

Métodos Matemáticos de la Física

OSCAR REULA

ACKNOWLEDGEMENTS

First of all, I want to thank Gloria Puente for supporting me throughout the time I wrote these notes, which took quite a bit of our shared hours, generally long nights. Secondly, to Bernardo González Kriegel who converted the first versions of these notes to \LaTeX , doing a great job with much enthusiasm. Also to Bob Geroch, with whom I discussed several topics of the notes and from whom I was also inspired through his writings and books, not only in content but also in style. Finally, to several generations of students who stoically assimilated a large amount of the material from these courses in a very short time.

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PREFACE

These notes, now turned into a book, originated as an attempt to condense in one place a large set of ideas, concepts, and mathematical tools that I consider basic for the understanding and daily work of a physicist today.

It usually happens that if a problem is formulated from a physical need, such as the description of some natural phenomenon, then it is well formulated, in the sense that a reasonable solution to it exists. This rule has generally been very fruitful and has particularly served as a guide to many mathematicians to make their way in unknown areas. But it has also served, particularly to many physicists, to work without worrying too much about formal aspects, whether analytical, algebraic, or geometric, and thus be able to concentrate on physical and/or computational aspects. Although this allows for the rapid development of some research, in the long run, it leads to stagnation because by proceeding in this way, one avoids facing problems that are very rich in terms of conceptualizing the phenomenon to be described. It is important to verify that the formulated problem has a mathematically and physically correct solution.

An example of this was the development, in the middle of the last century, of the modern theory of partial differential equations. Many of these equations arose because they describe physical phenomena: heat transmission, electromagnetic wave propagation, quantum waves, gravitation, etc. One of the first mathematical responses to the development of these areas was the Cauchy-Kowalevski theorem, which tells us that given a partial differential equation (under quite general circumstances), if an analytic function is given as data on a hypersurface (with certain characteristics), then there is a unique solution in a sufficiently small neighborhood of that hypersurface. It took a long time to realize that this theorem was not really relevant from the point of view of physical applications: there were equations admitted by the theorem such that if the data was not analytic, there was no solution! And in many cases, if they existed, they did not depend continuously on the given data, a small variation of the data produced a completely different solution. It was only in the middle of the last century that substantial progress was made on the problem, finding that there were different types of equations, hyperbolic, elliptic, parabolic, etc., that behaved differently and this reflected the different physical processes they simulated. Due to its relative novelty, this very important set of concepts is not part of the set of tools that many active physicists have, nor are they found in the textbooks usually used in undergraduate courses.

Like the previous one, there are many examples, particularly the theory of or-

dinary differential equations and geometry, without which it is impossible to understand many of the modern theories, such as relativity, elementary particle theories, and many phenomena of solid-state physics. As our understanding of the basic phenomena of nature advances, we realize that the most important tool for their description is geometry. This tool, among other things, allows us to handle a wide range of processes and theories with little in common among each other with a very reduced set of concepts, thus achieving a synthesis. These syntheses are what allow us to acquire new knowledge, since by adopting them we leave space in our minds to learn new concepts, which are in turn ordered more efficiently within our mental construction of the area.

These notes were originally intended for a four-month course. But in reality, they were more suited for an annual course or two semesters. Then, as more topics were incorporated into them, it became increasingly clear that they should be given in two semesters or one annual course. Basically, one course should contain the first chapters that include notions of topology, vector spaces, linear algebra, ending with the theory of ordinary differential equations. The task is considerably simplified if the students have previously had a good course on linear algebra. The correlation with physics subjects should be such that the course is prior to or concurrent with an advanced mechanics course. Emphasizing in it the fact that ultimately one is solving ordinary differential equations with a certain special structure. Using the concepts of linear algebra to find eigenmodes and the stability of equilibrium points. And finally using geometry to describe, albeit briefly, the underlying symplectic structure.

The second course consists of developing the tools to be able to discuss aspects of the theory of partial differential equations. It should be given before or concurrently with an advanced electromagnetism course, where emphasis should be placed on the type of equations that are solved (elliptic, hyperbolic), and the meaning of their initial or boundary conditions, as appropriate. Also using coherently the concept of *distribution*, which is far from being an abstract mathematical concept but is actually a concept that naturally appears in physics.

None of the content of these notes is original material, but some ways of presenting it are: for example, some proofs that are simpler than usual, or the way some content is integrated with the previous ones. Much of the material should be thought of as a first reading or an introduction to the topic. The interested reader should consult the cited books, from which I have drawn upon a lot, being these excellent and difficult to surpass.

BASIC CONCEPTS OF TOPOLOGY

1.1 Introduction

The notion of a set, while telling us that certain objects—the elements that comprise it—have something in common with each other, does not give us any idea of the *closeness* between these elements. On the other hand, if we consider for example the real numbers, this notion is present. We know, for example, that the number 2 is much closer to 1 than 423 is. The concept of a topology in a set that we will define below tries to capture precisely this notion of closeness which, as we will see, admits many gradations.

Definition: A **topological space** consists of a pair (X, \mathcal{T}) , where X is a set and \mathcal{T} is a collection of subsets of X satisfying the following conditions:

1. The subsets \emptyset and X of X are in \mathcal{T} .
2. If $O_\lambda, \lambda \in I$, is a one-parameter family of subsets of X in \mathcal{T} , then $\bigcup_I O_\lambda$ is also in \mathcal{T} .
3. If O and O' are in \mathcal{T} , then $O \cap O'$ is also in \mathcal{T} .

For simplicity, we will often denote the topological space (X, \mathcal{T}) simply by X when the topology is understood from context, as is common practice.

The elements of \mathcal{T} , subsets of X , are called the **open subsets** of X . The set \mathcal{T} itself is called a **topology** on X . Condition 2) tells us that infinite unions of elements of \mathcal{T} are also in \mathcal{T} , while condition 3) tells us that in general only finite intersections remain in \mathcal{T} . The following examples illustrate why this asymmetry exists, they also illustrate how giving a topology is essentially giving a notion of closeness between the points of the set in question.

Example: a) Let $\mathcal{T} = \{\emptyset, X\}$, that is, the only open subsets of X are the empty subset and the subset X . It is clear that this collection of subsets is a topology, as it satisfies the three required conditions. This topology is called the **indiscrete topology** on X . We can say that, with this topology, the points of X are arbitrarily close to each other, since if an open set contains one of them it contains all of them.

Example: b) Let $\mathcal{T} = \mathcal{P}(X)$, the collection of all subsets of X . Clearly, this collection also satisfies the conditions mentioned above and therefore it is also a topology on X , the so-called **discrete topology** on X . We can say that in this one all the points are arbitrarily separated from each other since, for example, given any point of X there is an open set that separates it from all the others, which consists of only the point in question.

Example: c) Let X be the set of real numbers, \mathbb{R} , and let $\mathcal{T} = \{O \subseteq \mathbb{R} \mid \text{if } r \in O, \exists \varepsilon > 0 \text{ such that if } |r - r'| < \varepsilon, r' \in O\}$, that is, the collection of open sets in the usual sense of the real numbers. Let's see that this collection satisfies the conditions to be a topology. Clearly, $\emptyset \in \mathcal{T}$ (since it has no r), as well as $\mathbb{R} \in \mathcal{T}$ (since it contains all r'), and thus condition 1) is satisfied. Let us examine the second condition: let $r \in \bigcup_I O_\lambda$ then $r \in O_\lambda$ for some λ and therefore there will exist $\varepsilon > 0$ such that all r' with $|r - r'| < \varepsilon$ are also in O_λ , and therefore in $\bigcup_I O_\lambda$. Finally, the third condition: let $r \in O \cap O'$ then r is in O and therefore there will exist $\varepsilon > 0$ such that all r' with $|r - r'| < \varepsilon$ will be in O ; as r is also in O' , there will exist $\varepsilon' > 0$ such that all r' with $|r - r'| < \varepsilon'$ will be in O' . Let $\varepsilon'' = \min\{\varepsilon, \varepsilon'\}$, then all r' with $|r - r'| < \varepsilon''$ will be in O and in O' and therefore in $O \cap O'$, so we conclude that this last set is also in \mathcal{T} . \mathbb{R} with this topology is called the **real line**.

Exercise: In the real line, as defined in the previous example, find an infinite intersection of open sets that is not open.

Example: d) Let $X = \mathbb{R} \times \mathbb{R} \equiv \mathbb{R}^2$, that is, the Cartesian product of \mathbb{R} with itself —the set of all pairs (x, y) , with $x, y \in \mathbb{R}$ — and let $\mathcal{T} = \{O \in \mathbb{R}^2 \mid \text{if } (x, y) \in O, \exists \varepsilon > 0 \text{ such that if } |x - x'| + |y - y'| < \varepsilon, (x', y') \in O\}$. Following the previous example, it can be seen that this is also a topological space and that this is the topology we usually use in \mathbb{R}^2 .

Definition: A **metric space** is a pair (X, d) consisting of a set X and a map $d : X \times X \rightarrow \mathbb{R}$, usually called **distance**, satisfying the following conditions for all $x, x', x'' \in X$:

1. Non-negativity: $d(x, x') \geq 0$, with equality only when $x = x'$.
2. Symmetry: $d(x, x') = d(x', x)$.
3. Triangle inequality: $d(x, x') + d(x', x'') \geq d(x, x'')$.

Exercise: Prove that any metric space has a topology *induced* by its metric in a similar way to \mathbb{R} in the previous example.

Exercise: Prove that, for any set, $d(x, y) = 1$ if $x \neq y$ and $d(x, x) = 0$ defines a distance. What topology does this distance induce on the set?

Clearly, a distance gives us a notion of closeness between points, in the precise form of a numerical value. A topology, by not generally giving us any number, gives us a much vaguer notion of closeness, but still generally interesting.

1.1.1 Terminology

We now give a summary of the usual terminology in this area, which is a direct generalization of the commonly used one.

Definition: We will call the **complement**, O^c , of the subset O of X the subset of all elements of X that are not in O .

Definition: We will say that a subset O of X is **closed** if its complement O^c is open.

Definition: A subset N of X is called a **neighborhood** of $x \in X$ if there is an open set O_x , with $x \in O_x$, contained in N .

Definition: We will call the **interior** of $A \in X$ the subset $Int(A)$ of X formed by the union of all open sets contained in A .

Definition: We will call the **closure** of $A \in X$ the subset $Cl(A)$ of X formed by the intersection of all closed sets containing A .

Definition: We will call the **boundary** of $A \in X$ the subset ∂A of X formed by $Cl(A) - Int(A) \equiv Int(A)^c \cap Cl(A)$.

Exercise: Let (X, d) be a metric vector space, prove that: a) $C_x^1 = \{x' | d(x, x') \leq 1\}$ is closed and is a neighborhood of x . b) $N_x^\varepsilon = \{x' | d(x, x') < \varepsilon\}$, $\varepsilon > 0$ is also a neighborhood of x . c) $Int(N_x^\varepsilon) = N_x^\varepsilon$ d) $Cl(N_x^\varepsilon) = \{x' | d(x, x') \leq \varepsilon\}$ e) $\partial N_x^\varepsilon = \{x' | d(x, x') = \varepsilon\}$.

Exercise: Let (X, \mathcal{T}) be a topological space and A a subset of X . Prove that: a) A is open if and only if each $x \in A$ has a neighborhood contained in A . b) A is closed if and only if each x in A^c (that is, not belonging to A) has a neighborhood that does not intersect A .

Exercise: Let (X, \mathcal{T}) be a topological space, let $A \in X$ and $x \in X$. Prove that: a) $x \in Int(A)$ if and only if x has a neighborhood contained in A . b) $x \in Cl(A)$ if and only if every neighborhood of x intersects A . c) $x \in \partial A$ if and only if every neighborhood of x contains points in A and points in A^c .

1.2 Derived Concepts

In the previous sections, we have seen that the concept of a topology leads us to a generalization of a series of ideas and derived concepts that we knew how to handle in

\mathbb{R}^n , revealing that they did not depend on the usual distance used in these spaces (the so-called Euclidean distance). It is then worth asking if there are still other possible generalizations. In this and the next subsection, we will study two more of them. These in turn open up a vast area of mathematics, which we will not cover in this course but is very important in what concerns modern physics.

The first of these notions is *continuity*.

1.2.1 Continuous Maps

Definition: Let $\varphi : X \rightarrow Y$ be a map between two topological spaces (see the box below). We will say that the map φ is **continuous** if given any open set O of Y , $\varphi^{-1}(O)$ is an open set of X .

Definition: A **map** $\phi : X \rightarrow Y$ between a set X and another Y is an assignment to *each* element of X of an element of Y .

This generalizes the usual concept of a function. Note that the map is defined for every element of X while, in general, its **image**, that is, the set $\phi(X) \equiv \{y \in Y \mid \exists x \in X \text{ such that } \phi(x) = y\}$, is not all of Y . In the case that it is, i.e. $\phi(X) = Y$, we will say that the map is **surjective**. On the other hand, if it is fulfilled that $\phi(x) = \phi(\tilde{x}) \implies x = \tilde{x}$, we will say that the map is **injective**. In such a case, there exists the inverse map to ϕ between the set $\phi(X) \subset Y$ and X . If the map is also surjective then its inverse is defined on all of Y and in this case, it is denoted by $\phi^{-1} : Y \rightarrow X$. It is also useful to consider the sets $\phi^{-1}(O) = \{x \in X \mid \phi(x) \in O\}$, sometimes called the **pre-image** of the set O under ϕ .

Clearly, the previous definition only uses topological concepts. Does it have anything to do with the usual *epsilon-delta* definition used in \mathbb{R}^n ? The answer is affirmative, as we will see below in our first theorem. But first, let's see some examples.

Example: a) Let X and Y be any topological spaces and let the topology on X be the discrete one. Then any map between X and Y is continuous. Indeed, for any open set O in Y , $\varphi^{-1}(O)$ is some subset in X , but in the discrete topology, every subset of X is an open set.

Example: b) Let X and Y be any topological spaces and let the topology on Y be the indiscrete one. Then any map between X and Y is also continuous. Indeed, the only open sets in Y are \emptyset and Y , but $\varphi^{-1}(\emptyset) = \emptyset$, while $\varphi^{-1}(Y) = X$, but whatever the topology of X , \emptyset and X are open sets.

From the previous examples, it might seem that our definition of continuity is not very interesting. But that is because we have taken cases with the *extreme* topologies. It is in the intermediate topologies where the definition becomes more useful.

Example: c) Let X and Y be real lines, and let $\varphi : X \rightarrow Y$ be a map such that $\varphi(x) = 1$

if $x \geq 0$, $\varphi(x) = -1$ if $x < 0$. This map is not continuous because, for example, while the interval $I = (1/2, 3/2) \subseteq Y$ is open, its pre-image $\varphi^{-1}(I) = \{x \mid x \geq 0\}$ is not open.

Theorem 1.1 *The map $\varphi : X \rightarrow Y$ is continuous if and only if given any point $x \in X$ and any neighborhood $M \subseteq Y$ of $\varphi(x)$, there exists a neighborhood $N \subseteq X$ of x such that $\varphi(N) \subset M$.*

This theorem provides an equivalent definition of continuity that is much closer to the intuitive concept of continuity.

Proof:

Suppose φ is continuous. Let x be a point of X , and $M \subseteq Y$ a neighborhood of $\varphi(x)$. By definition of neighborhood, there exists an open set $O \subset M$ in Y and containing $\varphi(x)$. By continuity, $N = \varphi^{-1}(O)$ is an open set of X , and as it contains x , it is a neighborhood of x . It is then fulfilled that $\varphi(N) \subset O \subset M$. Now to prove the reciprocal, suppose that given any point $x \in X$ and any neighborhood M of $\varphi(x)$, there exists a neighborhood N of x such that $\varphi(N) \subset M$. Then let O be any open set of Y , we must now show that $\varphi^{-1}(O)$ is an open set of X . Let x be any point of $\varphi^{-1}(O)$, then $\varphi(x) \in O$ and therefore O is a neighborhood of $\varphi(x)$, therefore there exists a neighborhood N of x such that $\varphi(N) \subset O$ and therefore $N \subset \varphi^{-1}(O)$. But then $\varphi^{-1}(O)$ contains a neighborhood of each of its points and therefore it is open.



Exercise: Let $\phi : X \rightarrow Y$ and $\psi : Y \rightarrow Z$ be continuous maps, prove that the composite map $\psi \circ \phi : X \rightarrow Z$ is also continuous. (Composition of maps preserves continuity.)

Induced Topology:

Let ϕ be a map between a set X and a topological space $\{Y, \mathcal{T}\}$. This map naturally provides, that is, without the help of any other structure, a topology on X , denoted by \mathcal{T}_ϕ and called the **topology induced** by ϕ on X . It is given by $\mathcal{T}_\phi = \{O \subset X \mid O = \phi^{-1}(Q), Q \in \mathcal{T}\}$, that is, O is an open set of X if there exists an open set Q of Y such that $O = \phi^{-1}(Q)$. In other words, the open sets in X are the pre-images of all the open sets in Y . Note that under this topology on X , the map ϕ is automatically continuous. In fact, it is by construction the “minimal” topology on X that renders ϕ continuous.

Exercise: Prove that this construction really defines a topology.

Not all topologies thus induced are of interest and, in general, they depend strongly on the map, as shown by the following example:

Example:

a) Let $X = Y = \mathbb{R}$ with the usual topology and let $\phi : \mathbb{R} \rightarrow \mathbb{R}$ be the function $\phi(x) = 17 \forall x \in \mathbb{R}$. This function is clearly continuous with respect to the topologies of X and Y , those of the real line. However, the topology induced on X by this map is the indiscrete one! $\mathcal{T}_\phi = \{\emptyset, X\}$. Interestingly, this “loss” of topological information can be intuitively associated to the information lost by ϕ in mapping the whole \mathbb{R} to a single point, 17. In fact, ϕ would be continuous with respect to *any* topology on X , so its continuity is not very interesting, as it does not tell us anything about the underlying topology.

b) Let X and Y be as in a) and let $\phi(x)$ be an invertible map, then \mathcal{T}_ϕ coincides with the topology of the real line.

1.2.2 Compactness

The second generalization that we are interested in corresponds to the concept of *compactness*. For this, we introduce the following definition: let X be a set, A a subset of X , and $C = \{A_\lambda\}$ a collection of subsets of X parameterized by a continuous or discrete variable λ . We say that this collection **covers** A if $A \subset \bigcup_\lambda A_\lambda$. In such a case, we also say that C is a **cover** of A . When A is a topological space and the sets $\{A_\lambda\}$ are open, we say that C is an **open cover** of A . Finally, a subcollection of C that also covers A is called a **subcover**.

Definition: We say that a subset A of a topological space is **compact** if given any collection $\{A_\lambda\}$ of *open sets* that covers A , there exists a *finite* subcollection of these A_λ that also covers A .

