rotation vector  $\boldsymbol{\omega}_r$  in the robot frame. The inputs are  $f_1, \dots, f_6$ .

2) Find a controller that makes the hexarotor able to dock on a platform at position  $\mathbf{p}_d$  with an orientation  $\mathbf{R}_d$ .

3) Illustrate on a simulation in the case where the position and the orientation for the platform (which corresponds to the target to be reached) are given by

$$\mathbf{p}_d = \begin{pmatrix} \sin \frac{3t}{10} \\ \cos \frac{4t}{10} \\ \frac{1}{10} \cdot \sin \frac{3t}{10} \end{pmatrix}, \quad \mathbf{R}_d = \text{Exp} \begin{pmatrix} \sin t \\ \cos 2t \\ t \end{pmatrix}.$$

#### EXERCISE 27.— *Scan satellite*

See the correction video at <https://youtu.be/MRD3xtGYZwE>

We consider a 2-dimensional satellite in orbit around the Earth. We assume that it is described by the state equations

$$\begin{cases} \dot{\mathbf{p}} &= \mathbf{R} \cdot \mathbf{v}_r \\ \dot{\mathbf{R}} &= \mathbf{R} \cdot \boldsymbol{\omega} \wedge \\ \dot{\mathbf{v}}_r &= \mathbf{R}^T \cdot \mathbf{g} - \boldsymbol{\omega} \wedge \mathbf{v}_r \\ \dot{\boldsymbol{\omega}} &= \boldsymbol{\tau} \\ \dot{\mathbf{R}}_w &= \mathbf{R}_w \cdot \boldsymbol{\omega}_w \wedge \\ \dot{\boldsymbol{\omega}}_w &= -\boldsymbol{\tau} \end{cases}$$

where  $\mathbf{R}_w$  and  $\omega_w$  define the angle position and rotation speed of an inertial wheel inside the satellite,  $\mathbf{p}$  is the position of the satellite,  $\mathbf{R}$  is its orientation,  $\mathbf{v}_r$  its speed expressed in its own frame and  $\omega$  its rotation speed. We assume that the gravity is  $\mathbf{g} = -\frac{\mathbf{p}}{\|\mathbf{p}\|^3}$ .

1) Provide a simulation for  $\tau = 0$  with the following initial conditions:

$$\mathbf{p}(0) = \begin{pmatrix} 0 \\ 1 \end{pmatrix}, \mathbf{v}_r(0) = \begin{pmatrix} 0.8 \\ 0 \end{pmatrix}, \mathbf{R}(0) = \mathbf{R}_w(0) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \omega(0) = \omega_w(0) = 0.$$

2) Find a controller so that the satellite is always oriented toward the center of the Earth.



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