Example: a) Let X be an infinite set of points with the discrete topology. Then a cover of X consists, for example, of all subsets of the form $\{x\}$, with $x \in X$. Since the topology of X is discrete, any cover is an open cover. But no finite number of the sets $\{x\}$ cover all of X , therefore X is not compact in this case. Clearly, if X had only a finite number of elements it would always be compact, regardless of its topology.

Example: b) Let X be any set with the indiscrete topology. Then X is compact. The only open sets of this set are \emptyset and X , so any open cover has X as one of its members and this alone is enough to cover X .

Thus, we see that this property strongly depends on the topology of the set. The relation with the intuitive concept of compactness is clear from the following example and exercise.

Example: c) Let X be the real line and $A = (0, 1)$. This subset is not compact because, for example, the following is an open cover of A that has no finite subcover: $\{A_n = (\frac{1}{n}, \frac{n-1}{n}) : n \in \mathbb{N}\}$

Exercise: Let X be the real line and $A = [0, 1]$. Prove that A is compact.

Proof:

Let $\{A_\lambda\}$ be an open cover of $[0, 1]$. Consider the set S of all $x \in [0, 1]$ such that $[0, x]$ is covered by *finitely* many A_λ . We shall see that $S = [0, 1]$. This set is not empty since 0 is in some open set of the covering, and therefore there exists $x > 0$ such that $[0, x]$ is inside this open set. Furthermore it is bounded above by 1. Therefore, S is a non-empty upper-bounded real set, so by the least-upper-bound property of \mathbb{R} it has a least upper bound. Let a be that number and A_{λ_0} be an element of the cover such that $a \in A_{\lambda_0}$. We assume $a < 1$ otherwise we would be done. Adding this set to the finite subcovering we still get a finite subcover containing a , and therefore also some other element $1 > b > a$. Contradicting the fact that a was a least upper bound.



Now let's see the relation between the two derived concepts of Topology, namely the continuity of maps between topological spaces and compactness. The fact that a map between topological spaces is continuous implies that this map is special, in the sense that *it carries or conveys information about the respective topologies and preserves the topological properties of the sets it associates*. This is seen in the following property, which—as shown by the following example—is very important.

Theorem 1.2 *Let X and Y be two topological spaces and $\phi : X \rightarrow Y$ a continuous map between them. Then if A is a compact subset of X , $\phi(A)$ is a compact subset of Y .*

Proof: Let $\{O_\lambda\}$ be a collection of open sets in Y that cover $\phi(A)$. Then the collection $\{\phi^{-1}(O_\lambda)\}$ covers A , but A is compact and therefore there will be a finite subcollection $\{\phi^{-1}(O_{\hat{\lambda}})\}$ of the former that also covers it. Therefore, the finite subcollection $O_{\hat{\lambda}}$ will also cover $\phi(A)$. Since this is true for any open cover of $\phi(A)$ we conclude that $\phi(A)$ is compact.

Example: Let A be a compact topological space and let $\phi : A \rightarrow \mathbb{R}$ be a continuous map between A and the real line. $\phi(A)$ is then a compact real set and therefore a closed and bounded set, but then this set will have a maximum and a minimum, that is, the map ϕ reaches its maximum and minimum in A .

Finally, another theorem of fundamental importance about compact sets, which shows that they have another property that makes them very interesting. For this, we introduce the following definitions, which also only use topological concepts. A **sequence** in a set X , $\{x_n\} = \{x_1, x_2, \dots\}$, with $x_n \in X$, is a map from \mathbb{N} to this set. Given a sequence $\{x_n\}$ in a topological space X , we say that $x \in X$ is a **limit point** of this sequence if given any open set $O \subseteq X$ containing x there exists a number N such that for all $n > N$, $x_n \in O$. We say that $x \in X$ is an **accumulation point** of this sequence if given any open set $O \subseteq X$ containing x , infinitely many elements of the sequence also belong to O .

Exercise: Find an example of a sequence in some topological space with different limit points.

Theorem 1.3 *Let A be compact. Then every sequence in A has an accumulation point.*

Proof: Suppose —contrary to the theorem’s assertion— that there exists a sequence $\{x_n\}$ in A without any accumulation points. That is, given any point $x \in A$ there exist a neighborhood O_x containing it and a number N_x such that if $n > N_x$ then $x_n \notin O_x$. Since this is valid for any x in A , the collection of sets $\{O_x | x \in A\}$ is an open cover of A , but A is compact and therefore it has a finite subcover. Let N be the maximum among the N_x of this finite subcover. But then $x_n \notin A$ for all $n > N$ which is absurd. ♠

Exercise: Prove that compact sets in the real line are the closed and bounded ones.

We can now ask the inverse question: if $A \subset X$ is such that every sequence has accumulation points, is it true then that A is compact? An affirmative answer would give us an alternative characterization of compactness, and this is affirmative for the case of the real line. In general, however, the answer is negative: there are topologies in which every sequence in a set has accumulation points in it, but this set is not compact. However, all the topologies we will see are **second countable** [See box] and in these the answer is affirmative.

In the real line, it is true that if $x \in \mathbb{R}$ is an accumulation point of a sequence $\{x_n\}$ then there exists a **subsequence**, $\{\tilde{x}_n\}$ (that is, a restriction of the map defining the sequence to an infinite subset of \mathbb{N}) having x as a limit point. This is also not true in the generality of topological spaces, but it is if we consider only those that are **first countable** [See box]. All the spaces we will see in this course are first countable.

***Countability of topological spaces.**

Definition: We say that a topological space $\{X, \mathcal{T}\}$ is first countable if for each $p \in X$ there exists a countable collection of open sets $\{O_n\}$ such that for every neighborhood N of p there exists i such that $O_i \subset N$.

Definition: We say that a topological space $\{X, \mathcal{T}\}$ is second countable if there is a countable collection of open sets such that any open set of X can be expressed as a union of sets from this collection.

Example:

- a) X with the indiscrete topology is first countable.
- b) X with the discrete topology is first countable. And second countable if and only if its elements are countable.

Exercise: Prove that the real line is first and second countable. Hint: For the first case, use the open sets $O_n = (p - \frac{1}{n}, p + \frac{1}{n})$ and for the second $O_{pqn} = (\frac{p}{q} - \frac{1}{n}, \frac{p}{q} + \frac{1}{n})$

***Separability of topological spaces.**

Definition: A topological space X is Hausdorff if given any pair of points of X , x and y , there exist neighborhoods O_x and O_y such that $O_x \cap O_y = \emptyset$.

Example:

- a) X with the indiscrete topology is not Hausdorff.
- b) X with the discrete topology is Hausdorff.

Exercise: Find a topology such that the integers are Hausdorff and compact.

Exercise: Prove that if a space is Hausdorff then if a sequence has a limit point, it is unique.

Exercise: Let X be compact, Y Hausdorff, and $\phi : X \rightarrow Y$ continuous. Prove that the images of closed sets are closed. Find a counterexample if Y is not Hausdorff.

Bibliography notes: This chapter is essentially a condensed version of Chapters 26, 27, 28, 29, and 30 of [1], see also [2], [6], and [21]. Topology is one of the most fascinating branches of mathematics, if you delve deeper you will be captivated! Of particular interest in physics is the notion of *Homotopy*, a good place to understand these ideas is Chapter 34 of [1].

2.1 Vector Spaces

Definition: A **Real Vector Space** consists of three things: – *i*) A set, V , whose elements will be called **vectors**. *ii*) A rule that assigns to each pair of vectors, v , u , a third vector that we will denote by $v + u$ and that we will call their **sum**. *iii*) A rule that assigns to each vector, v and to each real number a , a vector that we will denote by av and that we will call the **product** of v with a . – subject to the following conditions:

- 1.a) For any pair of vectors $u, v \in V$ it holds that,

$$u + v = v + u \quad (2.1)$$

- 1.b) There exists in V a unique element called **zero** and denoted by o , such that

$$o + v = v \quad \forall v \in V. \quad (2.2)$$

- 1.c) For any vector $u \in V$ there exists a unique vector in V , denoted $-u$, such that,

$$u + (-u) = o \quad (2.3)$$

- 2.a) For any pair of real numbers a and a' and any vector v it holds that,

$$a(a'v) = (aa')v.$$

- 2.b) For any vector v it holds that,

$$1v = v.$$

- 3.a) For any pair of real numbers a and a' and any vector v it holds that,

$$(a + a')v = av + a'v.$$

3.b) For any real number a and any pair of vectors v and v' it holds that,

$$a(v + v') = av + av'.$$

The first conditions involve only the rule of addition; these are actually the rules that define a group, a structure we will revisit in a later chapter, where addition represents the product between elements of the group. The next conditions involve only the rule of multiplication, while the last two deal with the relationship between these two operations. As we will see with examples later, real numbers can be replaced by any field, such as rationals, \mathbb{Q} , integers, \mathbb{Z} , complex numbers, \mathbb{C} , and even finite fields.

Example: The set of all n -tuples of real numbers with the usual operations of addition and multiplication *tuple by tuple*. This space is denoted by \mathbb{R}^n .

Example: Let S be any set and let V be the set of all functions from S to the reals, $v : S \rightarrow \mathbb{R}$, with the following operations of addition and multiplication: The sum of the function v with the function v' is the function (element of V) that assigns to the element s of S the value $v(s) + v'(s)$. The product of $a \in \mathbb{R}$ with the function v is the function that assigns to $s \in S$ the value $av(s)$. This example will appear very often in the following chapters.

Definition: We will say that a set of vectors $\{e_i\}$ $i = 1, \dots, n$ are **linearly independent** if $\sum_i a^i e_i = 0$ $a^i \in \mathbb{R}$, $i = 1, \dots, n \implies a^i = 0$, $i = 1, \dots, n$, that is, if any non-trivial linear combination of these vectors gives us a non-trivial vector.

Definition: The *Span* of a set of vectors $\{u_i\}$ is the set of all possible linear combinations of these elements. They generate a subspace of V , which we will denote by $\text{Span}\{u_i\} \subset V$.

If a finite number of linearly independent vectors, n , are sufficient to **span** V , [that is, if any vector in V can be obtained as a linear combination of these n vectors], then we will say that these form a **basis** of V and that the **dimension** of V is n , $\dim V = n$.

Exercise: Show that given a vector v and a basis, $\{e_i\}$ $i = 1, \dots, n$, there exists a unique linear combination of elements of the basis that determines it. That is, if $v = \sum_i v^i e_i$ and $v = \sum_i \tilde{v}^i e_i$, then $v^i = \tilde{v}^i$ for all $i = 1, \dots, n$.

Exercise: If we have two bases, $\{e_i\}$ and $\{\tilde{e}_i\}$, we can write the elements of one in terms of the other,

$$\tilde{e}_j = \sum_i P_j^i e_i, \quad e_i = \sum_l R_i^l \tilde{e}_l.$$

See that $R_i^l = P_i^{-1l}$.

Exercise: Let S be a finite set, $S = \{s_1, s_2, \dots, s_n\}$, find a basis of the vector space of all functions from S to \mathbb{R} . Find the dimension of this space.

Exercise: Show that the dimension of V is unique, that is, it does not depend on the basis used to define it.

Note that if in the previous example S consists of a finite number of elements, then the dimension of V is finite.¹ In the case that S had an infinite number of elements, we would say that the dimension of V would be infinite. In what follows in this chapter, we will only consider finite-dimensional vector spaces.

Let V be a vector space of dimension n and a basis of it, $\{e_i\} \ i = 1, \dots, n$. Given an n -tuple of real numbers, $\{c^i\}$ we then have determined an element of V , that is, the vector, $v = \sum_{i=1}^n c^i e_i$. On the other hand, we have seen in a previous exercise that given any vector, it determines a unique n -tuple of real numbers, the elements of v in that basis. We see that we then have an invertible map between V and the space of n -tuples, \mathbb{R}^n . This map is linear, assigning to the sum of two vectors v and \tilde{v} the n -tuple sum of the respective n -tuples. This map depends on the basis, but it still tells us that finite-dimensional vector spaces do not hold many surprises; they are always copies of \mathbb{R}^n .

2.1.1 Covectors and Tensors

Let V be a vector space of dimension n . Associated with this vector space, consider the set, $V^* = \{ \text{the space of linear maps } \omega : V \rightarrow \mathbb{R} \}$. This is also a vector space, called the **dual space to V** , or **space of covectors**, with addition and multiplication given by:

$$(\omega + \alpha\tau)(v) = \omega(v) + \alpha\tau(v) \quad \forall v \in V$$

with $\omega, \tau \in V^*$, $\alpha \in \mathbb{R}$.

What is its dimension? Note that if $\{e_i\} \ i = 1, \dots, n$ is a basis of V , that is, a set of linearly independent vectors that span V , we can define n elements of V^* (called covectors) by the relation

$$\theta^i(e_j) = \delta_j^i. \quad (2.4)$$

That is, we define the action of θ^i on the $\{e_j\}$ as in the equation above and then extend its action to any element of V by writing this element in the basis $\{e_i\}$ and using the fact that the action must be linear.

It can be easily seen that any $\rho \in V^*$ can be obtained as a linear combination of the covectors $\{\theta^j\}$, $j = 1, \dots, n$ and that these are linearly independent, therefore they form a basis and thus the dimension of V^* is also n .

Exercise: See that V^* is a vector space and that the $\{\theta^i\}$ really form a basis.

¹That is, a finite number of linearly independent vectors span V .

Exercise: See that if $v = \sum_{i=1}^n v^i e_i$ then,

$$v^i = \theta(v).$$

Exercise: Let V be the space of functions from a set with a finite number of elements, n , to the reals. Let a basis be given by:

$$e_i(a) := \begin{cases} 1 & \text{if } a \text{ is the } i\text{-th element} \\ 0 & \text{otherwise} \end{cases}$$

Find the corresponding co-basis of its dual space.

Since V and V^* have the same dimension, they are, as vector spaces, the same thing, but since there is no map that identifies them, we have to consider them as different.

What happens if we now take the dual of V^* ? Will we get more copies of V ? The answer is no, since there is a natural identification between V and its double dual V^{**} .

Indeed, to each $v \in V$ we can associate an element X_v of V^{**} , that is, a linear functional from V^* to \mathbb{R} , in the following way: $X_v(\omega) := \omega(v) \quad \forall \omega \in V^*$. That is, the element X_v of V^{**} associated with $v \in V$ is the one that, when acting on any covector ω , gives the number $\omega(v)$. Note that X_v acts linearly on the elements of V^* and therefore is an element of V^{**} . Are there elements of V^{**} that do not come from some vector in V ? The answer is no, since the map $X_v : V \rightarrow V^{**}$ is injective [$X_v(\omega) = 0 \quad \forall \omega \implies v = 0$] and therefore ² $\dim X_V = \dim V$. On the other hand $\dim V^{**} = \dim V^*$ since V^{**} is the dual of V^* and thus $\dim V = \dim V^* = \dim V^{**}$, which indicates that the map in question is also surjective and therefore invertible. This allows us to identify V and V^{**} and conclude that by dualizing we will not be able to construct more interesting vector spaces. In the case where the dimension of the vector space is not finite, this is no longer true and there are cases – frequently used – where $X_V \subset V^{**}$ strictly.

Exercise: Given a basis in V , $\{e_i\}$ and the corresponding co-basis, $\{\theta^j\}$. Define the co-co-basis in V^{**} , $\{E_i\}$. Find the relationship between the components of a vector of the form X_v in the basis $\{E_i\}$ and those of the vector v in the basis $\{e_i\}$.

Exercise: See that indeed $\dim X_V = \dim V$.

However, nothing prevents us from also considering **multilinear maps** ³ from $\underbrace{V \times V \times \dots \times V}_{k \text{ times}}$ to \mathbb{R} , or more generally,

$$\underbrace{V \times \dots \times V}_{k \text{ times}} \times \underbrace{V^* \times \dots \times V^*}_{l \text{ times}} \rightarrow \mathbb{R}.$$

²Denoting by X_V the image by $X_{(\cdot)}$ of V .

³That is, maps that are separately linear in each of their arguments.

The set of these maps (for each given pair (k, l)) is also a vector space, –with the obvious operations– and its elements are called **tensors of type** $\binom{l}{k}$.

Exercise: What is the dimension of these spaces as a function of the pair $\binom{l}{k}$?

Note: In finite dimensions, it can be shown that any tensor of type $\binom{l}{k}$ can be written as a linear combination of elements of the Cartesian product of k copies of V^* and l copies of V –where we have identified V with V^{**} –. For example, if t is of type $\binom{0}{2}$, –that is, a map that has as arguments two covectors–, then given a basis $\{e_i\}$ of V , and the corresponding basis of V^{**} , $\{E_i\}$ there will be $n \times n$ real numbers t^{ij} , $i = 1, \dots, n$ such that

$$t(\sigma, \omega) = \sum_{i,j=1}^n t^{ij} E_i(\sigma) E_j(\omega) = \sum_{i,j=1}^n t^{ij} \sigma(e_i) \omega(e_j), \quad \forall \sigma, \omega \in V^*. \quad (2.5)$$

But the set of linear combinations of Cartesian products of k copies of V^* and l copies of V is also a vector space, it is called the **outer product** of k copies of V^* and l copies of V and is denoted by

$$\underbrace{V^* \otimes V^* \otimes \dots \otimes V^*}_{k \text{ times}} \otimes \underbrace{V \otimes \dots \otimes V}_{l \text{ times}}.$$

Therefore, tensors can also be considered as elements of these outer products.

Example: a) Let t be of type $\binom{0}{2}$, that is, $t \in V^* \otimes V^*$. This is a bilinear map from $V \otimes V$ to \mathbb{R} , $t(v, u) \in \mathbb{R}$. Let t be symmetric [$t(v, u) = t(u, v)$] and non-degenerate [$t(v, \cdot) = 0 \in V^* \implies v = 0$]. Since t is non-degenerate, it defines an invertible map between V and its dual. Indeed, given $v \in V$, $t(v, \cdot)$ is an element of V^* . But if v and \tilde{v} determine the same element of V^* , that is, if $t(v, u) = t(\tilde{v}, u)$; $\forall u \in V$ then $v = \tilde{v}$, which can be seen by taking $u = v - \tilde{v}$ and using that t is non-degenerate. Since the dimensions of V and V^* are equal, the map thus defined is invertible.

Example: b) Let ε be an element of $\binom{0}{n}$ such that

$$\varepsilon(\dots, \underbrace{v}_i, \dots, \underbrace{u}_j, \dots) = -\varepsilon(\dots, \underbrace{u}_i, \dots, \underbrace{v}_j, \dots) \quad (2.6)$$

for any box i and j , that is, a totally antisymmetric tensor. Let e_i be a basis of V and $\varepsilon_{123\dots n} := \varepsilon(e_1, e_2, \dots, e_n)$, then any other component of ε in this basis will be either zero or $\varepsilon_{123\dots n}$ or $-\varepsilon_{123\dots n}$ depending on whether some e_i is repeated or is an even permutation of the above or an odd one. Indeed, for example,

$$\begin{aligned} \varepsilon_{3124\dots n} &:= \varepsilon(e_3, e_1, e_2, e_4, \dots, e_n) \\ &= -\varepsilon(e_1, e_3, e_2, e_4, \dots, e_n) \\ &= \varepsilon(e_1, e_2, e_3, e_4, \dots, e_n) \\ &= \varepsilon_{1234\dots n}. \end{aligned}$$

Therefore, given a basis, a single number, $\varepsilon_{123\dots n}$, is enough to determine the tensor ε and given another tensor $\tilde{\varepsilon}$ not identically zero with the properties mentioned above, there will exist a number α such that $\varepsilon = \alpha\tilde{\varepsilon}$. This last equality does not depend on the basis used and tells us that the dimension of the subspace of antisymmetric tensors of type $\binom{0}{n}$ is 1. Knowing one element is enough to generate the entire space by multiplying it by any real number.

Exercise: Let ε be not identically zero and u_i a set of $n = \dim V$ vectors of V . Show that these form a basis if and only if

$$\varepsilon(u_1, \dots, u_n) \neq 0. \quad (2.7)$$

Example: c) Let A be an element of $\binom{1}{1}$,

$$u \in V, ;; v \in V^* \rightarrow A(u, v^*) \in \mathbb{R}. \quad (2.8)$$

This implies that $A(u, \cdot)$ is also a vector (identifying V with V^{**} , the one that takes a form $\omega \in V^*$ and gives the number $A(u, \omega)$). Thus we have a **linear map** from V to V , that is, a linear operator on V .

Exercise: Let u_i be a basis of V and let a_i be the vectors $A(u_i, ;;)$, then

$$\varepsilon(a_1, \dots, a_n) = \varepsilon(A(u_1, \cdot), \dots, A(u_n, ;;))$$

is totally antisymmetric in the u_i and therefore proportional to itself;

$$\varepsilon(A(u_1, \cdot), \dots, A(u_n, \cdot)) \propto \varepsilon(u_1, \dots, u_n).$$

The proportionality constant is called the **determinant** of the operator A ,

$$\varepsilon(A(u_1, ;;), \dots, A(u_n, ;;)) =: \det(A) \varepsilon(u_1, \dots, u_n).$$

Problem 2.1 Show that this definition does not depend on the ε used nor on the basis and therefore is truly a function of the space of operators from V to \mathbb{R} .

Exercise: If A and B are two operators on V , then $A \cdot B(v) := A(B(v))$. Show that $\det(AB) = \det(A) \cdot \det(B)$.

2.1.2 Complexification

Another way to obtain vector fields from a given one, say V , is by extending the field where the multiplication operation is defined, if this is possible. The most common case is the **complexification** of a real vector space; in this case, the product is simply extended to complex numbers, resulting in a vector field of the same dimension. One way to obtain it, for example, is by taking a set of linearly independent vectors

from the initial space, that is, a basis, and considering all linear combinations with arbitrary complex coefficients. The space thus obtained is denoted by V^C . If the components of the vectors in V in the original basis were n-tuples of real numbers, they are now n-tuples of complex numbers. Since the basis is the same, the dimension is also the same. These extensions of vector spaces often appear, and we will see others later.

Multilinear maps must be extended in the same way, that is, for example, the dual of V will consist of all maps from V to C .

We can also take smaller fields, for example, Q^n or Z^n .

Example: Consider the vector space Q^n consisting of all n-tuples of rational numbers. In this space, the field is also the set of rationals. If we extend the field to the reals, we obtain R^n .

Example: Consider a space V and any basis in it, e_i . This choice characterizes a subspace of V , given by all elements of the form,

$$v = \sum_i^n m^i e_i \quad m^i \in Z.$$

Exercise: Now consider the set of all linear maps from this subspace to Z . What form do their elements take?

2.1.3 Quotient Spaces

The last way we will see to obtain vector spaces from other vector spaces is by taking **quotients**. Let V be a vector space and let $W \subset V$ be a subspace of it. We will call the **quotient space** the set of equivalent classes in V , where we will say that two vectors in V are equivalent if their difference is a vector in W . This space is denoted as V/W .

Exercise: Prove that this is an equivalence relation. [At the end of the chapter there is a box with a discussion of the relevant definitions and properties of the central concept of equivalence relations.]

Let's see that this is a vector space. The elements of V/W are equivalent classes; two elements of V , v and v' , belong to the same equivalent class if $v - v' \in W$. Let ζ and ζ' be two elements of V/W , that is, two equivalent classes of elements of V . We will define the operations proper to vector spaces as follows: $\zeta + \alpha\zeta'$ will be the equivalent class corresponding to an element \tilde{v} obtained by taking an element of V in ζ , say v , another in ζ' , say v' , and defining $\tilde{v} := v + \alpha v'$, we have $\tilde{\zeta} = \zeta + \alpha\zeta'$, where $\tilde{\zeta}$ is the equivalent class containing the element $\tilde{v} = v + \alpha v'$. To facilitate notation, the equivalent class containing a given element, say v , is usually represented as v . In this case, we have,

$$v + \alpha v' = v + \alpha v'.$$

Exercise: See that this definition does not depend on the choice of elements in the equivalent class taken to perform the operation. That is, consider two other elements in ζ and ζ' , say \hat{v} and \hat{v}' , and see that with them you obtain an element in the same class as $\tilde{v} = v + \alpha v'$.

Example: Let $V = \mathbb{R}^2$, that is, the space of 2-tuples of real numbers. Let v be any element. This element generates the one-dimensional space W_v consisting of all vectors of the form αv , for $\alpha \in \mathbb{R}$. The quotient space V/W_v is the space composed of lines parallel to v . That is, each line is an element of the quotient space, and there is a notion of addition and scalar multiplication among them.

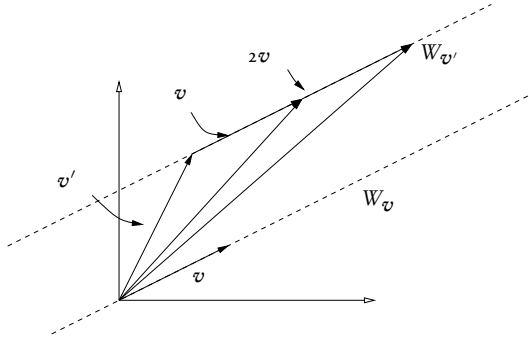


Figure 2.1: Geometric interpretation of the quotient space.

Exercise: Let V be the set of continuous functions from \mathbb{R} to \mathbb{R} and let W be the subset of those that vanish in the interval $[0, 1]$. See that this is a vector subspace. Consider the space V/W . What space can you associate with this?

2.2 Norms

Definition: A **norm** in a vector space V is a map $|x| : V \rightarrow \mathbb{R}^+$, satisfying for all $x, y \in V, \alpha \in \mathbb{R}$,

- i) $|x| \geq 0$ ($\Leftrightarrow x = 0$)
- ii) $|\alpha x| = |\alpha| |x|$
- iii) $|x + y| \leq |x| + |y|$

Examples: In \mathbb{R}^2 :

- a) $|(x, y)| := \max|x|, |y|$;
- b) $|(x, y)|_2 := \sqrt{x^2 + y^2}$, Euclidean norm;
- c) $|(x, y)|_1 := |x| + |y|$;
- d) $|(x, y)|_p := (|x|^p + |y|^p)^{\frac{1}{p}}$; ; ; ; ; $p \geq 1$;

e) In V let t be a positive definite symmetric tensor of type $\binom{0}{2}$, that is $t(u, v) = t(v, u)$, $t(u, u) \geq 0$ ($= \leftrightarrow; u = 0$). The function $|u|_t = \sqrt{t(u, u)}$ is a norm. Each tensor of this type generates a norm, but there are many norms that do not come from any tensor of this type. Give an example.

Exercise: Prove that $|t(u, v)|^2 \leq \|u\|_t \|v\|_t$. Hint: Consider the polynomial: $P(\lambda) := t(u + \lambda v, u + \lambda v)$.

Exercise: Prove that the given examples are indeed norms. Graph the level curves of the first four norms, that is, the sets $S_a = \{(x, y) \in \mathbb{R}^2; |(x, y)| = a\}$ and the "balls of radius a ", that is $B_a = \{(x, y) \in \mathbb{R}^2; |(x, y)| \leq a\}$.

Exercise: Prove that the map $d : V \times V \rightarrow \mathbb{R}^+$ given by $d(x, y) = |x - y|$ is a metric.

What is a norm geometrically? Given a vector $x \neq 0$ of V and any positive number, a , there is a unique number $\alpha > 0$ such that $|\alpha x| = a$. This indicates that the level surfaces of the norm, that is, the hypersurfaces $S_a = \{x \in V; |x| = a, a > 0\}$ form a smooth family of layers one inside the other, and each of them divides V into three disjoint sets, the *interior* of S_a —containing the element $x = 0$ —, S_a and the *exterior* of S_a . The *interior* of S_a is a convex set, that is, if x and y belong to the interior of S_a , then $\alpha x + (1 - \alpha)y$, $\alpha \in [0, 1]$ also belongs [since $|\alpha x + (1 - \alpha)y| \leq \alpha|x| + (1 - \alpha)|y| \leq \alpha a + (1 - \alpha)a = a$]. A level curve completely characterizes a norm in the sense that if we give a subset N of V , such that N : **a)** has the radial property, that is, given $x \neq 0$ there is a unique $\alpha > 0$ such that $\alpha x \in N$ and $-\alpha x \in N$ and **b)** is convex, then there is a unique norm such that N is the level surface S_1 . This norm is defined as follows: given x we know that there will be a unique $\alpha > 0$ such that $\alpha x \in N$ and then the norm of x will be $|x| := \frac{1}{\alpha}$.

Exercise: Prove that this is a norm. Hint, given any two vectors $x, y \in V$, then $\frac{x}{\|x\|}$ and $\frac{y}{\|y\|}$ are unitary and therefore are in N . But then we have that $\|\lambda \frac{x}{\|x\|} - (1 - \lambda) \frac{y}{\|y\|}\| \leq 1; \forall \lambda \in [0, 1]$. Now find a convenient value for λ .

From this image, we see that given two norms of a finite-dimensional vector space and a level surface of one, there will be level surfaces of the other that will have the first one inside or outside it. In the norms a) and b) of the previous example, we see that given a square containing zero, there will be two circles containing zero, one containing the square and the other contained by it. This leads us to the following theorem.

Theorem 2.1 *Let V be a finite-dimensional vector space. Then all its norms are equivalent to each other, in the sense that given $|\cdot|$ and $|\cdot|'$ there are positive constants M_1 and M_2 such that for all $x \in V$ it holds that $M_1|x| \leq |x'| \leq M_2|x|$.*

Proof: We will show that all are equivalent to the norm $|x|_1 = \sum_i |a^i|$, where the a^i are the components of x , with respect to a given basis e_i , $x = a^i e_i$.

Let any other norm, then

$$\begin{aligned}
 |,|x|,| &\leq |x| = |\sum_i^n a^i, e_i| \leq \sum_i^n |a^i, e_i| \\
 &\leq \sum_i^n |a^i|, |e_i| \leq (\max_{j=1, \dots, n} |e_j|) \sum_i |a^i| \\
 &= (\max_{j=1, \dots, n} |e_j|); |x|_1.
 \end{aligned} \tag{2.9}$$

And we have easily obtained the upper bound. Now let's see the lower bound. To do this, we must prove that the norm $|\cdot|$ is, as a function from V to \mathbb{R}^+ , a continuous function. This easily follows from the already found bound, indeed, let any other vector, $y = b^i, e_i$, then

$$\begin{aligned}
 |,|x| - |y|,| &\leq |x - y| \\
 &= (\max_{j=1, \dots, n} |e_j|); |x - y|_1
 \end{aligned} \tag{2.10}$$

This shows that the norm $|\cdot|$ is a continuous function with respect to the norm $|\cdot|_1$. Let S_1 be the level surface of radius 1 with respect to the metric $|\cdot|_1$. S_1 is a closed and bounded set and therefore compact. Therefore, by continuity, $|\cdot|$ has a maximum value, M_2 , and a minimum, M_1 , which give the sought inequality. The maximum value we have already found, the minimum is what allows us to bound the norm from below and conclude the theorem.

Notes:

i) In this proof, it is crucial that S_1 is compact. If V is infinite-dimensional, this is not the case, and there are many non-equivalent norms.

ii) For our purposes, any norm is sufficient –since if, for example, $f : V \rightarrow \mathbb{R}$ is continuous with respect to one norm, it is also continuous with respect to any other equivalent to it– and for simplicity, from now on, we will use the Euclidean norm.

iii) In this sense, the norms of finite-dimensional vector spaces are equivalent to the one generated by any positive-symmetric element of the exterior product of its dual with itself.

iv) Since equivalent norms generate the same topology, we see that in finite-dimensional vector spaces, there is a unique topology associated with all its possible norms. This is usually called the **strong topology**.

Exercise: Prove that the above is indeed an equivalence relation.

2.2.1 Induced Norms in V^*

The norms defined in V naturally induce norms in its dual, V^* . This is given by:

$$|\omega| := \max_{|v|=1} |\omega(v)|. \tag{2.11}$$

Exercise: Show that this is a norm and that $|\omega(v)| \leq |\omega| |v|$.

Exercise: Consider $V = \mathbb{R}^2$ with the norm: $|(x, y)| := \max|x|, |y|$. What is the induced norm in V^* ?

2.3 Linear Operator Theory

A **linear operator** A on a vector space V is a continuous map⁴ from V to V such that $\forall x, y \in V, \alpha \in \mathbb{R}, A(\alpha x + y) = \alpha A(x) + A(y)$, that is, a tensor of type $\binom{1}{1}$.

The set of linear operators \mathcal{L} is an algebra, that is, a vector space with a product. Indeed, if $A, B \in \mathcal{L}$, $\alpha \in \mathbb{R}$, then $A + \alpha B \in \mathcal{L}$ and also $A \cdot B$ (the operator that sends $x \in V$ to $A(B(x)) \in V$) also belongs to \mathcal{L} . Due to this property, we can also define non-linear functions from \mathcal{L} to \mathbb{R} and maps from \mathcal{L} to \mathcal{L} . To study the continuity and differentiability of these maps, we introduce a norm in \mathcal{L} , the most convenient being the following norm induced by the one used in V ,

$$\|A\|_{\mathcal{L}} = \max_{\|x\|_V=1} \|A(x)\|_V. \quad (2.12)$$

If V is finite-dimensional (which we will assume from now on), the vector space \mathcal{L} is also finite-dimensional and therefore all its norms are equivalent. The fact that the norm of A is finite again uses that $A : V \rightarrow V$ is continuous and that $\{x \in V / \|x\|_V = 1\}$ is compact, in the infinite-dimensional case, neither of these things is necessarily true, and within \mathcal{L} we only have a subspace of linear operators with finite (bounded) norms.

Exercise: Show that

$$\|A(v)\| \leq \|A\|_{\mathcal{L}} \|v\|.$$

Exercise: Using the result of the previous exercise, show that

$$\|AB\|_{\mathcal{L}} \leq \|A\|_{\mathcal{L}} \|B\|_{\mathcal{L}}.$$

Exercise: Show that $\|\cdot\|_{\mathcal{L}} : \mathcal{L} \rightarrow \mathbb{R}^+$ is a norm.

Next, we study various functions in the space of operators.

The determinant of an operator, introduced in the previous section, is a polynomial of degree $n = \dim V$ in A and therefore differentiable. Using the chain rule, we see that $\det(I + \varepsilon A)$ is differentiable in ε , and indeed a polynomial of degree n in ε . Each of the coefficients of this polynomial is a function of A . Of importance in what follows is the linear coefficient in A , which is obtained using the formula

$$\frac{d}{d\varepsilon} \det(I + \varepsilon A)|_{\varepsilon=0} = \frac{\varepsilon(A(u_1), u_2, \dots, u_n) + \dots + \varepsilon(u_1, \dots, A(u_n))}{\varepsilon(u_1, \dots, u_n)} \quad (2.13)$$

this function is called the **trace** of A and is denoted $\text{tr}(A)$.

⁴With respect to the topology induced by any of the equivalent norms of V .

Among the maps from \mathcal{L} to \mathcal{L} , consider the exponential map, defined as,

$$e^A = \sum_{i=0}^{\infty} \frac{A^i}{i!} = I + A + \frac{A^2}{2} + \dots \quad (2.14)$$

Theorem 2.2 $e^A \in \mathcal{L}$ if $A \in \mathcal{L}$ and e^{tA} is infinitely differentiable with respect to t .

Proof: Consider the Cauchy sequence $\{e_n^A\}$, where $e_n^A \equiv \sum_{i=0}^n \frac{A^i}{i!}$. This sequence is Cauchy because, taking $m > n$, we have

$$\begin{aligned} \|e_m^A - e_n^A\|_{\mathcal{L}} &= \left\| \frac{A^m}{(m)!} + \frac{A^{m-1}}{(m-1)!} + \frac{A^{m-2}}{(m-2)!} + \dots + \frac{A^{n+1}}{(n+1)!} \right\|_{\mathcal{L}} \\ &\leq \left\| \frac{A^m}{(m)!} \right\|_{\mathcal{L}} + \left\| \frac{A^{m-1}}{(m-1)!} \right\|_{\mathcal{L}} + \left\| \frac{A^{m-2}}{(m-2)!} \right\|_{\mathcal{L}} + \dots + \left\| \frac{A^{n+1}}{(n+1)!} \right\|_{\mathcal{L}} \\ &\leq \frac{\|A\|_{\mathcal{L}}^m}{(m)!} + \frac{\|A\|_{\mathcal{L}}^{m-1}}{(m-1)!} + \frac{\|A\|_{\mathcal{L}}^{m-2}}{(m-2)!} + \dots + \frac{\|A\|_{\mathcal{L}}^{n+1}}{(n+1)!} \\ &= \|e_m^A - e_n^A\|_{\mathcal{L}} \rightarrow 0. \end{aligned} \quad (2.15)$$

Where $e_n^A \equiv \sum_{i=0}^n \frac{\|A\|_{\mathcal{L}}^i}{i!}$ and the last implication follows from the fact that the numerical series $e^{\|A\|_{\mathcal{L}}}$ converges. But by completeness⁵ of \mathcal{L} , every Cauchy sequence converges to some element of \mathcal{L} that we will call e^A . The differentiability of e^{tA} follows from the fact that if a series $\sum_{i=0}^{\infty} f_i(t)$ is convergent and $\sum_{i=0}^{\infty} \frac{df_i}{dt}$ is uniformly convergent, then $\frac{d}{dt} \sum_{i=0}^{\infty} f_i(t) = \sum_{i=0}^{\infty} \frac{d}{dt} f_i(t)$ ♠

Exercise: Show that

- a) $e^{(t+s)A} = e^{tA} \cdot e^{sA}$,
- b) If A and B commute, that is if $AB = BA$, then $e^{A+B} = e^A e^B$.
- c) $d e^t(e^A) = e^{tr(A)}$
- d) $\frac{d}{dt} e^{tA} = A e^{tA}$.

Hint: For point c) use that e^A can also be defined as,

$$e^A = \lim_{m \rightarrow \infty} \left(I + \frac{A}{m} \right)^m.$$

⁵Every finite-dimensional real vector space is complete.

2.3.1 Matrix Representation

To describe certain aspects of linear operators, it is convenient to introduce the following matrix representation.

Let $\{\mathbf{u}_i\}$, $i = 1, \dots, n$ be a basis of V , that is, a set of linearly independent vectors of V [$\sum_{i=1}^n c^i \mathbf{u}_i = \mathbf{o} \implies c^i = 0$] that span it [if $\mathbf{v} \in V$, there exist numbers $\{v^i\}$, $i = 1, \dots, n$ such that $\mathbf{v} = \sum_{i=1}^n v^i \mathbf{u}_i$]. Applying the operator A to a member of the basis \mathbf{u}_i we obtain a vector $A(\mathbf{u}_i)$ which in turn can be expanded in the basis, $A(\mathbf{u}_i) = \sum_{j=1}^n A^j_i \mathbf{u}_j$. The matrix thus constructed, A^j_i , is a representation of the operator A in that basis. In this language, we see that the matrix A^j_i transforms the vector of components $\{v^i\}$ into the vector of components $\{A^j_i v^i\}$. Given a basis, $\{\mathbf{u}_i\}$, and a matrix A^j_i , we can construct a linear operator as follows: Given the basis, we define its co-basis, that is, a basis in V^* as $\{\theta^i\}$, $i = 1, \dots, n$, such that $\theta^i(\mathbf{u}_j) = \delta^i_j$, then $A = \sum_{i,j=1}^n A^j_i \mathbf{u}_j \theta^i$.

If we change the basis, the matrices representing the operators will change. Indeed, if we take another basis $\{\hat{\mathbf{u}}_i\}$ and write its components with respect to the previous basis as $\hat{\mathbf{u}}_i = P^k_i \mathbf{u}_k$, and therefore, $\hat{\theta}^j = (P^{-1})^j_l \theta^l$, then the relationship between the components of the operator A in both bases is given by

$$\hat{A}^j_i = A(\hat{\theta}^j, \hat{\mathbf{u}}_i) = (P^{-1})^j_l A^l_k P^k_i \quad \text{or} \quad \hat{A} = P^{-1} A P \quad (2.16)$$

that is, \hat{A} and A are **similar matrices**.

Exercise: See, from its definition (equation (2.13)), that in a basis we have, $\text{tr} A = \sum_{i=1}^n A^i_i$.

Exercise: See with an example in two dimensions that the definition of determinant conforms with the usual one when we use a basis.

2.3.2 Invariant Subspaces

Definition: Let $A : V \rightarrow V$ be an operator and let W be a subspace of V . We will say that W is an **invariant subspace** of A if $AW \subseteq W$.

The invariant subspaces of an operator are important because they allow us to understand its action. Note that given any operator A , there are always at least two invariant spaces, V and $\{\mathbf{o}\}$, and in reality, many more. For example, as we will see later, given any number between 1 and n ($= \dim V$), there exists an invariant subspace with that number as its dimension. The ones that truly encode the action of the operator are its **irreducible** invariant subspaces, that is, those that cannot be further decomposed into invariant subspaces such that their direct sum is the whole V ⁶

⁶A vector space is said to be the direct sum of two of its subspaces, W_1 and W_2 , and is denoted by

Example: Let V with $\dim(V) = 2$ and $A : V \rightarrow V$ given by, $A(u_1) = \lambda_1 u_1$, $A(u_2) = \lambda_2 u_2$, where (u_1, u_2) are linearly independent (and therefore a basis). Note that this completely defines the operator, since given any $v \in V$, we can uniquely write it as $v = v^1 u_1 + v^2 u_2$ and therefore,

$$A(v) = \lambda_1 v^1 u_1 + \lambda_2 v^2 u_2.$$

In this case, the invariant subspaces are $\text{Span}\{u_1\}$ and $\text{Span}\{u_2\}$, and clearly, we have $V = \text{Span}\{u_1\} \oplus \text{Span}\{u_2\}$. Since each of these invariant subspaces is one-dimensional, the action of the operator on them is simply a dilation, that is, the multiplication of their elements by a number. Note that in the case where $\lambda_1 = \lambda_2$, the operator is proportional to the identity and therefore we have infinite invariant spaces.

Exercise: Let V be the space from the previous example and let A be given by $A(u_1) = 0$, $A(u_2) = u_1$. Find its irreducible invariant subspaces. Do the same for the operator given by $A(u_1) = u_1$, $A(u_2) = a u_2 + u_1$. What happens when $a = 1$?

We will study in detail the one-dimensional invariant subspaces, note that they are irreducible. To study the invariant spaces, it is convenient to study the invariant subspaces of the operator when its action is extended to V^C , that is, the **complexification** of V .

Let's see that an operator always has at least one one-dimensional invariant subspace (and therefore always has a non-trivial irreducible invariant subspace).

Lemma 2.1 *Given $A : V^C \rightarrow V^C$, where V^C is finite-dimensional, there always exists a $u \in V^C$ and a $\lambda \in C$ such that,*

$$(A - \lambda I)u = 0 \quad (2.17)$$

Proof:

A solution to this equation consists of a scalar λ , called the **eigenvalue** of the operator A , and a vector u , called the **eigenvector** of the operator A . The subspace of V^C given by $\{\alpha u \mid \alpha \in C\}$ is the invariant subspace sought.

It is clear that the system has a solution if and only if $\det(A - \lambda I) = 0$. But this is a polynomial in λ of order equal to the dimension of V and therefore, by the Fundamental Theorem of Algebra, it has at least one solution or root, (generally complex), λ_1 , and therefore there will be, associated with it, at least one u_1 solution of (2.17) with $\lambda = \lambda_1$. ♠

The need to consider all these solutions is what leads us to treat the problem for complex vector spaces.

Exercise: In the infinite-dimensional case, the theorem is no longer true. Find an example of an operator on the set of infinite tuples without any eigenvector. Find it

$V = W_1 \oplus W_2$ if each element of V can be uniquely written as the sum of two elements, one from each of these subspaces.

by looking for infinite matrices constructed in such a way that they have only one eigenvector and consider the limit to infinite components.

Application: Schur's Triangulation Lemma

Definition: An $n \times n$ matrix A^j_i has an upper triangular form if $A^j_i = 0 \quad \forall j > i, j, i = 1, \dots, n$. That is, it is a matrix of the form,

$$A = \begin{pmatrix} A^1_1 & A^1_2 & \cdots & \cdots & A^1_n \\ 0 & A^2_2 & \cdots & \cdots & A^2_n \\ 0 & 0 & \ddots & \ddots & A^3_n \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & \cdots & 0 & A^n_n \end{pmatrix}. \quad (2.18)$$

As we will see later, in chapter ??, this is a very convenient form to understand the solutions to systems of ordinary differential equations. And the most attractive thing about it is that any operator has a matrix representation with an upper triangular form! Moreover, if an inner product is present, the basis for this representation can be chosen to be orthonormal.

Lemma 2.2 (Schur) *Let $A : V \rightarrow V$ be a linear operator acting on a finite-dimensional complex vector space V , n , and let (\cdot, \cdot) be an inner product on V . Then, there exists an orthonormal basis $\{u_i\}$, $i = 1, \dots, n$ with respect to which the matrix representation of A is upper triangular.*

Proof: Consider the eigenvalue-eigenvector problem for A ,

$$(A - \lambda I)(u) = 0. \quad (2.19)$$

As we have already seen, this problem always has at least one non-trivial solution, and therefore we have a pair (λ_1, u_1) solution of the problem. We take u_1 of unit norm as the first element of the basis to be determined. We then have

$$A^j_1 := \theta^j(A(u_1)) = \theta^j(\lambda_1 u_1) = \lambda_1 \delta^j_1,$$

which gives us the result for the first column of the matrix.

Now consider the space

$$V_{n-1} = \text{Span}\{u_1\}^\perp := \{u \in V | (u, u_1) = 0\}$$

and the operator from $V_{n-1} \rightarrow V_{n-1}$ given by $A_1 := (I - u_1 \theta^1)A$. Note that as we form an orthonormal basis, we already know the first member of the co-basis, $\theta^1 = (u_1, \cdot)$. The operator $P_1 := I - u_1 \theta^1$ satisfies $P_1(u_1) = 0$, $(P_1(v), u_1) = 0$ and $P_1 \cdot P_1 = P_1$, that is, it is a projection operator in the subspace V_{n-1} .

We then have that $A_1 : V_{n-1} \rightarrow V_{n-1}$. Therefore, in this space, we can also pose the eigenvalue-eigenvector equation,

$$(A_1 - \lambda I)u = ((I - u_1 \theta^1)A - \lambda I)u = 0. \quad (2.20)$$

We thus obtain a new pair (λ_2, u_2) , with $u_2 \in V_{n-1}$, and therefore perpendicular to u_1 and also, $Au_2 = \lambda_2 u_2 + u_1 \theta^1(A(u_2))$. Therefore

$$A^j_2 = \theta^j(A(u_2)) = \theta^j(\lambda_2 u_2 + u_1 \theta^1(A(u_2))) = \lambda_2 \delta^j_2 + \delta^j_1 A^1_2.$$

We thus see that with this choice of basis element, the second column of A satisfies the condition of the Lemma. The next step is to consider the subspace,

$$V_{n-2} = \text{Span}\{u_1, u_2\}^\perp = \{u \in V | (u_1, u) = (u_2, u) = 0\},$$

and there the eigenvalue-eigenvector equation for the operator $(I - u_1 \theta^1 - u_2 \theta^2)A$. Proceeding in this way, we generate the entire basis ♠

The previous proof used an inner product that is actually exogenous to the property itself. We did it this way because the proof is conceptually simple and useful in the case that we are in the presence of a given inner product. However, the previous theorem can be proven using the notion of quotient space and thus dispensing with the inner product. This proof can be obtained by doing the following exercises:

Exercise: Let $A : V \rightarrow V$ be a linear operator, let $W \subset V$ be invariant under A , that is, $A[W] \subset W$. This operator induces an operator in the quotient space V/W as follows: $\hat{A}\zeta = \tilde{\zeta}$ where $\tilde{\zeta}$ is the equivalent class to which $A(u)$ belongs when u belongs to ζ . See that this definition is consistent, that is, if we choose another element $v \in \zeta$, we obtain that $A(v) \in \tilde{\zeta}$.

Exercise: The induced operator will have at least one eigenvalue-eigenvector pair in V/W , that is, there will be a pair $(\hat{\lambda}, \zeta)$ such that $\hat{A}\zeta = \hat{\lambda}\zeta$. What does this equation mean in terms of the operator A in V ?

Exercise: Specialize the previous case when W is the subspace of V generated by an eigenvector of A . $Au_1 = \lambda_1 u_1$, $W_1 = \text{Span}\{u_1\}$. What does it mean, in terms of the space V and the operator A , that $\hat{A} : V/W_1 \rightarrow V/W_1$ has an eigenvalue-eigenvector pair?

Exercise: Iterate the previous procedure, taking at each step an element from each equivalent class of generalized eigenvectors to form a basis where the matrix representation of A is upper triangular.

Continuing now with the study of invariant subspaces. If $\det(A - \lambda I)$ has $m \leq n$ distinct roots, $\{\lambda_i\}$, $i = 1, \dots, m$, then there will be at least one complex eigen-

vector \mathbf{u}_i associated with each of them. Let's see that these form distinct invariant subspaces.

Lemma 2.3 *Let $\{(\lambda_i, \mathbf{u}_i)\}$ $i = 1 \dots m$ be a set of eigenvalue-eigenvector pairs. If $\lambda_i \neq \lambda_j \quad \forall i \neq j, i, j = 1 \dots m$, then these eigenvectors are linearly independent.*

Proof: Suppose by contradiction that they are not, and therefore there exist constants $c^i \in \mathbb{C}, i = 1, \dots, m-1$, such that

$$\mathbf{u}_m = \sum_{i=1}^{m-1} c^i \mathbf{u}_i \quad (2.21)$$

Applying A to both sides, we get

$$A\mathbf{u}_m = \lambda_m \mathbf{u}_m = \sum_{i=1}^{m-1} c^i \lambda_i \mathbf{u}_i \quad (2.22)$$

or,

$$0 = \sum_{i=1}^{m-1} c^i (\lambda_m - \lambda_i) \mathbf{u}_i. \quad (2.23)$$

We conclude that $\{\mathbf{u}_i\} \quad i = 1, \dots, m-1$ are linearly dependent. Due to 2.30 and the hypothesis that the eigenvalues are distinct, at least one of the coefficients must be non-zero, and therefore we can solve for one of the remaining eigenvectors in terms of the others $m-2$. Repeating this procedure $(m-1)$ times, we arrive at the conclusion that $\mathbf{u}_1 = 0$, which is a contradiction since, as we have seen, the eigenvector equation always has a non-trivial solution for each distinct eigenvalue ♠

If for each eigenvalue there is more than one eigenvector, then these form a higher-dimensional invariant vector subspace (reducible). Within each of these subspaces, we can take a basis composed of eigenvectors. The previous lemma ensures that all these eigenvectors thus chosen, for all eigenvalues, form a large linearly independent set.

Exercise: Convince yourself that the set of eigenvectors with the same eigenvalue forms a vector subspace.

If a given operator A has all its eigenvalues distinct, then the corresponding eigenvectors are linearly independent and equal in number to the dimension of V , that is, they generate a basis of $V^{\mathbb{C}}$. In that basis, the matrix representation of A is diagonal, that is, $A^j_i = \delta^j_i \lambda_i$. Each of its eigenvectors generates an irreducible invariant subspace, and together they generate $V^{\mathbb{C}}$. In each of them, the operator A acts merely by multiplication by λ_i . Note that the λ_i are generally complex, and therefore such multiplication is actually a rotation plus a dilation. Note that, unlike the basis of

Schur's triangulation lemma, this is not generally orthogonal with respect to any given inner product.⁷

Example: Let V be a vector space with $\dim(V) = 2$ and let A be given by $A(e_1) = e_2$, $A(e_2) = -e_1$, where (e_1, e_2) are any two linearly independent vectors. If we interpret them as two orthonormal vectors, then A is a rotation by $\pi/2$ in the plane.

Now let's calculate the determinant of $A - \lambda I$,

$$\begin{aligned} \det(A - \lambda I) &= \varepsilon((A - \lambda I)e_1, (A - \lambda I)e_2) / \varepsilon(e_1, e_2) \\ &= \varepsilon(e_2 - \lambda e_1, -e_1 - \lambda e_2) / \varepsilon(e_1, e_2) \\ &= 1 + \lambda^2. \end{aligned} \quad (2.24)$$

and therefore the eigenvalues are $\lambda_1 = \iota$, and $\lambda_2 = -\iota$. The eigenvectors are $u_1 = e_1 + \iota e_2$ and $u_2 = e_1 - \iota e_2 = \bar{u}_1$. We see then that the action of A in these subspaces is multiplication by $\pm \iota$ and that both invariant subspaces are genuinely complex. In this new basis, the space V is generated by all linear combinations of the form $zu_1 + \bar{z}u_2$, and the action of A is simply multiplication by ι of z .

If the multiplicity of any of the roots $\det(A - \lambda I) = 0$ is greater than one, there will be fewer eigenvalues than the dimension of the space, and therefore we will not be guaranteed to have enough eigenvectors to form a basis, as we can only guarantee the existence of one for each eigenvalue.

Example: Let V be the set of 2-tuples of real numbers with a generic element (a, b) and let A be given by $A(a, b) = (\lambda a + \epsilon b, \lambda b)$. Taking a basis, $e_1 = (1, 0)$, $e_2 = (0, 1)$, we see that its matrix representation is:

$$\begin{pmatrix} \lambda & \epsilon \\ 0 & \lambda \end{pmatrix} \quad (2.25)$$

We see that this operator has only one eigenvalue with multiplicity 2. But it has only one eigenvector, proportional to $e_1 = (1, 0)$. For future use, note that if we define $\Delta = A - \lambda I$, then $e_1 = \frac{1}{\epsilon} \Delta e_2$, and therefore, in the basis $\{\tilde{e}_1 = e_1, \tilde{e}_2 = \frac{1}{\epsilon} e_2\}$, the operator is represented by the matrix,

$$\begin{pmatrix} \lambda & 1 \\ 0 & \lambda \end{pmatrix} \quad (2.26)$$

We must therefore analyze what happens in these cases. To do this, let us define, given λ_i an eigenvalue of A , the following subspaces:

⁷Note, however, that it can be *declared* orthogonal by defining the inner product as $(u, v) = \sum_{i=1}^n \theta^i(\tilde{u})\theta^i(v)$

$$W_{\lambda_i p} = \{u \in V \mid (A - \lambda_i I)^p u = 0\} \quad (2.27)$$

Note that these are invariant spaces: $AW_{\lambda_i p} \subset W_{\lambda_i p}$. Moreover, $W_{\lambda_i p} \subset W_{\lambda_i p+1}$, and therefore, for a sufficiently large p ($p \leq n$), we will have that $W_{\lambda_i p} = W_{\lambda_i p+1}$, taking the minimum among the p 's where this occurs, we define $W_{\lambda_i} := W_{\lambda_i p}$. Note that if for some λ_i , $p = 1$, then the subspace W_{λ_i} is composed of eigenvectors. These are the maximum invariant spaces associated with the eigenvalue λ_i , indeed we have:

Lemma 2.4 *The only eigenvalue of A in W_{λ_i} is λ_i .*

Proof: Let λ be an eigenvalue of A in W_{λ_i} . Let's see that $\lambda = \lambda_i$. As we have already seen, there will be an eigenvector $\zeta \in W_{\lambda_i}$ with λ as the eigenvalue. Since it is in W_{λ_i} , there will be some $p \geq 1$ such that $(A - \lambda_i I)^p \zeta = 0$, but since it is an eigenvector, we have that $(\lambda - \lambda_i)^p \zeta = 0$, and therefore $\lambda = \lambda_i$ ♠

Now let's see that these subspaces are linearly independent and generate all of V^C . We will prove this theorem again in Chapter ??.

Theorem 2.3 (See Chapter ??, Theorem 5.3) *Given an operator $A : V \rightarrow V$, with eigenvectors $\{\lambda_i\}$, $i = 1 \dots m$, the space V^C admits a direct decomposition into invariant subspaces W_{λ_i} , where in each of them A has only λ_i as an eigenvalue.*

Proof:

The W_{λ_i} are independent. Let $v_1 + \dots + v_s = 0$, with $v_i \in W_{\lambda_i}$, then we must prove that each $v_i = 0$. Applying $(A - \lambda_2 I)^{p_2} \dots (A - \lambda_s I)^{p_s}$ to the previous sum, we get, $(A - \lambda_2 I)^{p_2} \dots (A - \lambda_s I)^{p_s} v_1 = 0$, but since λ_i , $i \neq 1$, is not an eigenvalue of A in W_{λ_1} , the operator $(A - \lambda_2 I)^{p_2} \dots (A - \lambda_s I)^{p_s}$ is invertible⁸ in that subspace, and therefore $v_1 = 0$. Continuing in this way, we see that all the v_i must be zero.

The W_{λ_i} generate all of V^C . Suppose by contradiction that this is not the case, and consider V^C/W , where W is the space generated by all the W_{λ_i} , that is, the space of all linear combinations of elements in the W_{λ_i} . The operator A acts on V^C/W [$A\{u\} = \{Au\}$], and therefore it has an eigenvalue-eigenvector pair there. This implies that for some element ζ of V^C in some equivalent class of V^C/W , we have:

$$A\zeta = \lambda\zeta + u_1 + \dots + u_s \quad (2.28)$$

where the u_i belong to each W_{λ_i} . Now suppose that $\lambda \neq \lambda_i \ \forall i = 1 \dots s$, then $A - \lambda I$ is invertible in each W_{λ_i} , and therefore there exist vectors $\zeta_i = (A - \lambda I)^{-1} u_i \in W_{\lambda_i}$.

⁸Note that $(A - \lambda_i I)^s|_{W_{\lambda_j}}$ is invertible if its determinant is non-zero. But $\det(A - \lambda_i I)^s = (\det(A - \lambda_i I))^s = (\lambda_j - \lambda_i)^{s \dim(W_{\lambda_j})} \neq 0$

But then $\tilde{\zeta} := \zeta - \zeta_1 - \dots - \zeta_s$ is an eigenvalue of A ! This is impossible since λ is not a root of the characteristic polynomial, nor is $\tilde{\zeta} = 0$, since belonging to ζ to V^C/W is not a linear combination of elements in the W_{λ_i} . We thus have a contradiction. Now suppose that $\lambda = \lambda_j$ for some $j \in \{1..s\}$. We can still define the vectors $\zeta_i = (A - \lambda_j I)^{-1} u_i$ for all $i \neq j$, and $\tilde{\zeta}$, where we only subtract from ζ all the ζ_i with $i \neq j$, therefore we have that

$$(A - \lambda_j I) \tilde{\zeta} = u_i \quad (2.29)$$

But applying $(A - \lambda_j I)^{p_j}$ to this equation, with p_j the minimum value for which $W_{\lambda_j p_j} = W_{\lambda_j p_j + 1}$, we get that $\tilde{\zeta} \in W_{\lambda_j}$, and thus another contradiction, therefore it can only be that V^C/W is the trivial space, and the W_{λ_i} generate all of V^C ♠

We see that we only need to study each of these subspaces W_{λ_i} to find all their irreducible parts (from now on we will suppress the subscript i). But in these subspaces, the operator A acts very simply!

Indeed, let $\Delta_\lambda : W_\lambda \rightarrow W_\lambda$ be defined by $\Delta_\lambda := A|_{W_\lambda} - \lambda I|_{W_\lambda}$, then Δ has only 0 as an eigenvalue, and therefore it is **nilpotent**, that is, there exists an integer $m \leq n$ such that $\Delta_\lambda^m = 0$.

Lemma 2.5 *Let $\Delta : W \rightarrow W$ be such that its only eigenvalue is 0, then Δ is nilpotent.*

Proof: Let $W^p := \Delta^p[W]$, then we have that $W^p \subseteq W^q$ if $p \geq q$. Indeed, $W^p = \Delta^p[W] = \Delta^q[\Delta^{p-q}[W]] \subset \Delta^q[W]$. Since the dimension of W is finite, it must happen that for some integer p , we will have that $W^p = W^{p+1}$, we see then that Δ^p acts injectively on W^p and therefore cannot have 0 as an eigenvalue. But we have seen that every operator has some eigenvalue, and therefore we have a contradiction unless $W^p = \{0\}$. That is, $\Delta^p = 0$ ♠

Nilpotent operators have the important property of generating a basis of the space in which they act from their repeated application on a smaller set of linearly independent vectors. Continuing now with the study of invariant subspaces. If $\det(A - \lambda I)$ has $1 \leq m \leq n$ distinct roots, $\{\lambda_i\}$, $i = 1, \dots, m$, then there will be at least one complex eigenvector u_i associated with each of them. Let's see that these form distinct invariant subspaces.

Lemma 2.6 *Let $\{(\lambda_i, u_i)\}$ $i = 1 \dots m$ be a set of eigenvalue-eigenvector pairs. If $\lambda_i \neq \lambda_j \quad \forall i \neq j$, $i, j = 1 \dots m$, then these eigenvectors are linearly independent.*

Proof: Suppose by contradiction that they are not, and therefore there exist constants

$c^i \in C, i = 1, \dots, m-1$, such that

$$\mathbf{u}_m = \sum_{i=1}^{m-1} c^i \mathbf{u}_i \quad (2.30)$$

Applying A to both sides, we get

$$A\mathbf{u}_m = \lambda_m \mathbf{u}_m = \sum_{i=1}^{m-1} c^i \lambda_i \mathbf{u}_i \quad (2.31)$$

or,

$$0 = \sum_{i=1}^{m-1} c^i (\lambda_m - \lambda_i) \mathbf{u}_i. \quad (2.32)$$

We conclude that $\{\mathbf{u}_i\} \ i = 1, \dots, m-1$ are linearly dependent. Due to 2.30 and the hypothesis that the eigenvalues are distinct, at least one of the coefficients must be non-zero, and therefore we can solve for one of the remaining eigenvectors in terms of the others $m-2$. Repeating this procedure $(m-1)$ times, we arrive at the conclusion that $\mathbf{u}_1 = 0$, which is a contradiction since, as we have seen, the eigenvector equation always has a non-trivial solution for each distinct eigenvalue ♠

If for each eigenvalue there is more than one eigenvector, then these form a higher-dimensional invariant vector subspace (reducible). Within each of these subspaces, we can take a basis composed of eigenvectors. The previous lemma ensures that all these eigenvectors thus chosen, for all eigenvalues, form a large linearly independent set.

Exercise: Convince yourself that the set of eigenvectors with the same eigenvalue forms a vector subspace.

If a given operator A has all its eigenvalues distinct, then the corresponding eigenvectors are linearly independent and equal in number to the dimension of V , that is, they generate a basis of V^C . In that basis, the matrix representation of A is diagonal, that is, $A^j_i = \delta^j_i \lambda_i$. Each of its eigenvectors generates an irreducible invariant subspace, and together they generate V^C . In each of them, the operator A acts merely by multiplication by λ_i . Note that the λ_i are generally complex, and therefore such multiplication is actually a rotation plus a dilation. Note that, unlike the basis of Schur's triangulation lemma, this is not generally orthogonal with respect to any given inner product.⁹

Example: Let V be a vector space with $\dim(V) = 2$ and let A be given by $A(\mathbf{e}_1) = \mathbf{e}_2$, $A(\mathbf{e}_2) = -\mathbf{e}_1$, where $(\mathbf{e}_1, \mathbf{e}_2)$ are any two linearly independent vectors. If we interpret them as two orthonormal vectors, then A is a rotation by $\pi/2$ in the plane.

⁹Note, however, that it can be *declared* orthogonal by defining the inner product as $(\mathbf{u}, \mathbf{v}) = \sum_{i=1}^n \theta^i(\mathbf{u})\theta^i(\mathbf{v})$

Now let's calculate the determinant of $A - \lambda I$,

$$\begin{aligned}\det(A - \lambda I) &= \varepsilon((A - \lambda I)e_1, (A - \lambda I)e_2) / \varepsilon(e_1, e_2) \\ &= \varepsilon(e_2 - \lambda e_1, -e_1 - \lambda e_2) / \varepsilon(e_1, e_2) \\ &= 1 + \lambda^2.\end{aligned}\tag{2.33}$$

and therefore the eigenvalues are $\lambda_1 = \iota$, and $\lambda_2 = -\iota$. The eigenvectors are $u_1 = e_1 + \iota e_2$ and $u_2 = e_1 - \iota e_2 = \bar{u}_1$. We see then that the action of A in these subspaces is multiplication by $\pm \iota$ and that both invariant subspaces are genuinely complex. In this new basis, the space V is generated by all linear combinations of the form $zu_1 + \bar{z}u_2$, and the action of A is simply multiplication by ι of z .

If the multiplicity of any of the roots $\det(A - \lambda I) = 0$ is greater than one, there will be fewer eigenvalues than the dimension of the space, and therefore we will not be guaranteed to have enough eigenvectors to form a basis, as we can only guarantee the existence of one for each eigenvalue.

Example: Let V be the set of 2-tuples of real numbers with a generic element (a, b) and let A be given by $A(a, b) = (\lambda a + \epsilon b, \lambda b)$. Taking a basis, $e_1 = (1, 0)$, $e_2 = (0, 1)$, we see that its matrix representation is:

$$\begin{pmatrix} \lambda & \epsilon \\ 0 & \lambda \end{pmatrix}\tag{2.34}$$

We see that this operator has only one eigenvalue with multiplicity 2. But it has only one eigenvector, proportional to $e_1 = (1, 0)$. For future use, note that if we define $\Delta = A - \lambda I$, then $e_1 = \frac{1}{\epsilon} \Delta e_2$, and therefore, in the basis $\{\tilde{e}_1 = e_1, \tilde{e}_2 = \frac{1}{\epsilon} e_2\}$, the operator is represented by the matrix,

$$\begin{pmatrix} \lambda & 1 \\ 0 & \lambda \end{pmatrix}\tag{2.35}$$

We must therefore analyze what happens in these cases. To do this, let us define, given λ_i an eigenvalue of A , the following subspaces:

$$W_{\lambda_i p} = \{u \in V \mid (A - \lambda_i I)^p u = 0\}\tag{2.36}$$

Note that these are invariant spaces: $AW_{\lambda_i p} \subset W_{\lambda_i p}$. Moreover, $W_{\lambda_i p} \subset W_{\lambda_i p+1}$, and therefore, for a sufficiently large p ($p \leq n$), we will have that $W_{\lambda_i p} = W_{\lambda_i p+1}$, taking the minimum among the p 's where this occurs, we define $W_{\lambda_i} := W_{\lambda_i p}$. Note that if for some λ_i , $p = 1$, then the subspace W_{λ_i} is composed of eigenvectors. These are the maximum invariant spaces associated with the eigenvalue λ_i , indeed we have:

Lemma 2.7 *The only eigenvalue of A in W_{λ_i} is λ_i .*

Proof: Let λ be an eigenvalue of A in W_{λ_i} . Let's see that $\lambda = \lambda_i$. As we have already seen, there will be an eigenvector $\zeta \in W_{\lambda_i}$ with λ as the eigenvalue. Since it is in W_{λ_i} , there will be some $p \geq 1$ such that $(A - \lambda_i I)^p \zeta = 0$, but since it is an eigenvector, we have that $(\lambda - \lambda_i)^p \zeta = 0$, and therefore $\lambda = \lambda_i$ ♠

Now let's see that these subspaces are linearly independent and generate all of V^C . We will prove this theorem again in Chapter ??.

Theorem 2.4 (See Chapter ??, Theorem 5.3) *Given an operator $A : V \rightarrow V$, with eigenvectors $\{\lambda_i\}$, $i = 1 \dots m$, the space V^C admits a direct decomposition into invariant subspaces W_{λ_i} , where in each of them A has only λ_i as an eigenvalue.*

Proof:

The W_{λ_i} are independent. Let $v_1 + \dots + v_s = 0$, with $v_i \in W_{\lambda_i}$, then we must prove that each $v_i = 0$. Applying $(A - \lambda_2 I)^{p_2} \dots (A - \lambda_s I)^{p_s}$ to the previous sum, we get, $(A - \lambda_2 I)^{p_2} \dots (A - \lambda_s I)^{p_s} v_1 = 0$, but since λ_i , $i \neq 1$, is not an eigenvalue of A in W_{λ_1} , the operator $(A - \lambda_2 I)^{p_2} \dots (A - \lambda_s I)^{p_s}$ is invertible¹⁰ in that subspace, and therefore $v_1 = 0$. Continuing in this way, we see that all the v_i must be zero.

The W_{λ_i} generate all of V^C . Suppose by contradiction that this is not the case, and consider V^C/W , where W is the space generated by all the W_{λ_i} , that is, the space of all linear combinations of elements in the W_{λ_i} . The operator A acts on V^C/W [$A\{u\} = \{Au\}$], and therefore it has an eigenvalue-eigenvector pair there. This implies that for some element ζ of V^C in some equivalent class of V^C/W , we have:

$$A\zeta = \lambda\zeta + u_1 + \dots + u_s \quad (2.37)$$

where the u_i belong to each W_{λ_i} . Now suppose that $\lambda \neq \lambda_i \forall i = 1 \dots s$, then $A - \lambda I$ is invertible in each W_{λ_i} , and therefore there exist vectors $\zeta_i = (A - \lambda I)^{-1} u_i \in W_{\lambda_i}$. But then $\tilde{\zeta} := \zeta - \zeta_1 - \dots - \zeta_s$ is an eigenvector of A ! This is impossible since λ is not a root of the characteristic polynomial, nor is $\tilde{\zeta} = 0$, since belonging to ζ to V^C/W is not a linear combination of elements in the W_{λ_i} . We thus have a contradiction. Now suppose that $\lambda = \lambda_j$ for some $j \in \{1 \dots s\}$. We can still define the vectors $\zeta_i = (A - \lambda_j I)^{-1} u_i$ for all $i \neq j$, and $\tilde{\zeta}$, where we only subtract from ζ all the ζ_i with $i \neq j$, therefore we have that

$$(A - \lambda_j I)\tilde{\zeta} = u_j \quad (2.38)$$

But applying $(A - \lambda_j I)^{p_j}$ to this equation, with p_j the minimum value for which $W_{\lambda_j p_j} = W_{\lambda_j p_j + 1}$, we get that $\tilde{\zeta} \in W_{\lambda_j}$, and thus another contradiction, therefore it can only be that V^C/W is the trivial space, and the W_{λ_i} generate all of V^C ♠

¹⁰Note that $(A - \lambda_i I)^s|_{W_{\lambda_j}}$ is invertible if its determinant is non-zero. But $\det(A - \lambda_i I)^s = (\det(A - \lambda_i I))^s = (\lambda_j - \lambda_i)^{s \dim(W_{\lambda_j})} \neq 0$

We see that we only need to study each of these subspaces W_{λ_i} to find all their irreducible parts (from now on we will suppress the subscript i). But in these subspaces, the operator A acts very simply!

Indeed, let $\Delta_\lambda : W_\lambda \rightarrow W_\lambda$ be defined by $\Delta_\lambda := A|_{W_\lambda} - \lambda I|_{W_\lambda}$, then Δ has only 0 as an eigenvalue, and therefore it is **nilpotent**, that is, there exists an integer $m \leq n$ such that $\Delta_\lambda^m = 0$.

Lemma 2.8 *Let $\Delta : W \rightarrow W$ be such that its only eigenvalue is 0, then Δ is nilpotent.*

Proof: Let $W^p := \Delta^p[W]$, then we have that $W^p \subseteq W^q$ if $p \geq q$. Indeed, $W^p = \Delta^p[W] = \Delta^q[\Delta^{p-q}[W]] \subset \Delta^q[W]$. Since the dimension of W is finite, it must happen that for some integer p , we will have that $W^p = W^{p+1}$, we see then that Δ^p acts injectively on W^p and therefore cannot have 0 as an eigenvalue. But we have seen that every operator has some eigenvalue, and therefore we have a contradiction unless $W^p = \{0\}$. That is, $\Delta^p = 0$ ♠

Nilpotent operators have the important property of generating a basis of the space in which they act by repeatedly applying them to a smaller set of linearly independent vectors.

Lemma 2.9 *Let $\Delta : W \rightarrow W$ be nilpotent, then there exists a basis of W consisting of elements of the form:*

$$\{\{v_1, \Delta v_1, \dots, \Delta^{p_1} v_1\}, \dots, \{\{v_d, \Delta v_d, \dots, \Delta^{p_d} v_d\}\}$$

where p_i is such that $\Delta^{p_i+1} v_i = 0$

Note that if $n = \dim W$ then $n = \sum_{i=1}^d p_i$. Each of these sets formed by repeated applications of an operator is called a **cycle**. In this case, the basis is formed by the elements of d cycles. Note that cycles are not necessarily unique entities; indeed, if we have two cycles with the same number of elements, then any linear combination of them will also be a cycle. Note that each cycle contains only one eigenvector.

Example: Consider the matrix,

$$\Delta := \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix} \quad (2.39)$$

its powers are,

$$\Delta^2 := \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \quad \Delta^3 := \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \quad \Delta^4 := \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \quad (2.40)$$

In this case, we have a single cycle, corresponding to the eigenvector $e_1 = (1, 0, 0, 0)$, which is the vector whose span is $\Delta^3[\mathbb{R}^4]$, the cycle is,

$$\begin{aligned} e_4 &= (0, 0, 0, 1) \\ e_3 &= (0, 0, 1, 0) = \Delta e_4 \\ e_2 &= (0, 1, 0, 0) = \Delta e_3 = \Delta^2 e_4 \\ e_1 &= (1, 0, 0, 0) = \Delta e_2 = \Delta^2 e_3 = \Delta^3 e_4. \end{aligned}$$

Example: Consider the matrix,

$$\Delta := \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix} \quad (2.41)$$

its non-trivial powers are,

$$\Delta^2 := \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix} \quad \Delta^3 := \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

In this case, we have two cycles, corresponding to the two eigenvectors, e_1 and e_3 .

Proof: We will prove it by induction on the dimension of W . If $n = 1$, we take any vector to generate the basis, since in this case $\Delta = 0$. We now assume it is true for any dimension less than n . In particular, since Δ has a zero eigenvalue, $\dim(\ker \Delta) \geq 1$, and therefore we have that $W' (= \Delta(W))$ has a dimension less than n , say n' , and by the inductive hypothesis, a basis of the form

$$\{\{v'_1, \Delta v'_1, \dots, \Delta^{p'_1} v'_1\}, \dots, \{\{v'_{d'}, \Delta v'_{d'}, \dots, \Delta^{p'_{d'}} v'_{d'}\}\}.$$

To form a basis of W , we will add to these vectors d' vectors v_i such that $\Delta v_i = v'_i$, $i = 1, \dots, d'$. This can always be done since $v'_i \in W' = \Delta W$. We thus see that we have increased the set of vectors to

$$\{\{v_1, \Delta v_1, \dots, \Delta^{p'_1+1} v_1\}, \dots, \{\{v_{d'}, \Delta v_{d'}, \dots, \Delta^{p'_{d'}+1} v_{d'}\}\},$$

that is, we now have $r = \sum_{i=1}^{d'} (p'_i + 1) = n' + d'$ vectors. To obtain a basis, we must then increase this set with $n - n' - d'$ vectors. Note that this number is non-negative; indeed, $n - n' = \dim(\ker \Delta) \geq \dim(\ker \Delta \cap W') = d'$, and it is precisely the dimension of the subspace of $\ker \Delta$ that is not in W' . We then complete the proposed basis for W by incorporating into the already obtained set $n - n' - d'$ vectors $\{z_i\}, i = 1, \dots, n - n' - d'$ from the null space of Δ that are linearly independent among themselves and with the other elements of $\ker \Delta$ in W and that are also not in W' . We have thus obtained a set of $d = d' + n - n' - d' = n - n'$ cycles. Let's see that the set thus obtained is a basis. Since they are n in number, we only need to see that they are linearly independent. We must then prove that if we have constants $\{C_{i,j}\}, i = 1..d, j = 0..p'_i + 1$ such that

$$0 = \sum_{i=1}^d \sum_{j=0}^{p'_i+1} C_{ij} \Delta^{p_i} v_i \quad (2.42)$$

then $C_{ij} = 0$. Applying Δ to this relation, we get,

$$\begin{aligned} 0 &= \Delta \sum_{i=1}^d \sum_{j=0}^{p'_i+1} C_{ij} \Delta^{p_i} v_i \\ &= \sum_{i=1}^{d'} \sum_{j=0}^{p'_i} C_{ij} \Delta^{p'_i} v'_i, \end{aligned} \quad (2.43)$$

where we have used that $\Delta^{p'_i+1} v'_i = 0$. But this is the orthogonality relation of the basis of W' , and therefore we conclude that $C_{ij} = 0 \forall i \leq d', j \leq p'_i$. The initial relation is then reduced to

$$\begin{aligned} 0 &= \sum_{i=1}^d C_{i p'_i+1} \Delta^{p'_i+1} v_i \\ &= \sum_{i=1}^{d'} C_{i p'_i} \Delta^{p'_i} v'_i + \sum_{i=d'+1}^d C_{i1} z_i, \end{aligned} \quad (2.44)$$

but the members of the first summation are part of the basis of W' and therefore linearly independent among themselves, while those of the second are a set of elements outside W' chosen to be linearly independent among themselves and with those of the first summation, and therefore we conclude that all C_{ij} are zero ♠

Alternative Proof: Alternatively, the previous lemma can be proven constructively. Indeed, if $m + 1$ is the power for which Δ is nullified, we can take the space $W^m = \Delta^m[W]$ and a basis $\{v_i^m\}$ of it. Note that all elements of W^m are eigenvectors of A , and therefore the elements of the basis. Then we consider the space $W^{m-1} = \Delta^{m-1}[W]$. Note that $W^m = \Delta^m[W] = \Delta^{m-1}[\Delta[W]]$ and $\Delta[W] \subset W$, therefore $W^m \subset W^{m-1}$. Since $W^m = \Delta[W^{m-1}]$, for each vector v_i^m of the basis $\{v_i^m\}$ of

W^m there will be a vector v_i^{m-1} such that $\Delta v_i^{m-1} = v_i^m$. Since $W^m \subset W^{m-1}$, the set $\{v_i^m\} \cup \{v_i^{m-1}\}$ is contained in W^{m-1} . Note that $\dim W^m = \dim(\Delta[W^{m-1}]) \leq \dim(\ker \Delta \cap W^{m-1})$, since $W^m \subset W^{m-1}$ and all its elements belong to $\ker \Delta \cap W^{m-1}$.

Adding to the previous set a set $\{z_j\}$ of $\dim(\ker \Delta \cap W^{m-1}) - \dim(\ker \Delta \cap W^m)$ vectors from the null space of Δ in W^{m-1} , such that they are linearly independent among themselves and with the elements of the basis of W^m , we obtain a set of $\dim W^{m-1}$ vectors. Note that the mentioned choice of elements $\{z_j\}$ can be made since it is merely an extension of the basis $\{v_i^m\}$ of W^m to a basis of $\ker \Delta \cap W^{m-1}$. Now let's prove that they are linearly independent and therefore form a basis of W^{m-1} . To do this, we need to prove that if

$$0 = \sum_i C_i^m v_i^m + \sum_i C_i^{m-1} v_i^{m-1} + \sum_j C_j^z z_j \quad (2.45)$$

then each of the coefficients C_i^m , C_i^{m-1} , C_j must be zero. Multiplying the previous expression by Δ , we get,

$$0 = \sum_i C_i^{m-1} \Delta v_i^{m-1} = \sum_i C_i^{m-1} v_i^m, \quad (2.46)$$

but then the linear independence of the basis of W^m ensures that the $\{C_i^{m-1}\}$ are all zero. We then have,

$$0 = \sum_i C_i^m v_i^m + \sum_j C_j^z z_j. \quad (2.47)$$

But these vectors were chosen to be linearly independent among themselves and therefore all the coefficients in this sum must be zero. We thus see that the set $\{v_i^m\} \cup \{v_i^{m-1}\} \cup \{z_j\}$ forms a cyclic basis of W^{m-1} . Continuing with W^{m-2} and so on, we obtain a cyclic basis for all of W ♠

We thus see that the irreducible invariant subspaces of an operator are constituted by cycles within invariant subspaces associated with a given eigenvector. Each cycle contains a unique eigenvalue of the operator. We will denote the subspaces generated by these cycles (and usually also called cycles) by $C_{\lambda_i}^k$, where the lower index refers to the eigenvalue of the cycle and the upper index indexes the different cycles within each W_{λ_i} .

Exercise: Show that the obtained cycles are irreducible invariant subspaces of A .

2.3.3 Jordan Canonical Form

Definition: Let $A: V \rightarrow V$ be a linear operator. We will say that A is of Jordan type with eigenvalue λ if there exists a basis $\{u_i\}$ of V such that ¹¹

¹¹In the sense that $A(v) = \lambda \sum_{i=1}^n u_i \theta^i(v) + \sum_{i=2}^n u_i \theta^{i-1}(v) \quad \forall v \in V$

$$A = \lambda \sum_{i=1}^n u_i \theta^i + \sum_{i=2}^n u_i \theta^{i-1} \equiv \lambda I + \Delta \quad (2.48)$$

where $\{\theta^i\}$ is the co-basis of the basis $\{u_i\}$.

That is, in this basis, the components of A form a matrix A^j_i given by

$$A = \begin{pmatrix} \lambda & 1 & 0 & \cdots & 0 \\ 0 & \ddots & \ddots & \ddots & \vdots \\ & & \lambda & 1 & 0 \\ & 0 & & \lambda & 1 \\ & & & & \lambda \end{pmatrix} \quad (2.49)$$

Note that the matrix Δ is n -nilpotent, that is $\Delta^n = 0$.

Not every operator is of Jordan type –find one that is not– but it is clear that the restriction of any operator to one of its irreducible invariant subspaces (cycles) is. This can be seen by conveniently numbering the elements of the basis generated by the cycle.

Exercise: Find such an ordering of the elements of the basis.

Therefore, we can summarize the results found earlier in the following theorem about the matrix representations of any operator acting in a finite-dimensional space.

Theorem 2.5 de Jordan Let $A : V \rightarrow V$ be a linear operator acting in a complex vector space V . Then, there exists a unique decomposition into a direct sum¹² of V into subspaces $C_{\lambda_i}^k$, $V = C_{\lambda_1}^1 \oplus \cdots \oplus C_{\lambda_1}^{k_1} \oplus \cdots \oplus C_{\lambda_d}^1 \oplus \cdots \oplus C_{\lambda_d}^{k_d}$, $d \leq n$ such that

i) The $C_{\lambda_i}^k$ are invariant under the action of A , that is, $A C_{\lambda_i}^k \subseteq C_{\lambda_i}^k$

ii) The $C_{\lambda_i}^k$ are irreducible, that is, there are no invariant subspaces of $C_{\lambda_i}^k$ such that their sums are the entire $C_{\lambda_i}^k$.

iii) Due to property i), the operator A induces in each $C_{\lambda_i}^k$ an operator $A_i : C_{\lambda_i}^k \rightarrow C_{\lambda_i}^k$, which is of Jordan type with λ_i being one of the roots of the polynomial of degree n_i ,

$$\det(A_i - \lambda_i I) = 0. \quad (2.50)$$

This theorem tells us that given A , there exists a basis, generally complex, such that the matrix of its components has the form of square diagonal blocks of $n_i \times n_i$, where n_i is the dimension of the subspace $C_{\lambda_i}^k$, each with the form given in 2.49. This form of the matrix is called the **Jordan canonical form**.

¹²Recall that a vector space V is said to be the direct sum of two vector spaces W and Z , and we denote it as $V = W \oplus Z$ if each vector in V can be obtained in a unique way as the sum of an element in W and another in Z .

Exercise: Show that the roots, λ_i , that appear in the operators A_i are invariant under similarity transformations, that is, $\lambda_i(A) = \lambda_i(PAP^{-1})$.

Example: Let $A : C^3 \rightarrow C^3$, then $\det(A - \lambda I)$ is a polynomial of degree 3, and therefore has three roots. If these are distinct, there will be at least three invariant and irreducible subspaces of C^3 , but $\dim C^3 = 3$ and therefore each of them has $n_i = 1$. The Jordan canonical form is then,

If two of them coincide, we have two possibilities: either we have three subspaces, in which case the Jordan canonical form will be

$$\begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{pmatrix} \quad (2.51)$$

or, we have two subspaces, one necessarily of dimension 2, the Jordan canonical form will be

$$\begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{pmatrix} \quad (2.52)$$

If all three roots coincide, then there will be three possibilities,

$$\begin{pmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{pmatrix}, \quad \begin{pmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{pmatrix} \text{ and } \begin{pmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{pmatrix} \quad (2.53)$$

Example: We will now illustrate the case of coincident eigenvalues in two dimensions. This case is not generic, in the sense that any perturbation in the system –that is, any minimal change in the equations– separates the roots, making them distinct.

Let $A : C^2 \rightarrow C^2$. In this case, the characteristic polynomial $\det(A - \lambda I)$ has only two roots, which we will assume are coincident, $\lambda_1 = \lambda_2 = \lambda$. We find ourselves with two possibilities: either there are two linearly independent eigenvectors u_1 and u_2 , in which case $V = B_1 \oplus B_2$ and A is diagonalizable ($A = \text{diag}(\lambda, \lambda) = I\lambda$), or there is only one eigenvector \tilde{u}_1 . In this case, let \tilde{u}_2 be any vector linearly independent of \tilde{u}_1 , then $A\tilde{u}_2 = c^1\tilde{u}_1 + c^2\tilde{u}_2$ for some scalars c^1 and c^2 in C . Calculating the determinant of $A - \tilde{\lambda}I$ in this basis, we get $(\lambda - \tilde{\lambda})(c^2 - \tilde{\lambda})$, but λ is a double root and therefore $c^2 = \lambda$.

Reordering and rescaling the bases $u_1 = \tilde{u}_1$, $u_2 = c^1\tilde{u}_2$, we obtain

$$\begin{aligned} Au_1 &= \lambda u_1 \\ Au_2 &= \lambda u_2 + u_1, \end{aligned} \quad (2.54)$$

and therefore

$$A = \lambda(u_1 \oplus \theta^1 + u_2 \oplus \theta^2) + u_1 \oplus \theta^2, \quad (2.55)$$

where $\{\theta^i\}$ is the co-basis of the basis $\{u_i\}$.

Note that $(A - \lambda I)u_2 = u_1$ and $(A - \lambda I)u_1 = 0$, that is, $\Delta^2 = (A - \lambda I)^2 = 0$.

As we will see later in physical applications, the invariant subspaces have a clear physical meaning, they are called normal modes –one-dimensional case– and cycles –in other cases.

2.3.4 Similarity Relation

In physics applications, the following equivalence relation is common: [See box at the end of the chapter.] We will say that the operator A is similar to the operator B if there exists another operator P , invertible, such that

$$A = PBP^{-1}. \quad (2.56)$$

That is, if we *rotate* V with an invertible operator P and then apply A , we obtain the same action as if we first apply B and then *rotate* with P .

Exercise:

- Prove that similarity is an equivalence relation.
- Prove that the functions and maps defined above are the same in the different equivalent classes, that is,

$$\begin{aligned} \det(PAP^{-1}) &= \det(A) \\ \text{tr}(PAP^{-1}) &= \text{tr}(A) \\ e^{PAP^{-1}} &= Pe^AP^{-1}. \end{aligned} \quad (2.57)$$

2.4 Adjoint Operators

Let A be a linear operator between two vector spaces, $A : V \rightarrow W$, that is, $A \in \mathcal{L}(V, W)$. Since V and W have dual spaces, this operator naturally induces a linear operator from W' to V' , called its **dual**,

$$A'(\omega)(v) := \omega(A(v)) \quad \forall \omega \in W', \quad v \in V. \quad (2.58)$$

That is, the operator that when applied to an element $\omega \in W'$ gives us the element $A'(\omega)$ of V' which, when acting on $v \in V$, gives the number $\omega(A(v))$. See figure.

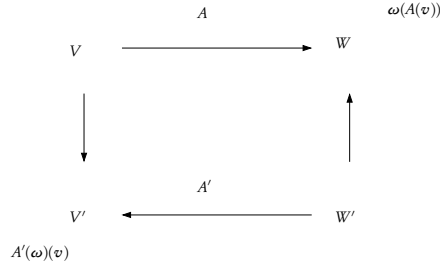


Figure 2.2: Diagram of the dual operator.

Note that this is a linear operator since,

$$\begin{aligned} A'(\alpha\omega + \sigma)(v) &= (\alpha\omega + \sigma)(A(v)) \\ &= \alpha\omega(A(v)) + \sigma(A(v)) \\ &= \alpha A'(\omega)(v) + A'(\sigma)(v). \end{aligned} \quad (2.59)$$

In the matrix representation, this operator is represented merely by the same matrix as the original, but now acting on the left, $A'(\omega)_i = \omega_j A^j_i$, that is, $A'^j_i = A^j_i$.

If there are norms defined in V and W and we define the norm of $A : V \rightarrow W$ in the usual way,

$$\|A\| := \sup_{\|v\|_V=1} \{\|A(v)\|_W\} \quad (2.60)$$

Then we see that

$$\begin{aligned}
\|A'\| &:= \sup_{\|\omega\|_{W'}=1} \{\|A'(\omega)\|_{V'}\} \\
&= \sup_{\|\omega\|_{W'}=1} \{ \sup_{\|v\|_V=1} \{|A'(\omega)(v)|\} \} \\
&= \sup_{\|\omega\|_{W'}=1} \{ \sup_{\|v\|_V=1} \{|\omega(A(v))|\} \} \\
&\leq \sup_{\|\omega\|_{W'}=1} \{ \sup_{\|v\|_V=1} \{\|\omega\|_{W'}\|(A(v))\|_W\} \} \\
&= \sup_{\|v\|_V=1} \{\|(A(v))\|_W\} \\
&= \|A\|.
\end{aligned} \tag{2.61}$$

Thus we see that if an operator is bounded, then its dual is also bounded. In fact, the equality of the norms can be proven, but this would require introducing new tools (in the most general case, the Hahn-Banach theorem) that we do not wish to incorporate in this text.

Let's see what the components of the dual of an operator are in terms of the components of the original operator. Let $\{e_i\}$, $\{\theta^i\}$, $i = 1, \dots, n$ be a basis and respectively a co-basis of V , and let $\{\hat{e}_i\}$, $\{\hat{\theta}^i\}$, $i = 1, \dots, m$ be a pair of basis and co-basis of W . We then have that the components of A with respect to these bases are: $A^i_j := \hat{\theta}^i(A(e_j))$, that is, $A(v) = \sum_{i=1}^m \sum_{j=1}^n A^i_j \hat{e}_i \theta^j(v)$.

Therefore, if v has components (v^1, v^2, \dots, v^n) in the basis $\{e_i\}$, $A(v)$ has components

$$(\sum_{i=1}^n A^1_i v^i, \sum_{i=1}^n A^2_i v^i, \dots, \sum_{i=1}^n A^m_i v^i)$$

in the basis $\{\hat{e}_i\}$

Now let's see the components of A' . By definition we have,

$$A'^i_j := A'(\hat{\theta}^i)(e_j) = \hat{\theta}^i(A(e_j)) = A^i_j.$$

That is, the same components, but now the matrix acts on the left on the components $(\omega_1, \omega_2, \dots, \omega_m)$ of an element ω of W' in the co-basis $\{\hat{\theta}^i\}$. The components of $A'(\omega)$ in the co-basis $\{\theta^i\}$ are,

$$(\sum_{i=1}^m A^i_1 \omega_i, \sum_{i=1}^m A^i_2 \omega_i, \dots, \sum_{i=1}^m A^i_n \omega_i).$$

A particularly interesting case of this construction is when $W = V$ and this is a space with an inner product, that is, a Hilbert space. In this case, the inner product gives us a canonical map between V and its dual V' :

$$\phi: V \rightarrow V', \quad \phi(v) := \langle v, \cdot \rangle. \tag{2.62}$$

This map is injective, and since V and V' have the same dimension, it is also surjective, and therefore invertible. That is, given $\omega \in V'$, there exists $v = \phi^{-1}(\omega) \in V$ such that $\langle v, \cdot \rangle = \omega$. Note then that $\phi^{-1} : V' \rightarrow V$ satisfies,

$$\langle \phi^{-1}(\omega), u \rangle = \langle v, u \rangle = \omega(u).$$

If $A : V \rightarrow V$, then $A' : V' \rightarrow V'$ can also be considered as an operator between V and V which we will call A^* .

With the help of this map, we define $A^* : V \rightarrow V$ given by: (See figure)

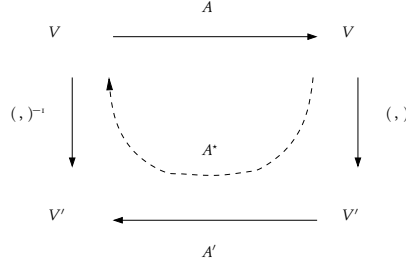


Figure 2.3: Diagram of the star operator.

$$A^*(v) := \phi^{-1}(A'(\phi(v))). \quad (2.63)$$

In terms of the inner product, this is:

$$\langle A^*(v), u \rangle = \langle \phi^{-1}(A'(\phi(v))), u \rangle = A'(\phi(v))(u) = \phi(v)(A(u)) = \langle v, A(u) \rangle. \quad (2.64)$$

In its matrix representation, this operator is,

$$A^{*j}_i = t_{il} A^l_k (t^{-1})^{kj}, \quad \text{real inner product} \quad (2.65)$$

$$A^{*j}_i = t_{il} \bar{A}^l_k (t^{-1})^{kj}, \quad \text{complex inner product} \quad (2.66)$$

where t_{li} is the representation of the inner product and $(t^{-1})^{jk}$ that of its inverse ($t_{il}(t^{-1})^{lk} = \delta^k_i$). If we choose a basis such that $t_{ik} = \delta_{ik}$, the Kronecker delta, the matrix representation of the adjoint is simply the transpose matrix, $A^{*j}_i = A^{\dagger j}_i = A^i_j$,

A particularly interesting subset of operators is those for which $A = A^*$. These operators are called **Hermitian** or **Self-adjoint**.¹³

Self-adjoint operators have important properties:

Lemma 2.10 *Let $M = \text{Span}\{u_1, u_2, \dots, u_m\}$, where $\{u_i\}$ are a set of eigenvalues of A , a self-adjoint operator. Then M and M^\perp are invariant spaces of A .*

¹³In the case of infinite dimension, these names do not coincide for some authors.

Proof: The first statement is clear and general, the second depends on the Hermiticity of A . Let $v \in M^\perp$ be any vector, let's see that $A(v) \in M^\perp$. Let $u \in M$ be arbitrary, then

$$\langle u, A(v) \rangle = \langle A(u), v \rangle = 0, \quad (2.67)$$

since $A(u) \in M$ if $u \in M$ ♠

This property has the following corollary:

Corollary 2.1 *Let $A : H \rightarrow H$ be self-adjoint. Then the eigenvectors of A form an orthonormal basis of H .*

Proof: $A : H \rightarrow H$ has at least one eigenvector, let's call it u_1 . Now consider its restriction to the space perpendicular to u_1 which we also denote by A since by the previous lemma this is an invariant space, $A : \{u_1\}^\perp \rightarrow \{u_1\}^\perp$. This operator also has an eigenvector, say u_2 and $u_1 \perp u_2$. Now consider the restriction of A to $\text{Span}\{u_1, u_2\}^\perp$, there we also have $A : \text{Span}\{u_1, u_2\}^\perp \rightarrow \text{Span}\{u_1, u_2\}^\perp$ and therefore an eigenvector of A , u_3 with $u_3 \perp u_1, u_3 \perp u_2$. Continuing in this way, we end up with $n = \dim H$ eigenvectors all orthogonal to each other ♠

This theorem has several extensions to the case where the vector space is of infinite dimension. Later in chapter ?? we will see one of them.

Note that in this basis A is diagonal and therefore we have

Corollary 2.2 *Every self-adjoint operator is diagonalizable*

Also note that if u is an eigenvector of A self-adjoint, with eigenvalue λ , then,

$$\bar{\lambda} \langle u, u \rangle = \langle A(u), u \rangle = \langle u, A(u) \rangle = \lambda \langle u, u \rangle \quad (2.68)$$

and therefore $\bar{\lambda} = \lambda$, that is,

Lemma 2.11 *The eigenvalues of a self-adjoint operator are real*

Let's see what the condition of Hermiticity means in terms of the components of the operator in an orthonormal basis. Let A be a self-adjoint operator and let $\{e_i\}, i = 1, \dots, n$ be an orthonormal basis of the space where it acts. We have that $\langle A(e_i), e_j \rangle = \langle e_i, A(e_j) \rangle$ and therefore, noting that $I = \sum_{i=1}^n e_i \theta^i$, we obtain,

$$\begin{aligned} 0 &= \left\langle \sum_{k=1}^n e_k \theta^k (A(e_i)), e_j \right\rangle - \left\langle e_i, \sum_{l=1}^n e_l \theta^l (A(e_j)) \right\rangle \\ &= \sum_{k=1}^n \bar{A}_i^k \langle e_k, e_j \rangle - \sum_{l=1}^n A_j^l \langle e_i, e_l \rangle \\ &= \sum_{k=1}^n \bar{A}_i^k \delta_{kj} - \sum_{l=1}^n A_j^l \delta_{li} \\ &= \bar{A}_j^i - A_j^i \end{aligned} \quad (2.69)$$

from which we conclude that

$$\bar{A}^j_i = A^i_j \quad (2.70)$$

that is, the transpose matrix is the complex conjugate of the original. In the case of real matrices, we see that the condition is that in this basis the matrix is equal

2.5 Adjoint Operators

Let A be a linear operator between two vector spaces, $A : V \rightarrow W$, that is, $A \in \mathcal{L}(V, W)$. Since V and W have dual spaces, this operator naturally induces a linear operator from W' to V' , called its **dual**,

$$A'(\omega)(v) := \omega(A(v)) \quad \forall \omega \in W', v \in V. \quad (2.71)$$

That is, the operator that when applied to an element $\omega \in W'$ gives us the element $A'(\omega)$ of V' which, when acting on $v \in V$, gives the number $\omega(A(v))$. See figure.

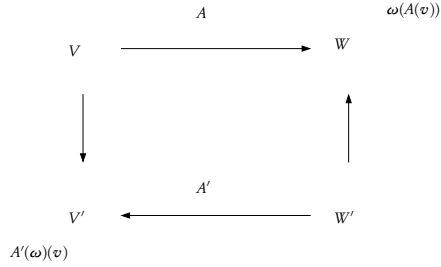


Figure 2.4: Diagram of the dual operator.

Note that this is a linear operator since,

$$\begin{aligned} A'(\alpha\omega + \sigma)(v) &= (\alpha\omega + \sigma)(A(v)) \\ &= \alpha\omega(A(v)) + \sigma(A(v)) \\ &= \alpha A'(\omega)(v) + A'(\sigma)(v). \end{aligned} \quad (2.72)$$

In the matrix representation, this operator is represented merely by the same matrix as the original, but now acting on the left, $A'(\omega)_i = \omega_j A^j_i$, that is, $A'^j_i = A^j_i$.

If there are norms defined in V and W and we define the norm of $A : V \rightarrow W$ in the usual way,

$$\|A\| := \sup_{\|v\|_V=1} \{\|A(v)\|_W\} \quad (2.73)$$

Then we see that

$$\begin{aligned}
\|A'\| &:= \sup_{\|\omega\|_{W'}=1} \{\|A'(\omega)\|_{V'}\} \\
&= \sup_{\|\omega\|_{W'}=1} \{ \sup_{\|v\|_V=1} \{|A'(\omega)(v)|\} \} \\
&= \sup_{\|\omega\|_{W'}=1} \{ \sup_{\|v\|_V=1} \{|\omega(A(v))|\} \} \\
&\leq \sup_{\|\omega\|_{W'}=1} \{ \sup_{\|v\|_V=1} \{\|\omega\|_{W'}\|(A(v))\|_W\} \} \\
&= \sup_{\|v\|_V=1} \{\|(A(v))\|_W\} \\
&= \|A\|.
\end{aligned} \tag{2.74}$$

Thus we see that if an operator is bounded, then its dual is also bounded. In fact, the equality of the norms can be proven, but this would require introducing new tools (in the most general case, the Hahn-Banach theorem) that we do not wish to incorporate in this text.

Let's see what the components of the dual of an operator are in terms of the components of the original operator. Let $\{e_i\}$, $\{\theta^i\}$, $i = 1, \dots, n$ be a basis and respectively a co-basis of V , and let $\{\hat{e}_i\}$, $\{\hat{\theta}^i\}$, $i = 1, \dots, m$ be a pair of basis and co-basis of W . We then have that the components of A with respect to these bases are: $A^i_j := \hat{\theta}^i(A(e_j))$, that is, $A(v) = \sum_{i=1}^m \sum_{j=1}^n A^i_j \hat{e}_i \theta^j(v)$.

Therefore, if v has components (v^1, v^2, \dots, v^n) in the basis $\{e_i\}$, $A(v)$ has components

$$(\sum_{i=1}^n A^1_i v^i, \sum_{i=1}^n A^2_i v^i, \dots, \sum_{i=1}^n A^m_i v^i)$$

in the basis $\{\hat{e}_i\}$

Now let's see the components of A' . By definition we have,

$$A'^i_j := A'(\hat{\theta}^i)(e_j) = \hat{\theta}^i(A(e_j)) = A^i_j.$$

That is, the same components, but now the matrix acts on the left on the components $(\omega_1, \omega_2, \dots, \omega_m)$ of an element ω of W' in the co-basis $\{\hat{\theta}^i\}$. The components of $A'(\omega)$ in the co-basis $\{\theta^i\}$ are,

$$(\sum_{i=1}^m A^i_1 \omega_i, \sum_{i=1}^m A^i_2 \omega_i, \dots, \sum_{i=1}^m A^i_n \omega_i).$$

A particularly interesting case of this construction is when $W = V$ and this is a space with an inner product, that is, a Hilbert space. In this case, the inner product gives us a canonical map between V and its dual V' :

$$\phi: V \rightarrow V', \quad \phi(v) := \langle v, \cdot \rangle. \tag{2.75}$$

This map is injective, and since V and V' have the same dimension, it is also surjective, and therefore invertible. That is, given $\omega \in V'$, there exists $v = \phi^{-1}(\omega) \in V$ such that $\langle v, \cdot \rangle = \omega$. Note then that $\phi^{-1} : V' \rightarrow V$ satisfies,

$$\langle \phi^{-1}(\omega), u \rangle = \langle v, u \rangle = \omega(u).$$

If $A : V \rightarrow V$, then $A' : V' \rightarrow V'$ can also be considered as an operator between V and V which we will call A^* .

With the help of this map, we define $A^* : V \rightarrow V$ given by: (See figure)

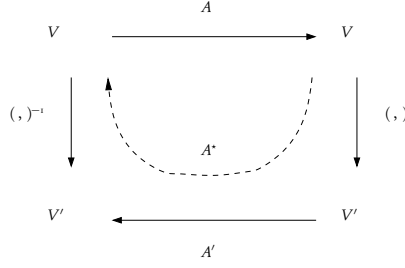


Figure 2.5: Diagram of the star operator.

$$A^*(v) := \phi^{-1}(A'(\phi(v))). \quad (2.76)$$

In terms of the inner product, this is:

$$\langle A^*(v), u \rangle = \langle \phi^{-1}(A'(\phi(v))), u \rangle = A'(\phi(v))(u) = \phi(v)(A(u)) = \langle v, A(u) \rangle. \quad (2.77)$$

In its matrix representation, this operator is,

$$A^{*j}_i = t_{il} A^l_k (t^{-1})^{kj}, \quad \text{real inner product} \quad (2.78)$$

$$A^{*j}_i = t_{il} \bar{A}^l_k (t^{-1})^{kj}, \quad \text{complex inner product} \quad (2.79)$$

where t_{li} is the representation of the inner product and $(t^{-1})^{jk}$ that of its inverse ($t_{il}(t^{-1})^{lk} = \delta^k_i$). If we choose a basis such that $t_{ik} = \delta_{ik}$, the Kronecker delta, the matrix representation of the adjoint is simply the transpose matrix, $A^{*j}_i = A^{\dagger j}_i = A^i_j$,

A particularly interesting subset of operators is those for which $A = A^*$. These operators are called **Hermitian** or **Self-adjoint**.¹⁴

Self-adjoint operators have important properties:

Lemma 2.12 *Let $M = \text{Span}\{u_1, u_2, \dots, u_m\}$, where $\{u_i\}$ are a set of eigenvalues of A , a self-adjoint operator. Then M and M^\perp are invariant spaces of A .*

¹⁴In the case of infinite dimension, these names do not coincide for some authors.

Proof: The first statement is clear and general, the second depends on the Hermiticity of A . Let $v \in M^\perp$ be any vector, let's see that $A(v) \in M^\perp$. Let $u \in M$ be arbitrary, then

$$\langle u, A(v) \rangle = \langle A(u), v \rangle = 0, \quad (2.80)$$

since $A(u) \in M$ if $u \in M$ ♠

This property has the following corollary:

Corollary 2.3 *Let $A : H \rightarrow H$ be self-adjoint. Then the eigenvectors of A form an orthonormal basis of H .*

Proof: $A : H \rightarrow H$ has at least one eigenvector, let's call it u_1 . Now consider its restriction to the space perpendicular to u_1 which we also denote by A since by the previous lemma this is an invariant space, $A : \{u_1\}^\perp \rightarrow \{u_1\}^\perp$. This operator also has an eigenvector, say u_2 and $u_1 \perp u_2$. Now consider the restriction of A to $\text{Span}\{u_1, u_2\}^\perp$, there we also have $A : \text{Span}\{u_1, u_2\}^\perp \rightarrow \text{Span}\{u_1, u_2\}^\perp$ and therefore an eigenvector of A , u_3 with $u_3 \perp u_1$, $u_3 \perp u_2$. Continuing in this way, we end up with $n = \dim H$ eigenvectors all orthogonal to each other ♠

This theorem has several extensions to the case where the vector space is of infinite dimension. Later in chapter ?? we will see one of them.

Note that in this basis A is diagonal and therefore we have

Corollary 2.4 *Every self-adjoint operator is diagonalizable*

Also note that if u is an eigenvector of A self-adjoint, with eigenvalue λ , then,

$$\bar{\lambda} \langle u, u \rangle = \langle A(u), u \rangle = \langle u, A(u) \rangle = \lambda \langle u, u \rangle \quad (2.81)$$

and therefore $\bar{\lambda} = \lambda$, that is,

Lemma 2.13 *The eigenvalues of a self-adjoint operator are real*

Let's see what the condition of Hermiticity means in terms of the components of the operator in an orthonormal basis. Let A be a self-adjoint operator and let $\{e_i\}$, $i = 1, \dots, n$ be an orthonormal basis of the space where it acts. We have that $\langle A(e_i), e_j \rangle = \langle e_i, A(e_j) \rangle$ and therefore, noting that $I = \sum_{i=1}^n e_i \theta^i$, we obtain,

$$\begin{aligned} 0 &= \left\langle \sum_{k=1}^n e_k \theta^k (A(e_i)), e_j \right\rangle - \left\langle e_i, \sum_{l=1}^n e_l \theta^l (A(e_j)) \right\rangle \\ &= \sum_{k=1}^n \bar{A}_i^k \langle e_k, e_j \rangle - \sum_{l=1}^n A_j^l \langle e_i, e_l \rangle \\ &= \sum_{k=1}^n \bar{A}_i^k \delta_{kj} - \sum_{l=1}^n A_j^l \delta_{li} \\ &= \bar{A}_j^i - A_j^i \end{aligned} \quad (2.82)$$

from which we conclude that

$$\bar{A}^i{}_i = A^i{}_i \quad (2.83)$$

so the transpose matrix is the complex conjugate of the original. In the case of real matrices, we see that the condition is that in that basis the matrix is equal to its transpose, which is usually denoted by saying that the matrix is symmetric.

An interesting property of self-adjoint operators is that their norm is equal to the supremum of the magnitudes of their eigenvalues. Since the calculation demonstrating this will be used later in the course, we provide a demonstration of this fact below.

Lemma 2.14 *If A is self-adjoint then $\|A\| = \sup\{|\lambda_i|\}$.*

Proof: Let $F(u) := \langle A(u), A(u) \rangle$ defined on the sphere $\|u\| = 1$. Since this set is compact (here we are using the fact that the space is finite-dimensional) it has a maximum which we will denote by u_o . Note then that $F(u_o) := \|A\|^2$. Since $F(u)$ is differentiable on the sphere, it must satisfy

$$\frac{d}{d\lambda} F(u_o + \lambda \delta u)|_{\lambda=0} = 0, \quad (2.84)$$

along any curve tangent to the sphere at the point u_o , that is, for all δu such that $\langle u_o, \delta u \rangle = 0$. See figure.

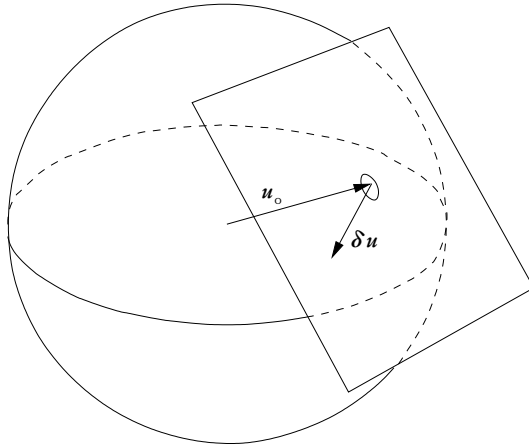


Figure 2.6: Normal and tangent vectors to the sphere.

But

$$\begin{aligned}
\frac{d}{d\lambda} F(u_o + \lambda \delta u)|_{\lambda=0} &= \frac{d}{d\lambda} \langle A(u_o + \lambda \delta u), A(u_o + \lambda \delta u) \rangle|_{\lambda=0} \\
&= \langle A(\delta u), A(u_o) \rangle + \langle A(u_o), A(\delta u) \rangle \\
&= 2\Re \langle A^* A u_o, \delta u \rangle \\
&= 0 \quad \forall \delta u, \quad \langle u_o, \delta u \rangle = 0
\end{aligned} \tag{2.85}$$

Since δu is arbitrary in $\{u_o\}^\perp$ this simply implies

$$A^* A(u_o) \in \{\delta u\}^\perp = \{u_o\}^{\perp\perp} = \{u_o\}. \tag{2.86}$$

and therefore there will exist $\alpha \in \mathbb{C}$ such that

$$A^* A(u_o) = \alpha u_o. \tag{2.87}$$

Taking the inner product with u_o we get,

$$\begin{aligned}
\langle A^* A(u_o), u_o \rangle &= \bar{\alpha} \langle u_o, u_o \rangle \\
&= \langle A(u_o), A(u_o) \rangle \\
&= \|A\|^2
\end{aligned} \tag{2.88}$$

and therefore we have that $\alpha = \bar{\alpha} = \|A\|^2$.

Now let $v := Au_o - \|A\|u_o$, then, now using that A is self-adjoint, we have,

$$\begin{aligned}
A(v) &= AA(u_o) - \|A\|A(u_o) \\
&= A^* A(u_o) - \|A\|A(u_o) \\
&= \|A\|^2 u_o - \|A\|A(u_o) \\
&= \|A\|(\|A\|u_o - A(u_o)) \\
&= -\|A\|v,
\end{aligned} \tag{2.89}$$

therefore either v is an eigenvector of A , with eigenvalue $\lambda = -\|A\|$, or $v = 0$, in which case u_o is an eigenvector of A with eigenvalue $\lambda = \|A\|$ ♠

2.6 Unitary Operators

Another subclass of linear operators that appears very often in physics when there is a privileged inner product is that of **unitary operators**, that is, those such that their action preserves the inner product,

$$\langle U(u), U(v) \rangle = \langle u, v \rangle, \quad \forall u, v \in H. \tag{2.90}$$

The most typical case of a unitary operator is a transformation that sends one orthonormal basis to another. Usual examples are rotations in \mathbb{R}^n .

We also observe that $\|U\| = \sup_{\|v\|=1} \{\|U(v)\|\} = 1$.

Note that

$$\langle U(u), U(v) \rangle = \langle U^* U(u), v \rangle = \langle u, v \rangle, \quad \forall u, v \in H, \quad (2.91)$$

that is,

$$U^* U = I \quad (2.92)$$

and therefore,

$$U^{-1} = U^*. \quad (2.93)$$

Let's see what the eigenvalues of a unitary operator U are. Let v_1 be an eigenvector of U (we know it has at least one), then,

$$\begin{aligned} \langle U(v_1), U(v_1) \rangle &= \lambda_1 \bar{\lambda}_1 \langle v_1, v_1 \rangle \\ &= \langle v_1, v_1 \rangle \end{aligned} \quad (2.94)$$

and therefore $\lambda_1 = e^{i\theta_1}$ for some angle θ_1 .

If the operator U represents a non-trivial rotation in \mathbb{R}^3 , then, given that we have an odd number of eigenvalues, there will be one that is real, the other two complex conjugates of each other. The eigenvector corresponding to the real eigenvalue defines the axis that remains fixed in that rotation. If we have more than one eigenvalue, then their corresponding eigenvectors are orthogonal, indeed, let v_1 and v_2 be two eigenvectors then

$$\begin{aligned} \langle U(v_1), U(v_2) \rangle &= \bar{\lambda}_1 \lambda_2 \langle v_1, v_2 \rangle \\ &= \langle v_1, v_2 \rangle \end{aligned} \quad (2.95)$$

and therefore if $\lambda_1 \neq \lambda_2$ we must have $\langle v_1, v_2 \rangle = 0$.

Exercise: Show that if A is a self-adjoint operator, then $U := e^{iA}$ is a unitary operator.

Equivalence Relations.

Definition: An **equivalence relation**, \approx , between elements of a set X is a relation that satisfies the following conditions:

1. Reflexive: If $x \in X$, then $x \approx x$.
2. Symmetric: If $x, x' \in X$ and $x \approx x'$, then $x' \approx x$.
3. Transitive: If $x, x', x'' \in X$, $x \approx x'$ and $x' \approx x''$, then $x \approx x''$.

Note that the first property ensures that each element of X satisfies an equivalence relation with some element of X , in this case with itself. Equivalence relations often appear in physics, essentially when we use a mathematical entity to describe a physical process that has superfluous parts with respect to this process and therefore we would like to ignore them. This is achieved by declaring two entities that describe the same phenomenon as equivalent entities.

Example: Let X be the set of real numbers and let $x \approx y$ if and only if there exists an integer n such that $x = n + y$, this is clearly an equivalence relation. This is used when we are interested in describing something using the straight line but that in reality should be described using a circle of unit circumference.

Given an equivalence relation in a set, we can group the elements of this set into **equivalence classes** of elements, that is, into subsets where all their elements are equivalent to each other and there is no element outside this subset that is equivalent to any of the elements of the subset. (If X is the set, $Y \subset X$ is one of its equivalence classes, and if $y \in Y$, then $y \approx y'$ if and only if $y' \in Y$.)

The fundamental property of equivalence relations is the following.

Theorem 2.6 *An equivalence relation in a set X allows regrouping its elements into equivalence classes such that each element of X is in one and only one of these.*

Proof: Let $x \in X$ and Y be the subset of all elements of X equivalent to x . Let's see that this subset is an equivalence class. Let y and y' be two elements of Y , that is, $y \approx x$ and $y' \approx x$, but by the transitivity property, $y \approx y'$. If $y \in Y$ and $z \notin Y$, then $y \not\approx z$, because otherwise z would be equivalent to x and therefore would be in Y . Finally, note that by reflexivity, x is also in Y . It only remains to see that if y is in Y and also in another equivalence class, say Z , then $Y = Z$. Since $y \in Y$, then y is equivalent to every element of Y , and since $y \in Z$, then y is equivalent to every element of Z , but by transitivity, every element of Y is equivalent to every element of Z , but since these are equivalence classes and therefore each contains all its equivalent elements, both must coincide.

Exercise: What are the equivalence classes of the previous examples?

2.7 Problems

Problem 2.2 Let the operator $A : V \rightarrow V$ where $\dim V = n$, such that $Ax = \lambda x$. Calculate $\det A$ and $\text{tr} A$.

Problem 2.3 Let $V = \mathbb{R}^3$ and x be any non-zero vector. Find geometrically and analytically the quotient space V/W_x , where W_x is the space generated by x . Take another vector, x' , linearly independent of the first and now calculate $V/W_{(x,x')}$.

Problem 2.4 The norm on operators is defined as:

$$\|A\|_{\mathcal{L}} = \max_{\|x\|_V=1} \|A(x)\|_V. \quad (2.96)$$

Find the norms of the following operators, given by their matrix representation with respect to a basis and where the norm in the vector space is the Euclidean norm with respect to that basis.

a)

$$\begin{pmatrix} 3 & 5 \\ 4 & 1 \end{pmatrix} \quad (2.97)$$

b)

$$\begin{pmatrix} 3 & 5 & 2 \\ 4 & 1 & 7 \\ 8 & 3 & 2 \end{pmatrix} \quad (2.98)$$

Problem 2.5 Let V be any vector space and let $\|\cdot\|$ be a Euclidean norm in that space. The Hilbert-Schmidt norm of an operator is defined as:

$$\|A\|_{HS}^2 = \sum_{i,j=1}^n |A^j_i|^2. \quad (2.99)$$

where the basis used has been orthonormal with respect to the Euclidean norm.

a) Show that this is a norm.

b) Show that $\|A\|_{\mathcal{L}} \leq \|A\|_{HS}$.

c) Show that $\sum_{j=1}^n |A^j_k|^2 \leq \|A\|_{\mathcal{L}}^2$ for each k . Therefore $\|A\|_{HS}^2 \leq n\|A\|_{\mathcal{L}}^2$, and the two norms are equivalent.

Hint: use that $\theta^j(A(u)) = \theta^j(A)(u)$ and then that $|\theta(u)| \leq \|\theta\| \|u\|$.

Problem 2.6 Calculate the eigenvalues and eigenvectors of the following matrices:

a)

$$\begin{pmatrix} 3 & 6 \\ 4 & 1 \end{pmatrix} \quad (2.100)$$

b)

$$\begin{pmatrix} 3 & 6 \\ 0 & 1 \end{pmatrix} \quad (2.101)$$

c)

$$\begin{pmatrix} 2 & 4 & 2 \\ 4 & 1 & 0 \\ 3 & 3 & 1 \end{pmatrix} \quad (2.102)$$

Problem 2.7 Bring the following matrices to upper triangular form: Note: From the transformation of the bases.

a)

$$\begin{pmatrix} 3 & 4 \\ 2 & 1 \end{pmatrix} \quad (2.103)$$

b)

$$\begin{pmatrix} 2 & 4 & 2 \\ 4 & 1 & 0 \\ 3 & 3 & 1 \end{pmatrix} \quad (2.104)$$

c)

$$\begin{pmatrix} 1 & 4 & 3 \\ 4 & 4 & 0 \\ 3 & 3 & 1 \end{pmatrix} \quad (2.105)$$

Problem 2.8 Show again that $\det e^{\mathbf{A}} = e^{\text{tr} \mathbf{A}}$. Hint: express the matrix representation of \mathbf{A} in a basis where it has the Jordan canonical form. Alternatively, use a basis where \mathbf{A} is upper triangular and see that the product of upper triangular matrices gives an upper triangular matrix and therefore the exponential of an upper triangular matrix is also upper triangular.

Bibliography notes: This chapter is based on the following books: [1], [3], [5] and [18]. The following are also of interest, [11] and [12]. Linear algebra is one of the largest and most prolific areas of mathematics, especially when dealing with infinite-dimensional spaces, which is usually called real analysis and operator theory. In my personal experience, most problems end up reducing to an algebraic problem and one feels that progress has been made when that problem can be solved.

3.1 Manifolds

There are several reasons that justify the study of the concept of a manifold, or more generally of differential geometry, by physicists. One is that manifolds naturally appear in physics and therefore we cannot avoid them. Only in elementary courses can they be circumvented through vector calculus in \mathbb{R}^n . Thus, for example, to study the movement of a particle restricted to move on a sphere, we imagine the latter embedded in \mathbb{R}^3 and use the natural coordinates of \mathbb{R}^3 to describe its movements.

The second reason is that the concept of a manifold is of great conceptual utility, since, for example, in the case of a particle moving in \mathbb{R}^3 , it allows us to clearly distinguish between the position of a particle and its velocity vector as mathematical entities of different nature. This fact is masked in \mathbb{R}^3 since this special type of manifold has the structure of a vector space.

Since the time of Galileo, we know that the language of physics is mathematics. Like any language, its utility goes beyond its daily use to understand each other and work on our ideas. Language allows for a synthesis of concepts and knowledge that encapsulates an entire area of knowledge into a smaller number of concepts. This allows future generations to understand an immense amount of knowledge that for previous generations were disparate aspects of reality as particular aspects of the same trunk of knowledge. The clearest example of this is the theories of the standard model of particles, which unify under the same phenomenon what we previously understood as distinct properties of matter. In particular, these theories naturally and deterministically incorporate elements of geometry, such as fiber bundles and connections, symmetries, etc.

A manifold is a generalization of Euclidean spaces \mathbb{R}^n in which one preserves the concept of continuity, that is, its topology in the local sense but discards its character as a vector space. A manifold of dimension n is, in imprecise terms, a set of points that locally is like \mathbb{R}^n , but not necessarily in its global form.

An example of a two-dimensional manifold is the sphere, S^2 . If we look at a sufficiently small neighborhood, U_p , of any point of S^2 , we see that it is similar to a neighborhood of the plane, \mathbb{R}^2 , in the sense that we can define a continuous and invertible map between both neighborhoods. Globally, the plane and the sphere are

topologically distinct, as there is no continuous and invertible map between them. [See figure 3.1.]

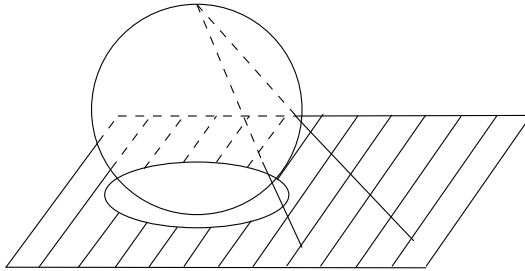


Figure 3.1: An atlas of the sphere.

As we said before, one might object to the need, in the previous example, to introduce the concept of a manifold, since one could consider S^2 as the subset of \mathbb{R}^3 such that $x_1^2 + x_2^2 + x_3^2 = 1$. The answer to this objection is that in physics one must follow the rule of economy of concepts and objects and discard everything that is not fundamental to the description of a phenomenon: if we want to describe things happening on the sphere, why do we need a space of more dimensions? This rule of economy forces us to refine concepts and discard everything superfluous. It is in this way that we advance in our maturation as physicists. It is the way we truly penetrate the mysteries of nature.

Note that the first way of working with the sphere is what we normally do when we seek to locate points and paths on the globe. In fact, we use flat maps, formerly called charts, to describe what happens in our cities and countries. When we want to have a collection of maps that cover the entire globe, we acquire an atlas, that is, a set of maps that cover the entire globe and have sectors in common between one and another. Some of these sectors are internal, for example when they describe a city within a country (on the map that covers that country), or on the edges when we go from one sheet to another. Only when we want to see the global structure of the globe, for example if we want to take a plane trip covering a large part of the globe, do we use a small version of the planet as implanted in \mathbb{R}^3 .

We now give a series of definitions to finally arrive at the definition of an n -dimensional manifold.

Definition: Let M be a set. A **chart of M** is a pair (U, φ) where U is a subset of M and φ an injective map between U and \mathbb{R}^n , such that its image, $\varphi[U]$ is open in \mathbb{R}^n .

Definition: An **atlas of M** is a collection of charts $\{(U_i, \varphi_i)\}$ satisfying the following conditions: [See figure 3.2.]

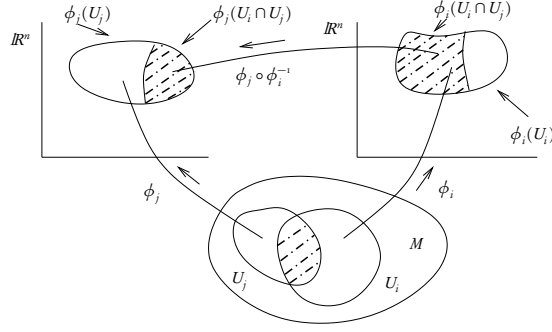


Figure 3.2: Relationship between charts.

1. The U_i cover M , ($M = \bigcup_i U_i$).
2. If two charts overlap then $\varphi_i(U_i \cap U_j)$ is also an open set in \mathbb{R}^n .
3. The map $\varphi_j \circ \varphi_i^{-1} : \varphi_i[U_i \cap U_j] \rightarrow \varphi_j[U_i \cap U_j]$ is continuous, injective, and surjective.

Condition 1 gives us a notion of *closeness* in M induced from the analogous notion in \mathbb{R}^n . Indeed, we can say that a sequence of points $\{p_k\}$ in M converges to p in U_i if there exists k_0 such that $\forall k > k_0$, $p_k \in U_i$ and the sequence $\{\varphi_i(p_k)\}$ converges to $\varphi_i(p)$. Another way to see this is that if after this construction of a manifold we impose that the maps φ_i are continuous, then we induce a unique topology on M . [See figure 3.3.]

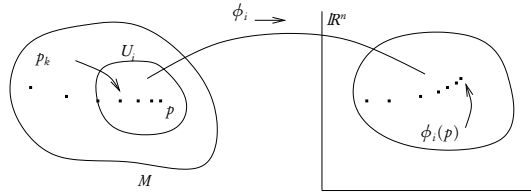


Figure 3.3: Sequences in M .

Condition 2 simply ensures that this notion is consistent. If $p \in U_i \cap U_j$ then the fact that the sequence converges is independent of whether we use the chart (U_i, φ_i) or (U_j, φ_j) .

Condition 3 allows us to encode in the maps $\varphi_j \circ \varphi_i^{-1}$ from \mathbb{R}^n to \mathbb{R}^n the global topological information necessary to distinguish, for example, whether M is a sphere,

a plane, or a torus. Therein lies, for example, the information that there is no continuous and invertible map between S^2 and \mathbb{R}^2 . But also, if we require these maps to be differentiable, it is what will allow us to formulate differential calculus on M . Indeed, note that in condition 3 we speak of the continuity of the map $\varphi_j \circ \varphi_i^{-1}$, which is well-defined because it is a map between \mathbb{R}^n and \mathbb{R}^n . Similarly, we can speak of the differentiability of these maps.

We will say that **an atlas** $\{(U_i, \varphi_i)\}$ is C^p if the maps $\varphi_j \circ \varphi_i^{-1}$ are p -times differentiable and their p -th derivative is continuous.

One might be tempted to define the manifold M as the pair consisting of the set M and an atlas $\{(U_i, \varphi_i)\}$, but this would lead us to consider as different manifolds, for example, the plane with an atlas given by the chart $(\mathbb{R}^2, (x, y) \rightarrow (x, y))$ and the plane with an atlas given by the chart $(\mathbb{R}^2, (x, y) \rightarrow (x, -y))$.

To remedy this inconvenience, we introduce the concept of equivalence between atlases.

Definition: We will say that **two atlases are equivalent** if their union is also an atlas.

Exercise: Prove that this is indeed an equivalence relation \approx , that is, it satisfies:

- i) $A \approx A$
- ii) $A \approx B \implies B \approx A$
- iii) $A \approx B, B \approx C \implies A \approx C$.

With this equivalence relation, we can divide the set of atlases of M into different **equivalent classes**. [Remember that each equivalent class is a set where all its elements are equivalent to each other and such that there is no element equivalent to these that is not in it.]

Definition: We will call **manifold M of dimension n and differentiability p** the pair consisting of the set M and an equivalent class of atlases, $\{\varphi_i : U_i \rightarrow \mathbb{R}^n\}$, in C^p .

It can be shown that to uniquely characterize the manifold M it is sufficient to give the set M and an atlas. If we have two atlases of M , then either they are equivalent and thus represent the same manifold, or they are not and then represent different manifolds.

The definition of a manifold that we have introduced is still too general for usual physical applications, in the sense that the allowed topologies can still be pathological from the point of view of physics. Therefore, in this course, we will impose an extra condition on manifolds. We will assume that they are **separable or Hausdorff**. That is, if p and $q \in M$, distinct, then: either they belong to the domain of the same chart U_i (in which case there exist neighborhoods W_p of $\varphi_i(p)$ and W_q of $\varphi_i(q)$ such that $\varphi_i^{-1}(W_p) \cap \varphi_i^{-1}(W_q) = \emptyset$, that is, the points have disjoint neighborhoods) or there exist U_i and U_j with $p \in U_i, q \in U_j$ and $U_i \cap U_j = \emptyset$, which also implies that they have disjoint neighborhoods. This is a property in the topology of M that essentially says we can separate points of M . An example of a non-Hausdorff manifold is the following.

Example: M , as a set, consists of three intervals of the line $I_1 = (-\infty, 0]$, $I_2 = (-\infty, 0]$ and $I_3 = (0, +\infty)$.
An atlas of M is $\{(U_1 = I_1 \cup I_3, \varphi_1 = id), (U_2 = I_2 \cup I_3, \varphi_2 = id)\}$. [See figure 3.4.]

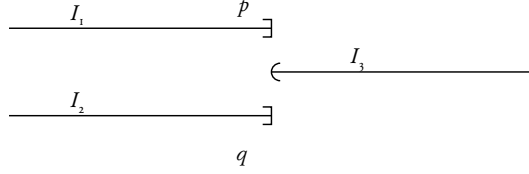


Figure 3.4: Example of a non-Hausdorff manifold.

Exercise: Prove that it is an atlas.

Note that given any neighborhood W_1 of $\varphi_1(0)$ in \mathbb{R} and any neighborhood W_2 of $\varphi_2(0)$ we necessarily have $\varphi_1^{-1}(W_1) \cap \varphi_2^{-1}(W_2) \neq \emptyset$.

3.2 Differentiable Functions on M

From now on we will assume that M is a C^∞ manifold, that is, all its maps $\varphi_i \circ \varphi_j^{-1}$ are infinitely differentiable. Although mathematically this is a restriction, it is not in physical applications. In these, M is generally the space of possible states of the system and therefore its points cannot be determined with absolute certainty, as every measurement involves some error. This indicates that through measurements we could never know the degree of differentiability of M . For convenience, we will assume it is C^∞ .

A function on M is a map $f : M \rightarrow \mathbb{R}$, that is, a map that assigns a real number to each point of M . The information encoded in the atlas on M allows us to say how smooth f is.

Definition: We will say that f is **p -times continuously differentiable** at the point $q \in M$, $f \in C_q^p$ if given (U_i, φ_i) with $q \in U_i$, $f \circ \varphi_i^{-1} : \varphi_i(U_i) \subset \mathbb{R}^n \rightarrow \mathbb{R}$ is p -times continuously differentiable at $\varphi_i(q)$.

Note that this property is independent of the chart used [as long as we consider only charts from the compatible class of atlases]. We will say that $f \in C^p(M)$ if $f \in C_q^p \forall q \in M$. [See figure 3.5.]

In practice, one defines a particular function $f \in C^p(M)$ by introducing functions $f_i : \varphi_i(U_i) \subset \mathbb{R}^n \rightarrow \mathbb{R}$ (that is, $f_i(x^j)$ where x^j are the Cartesian coordinates

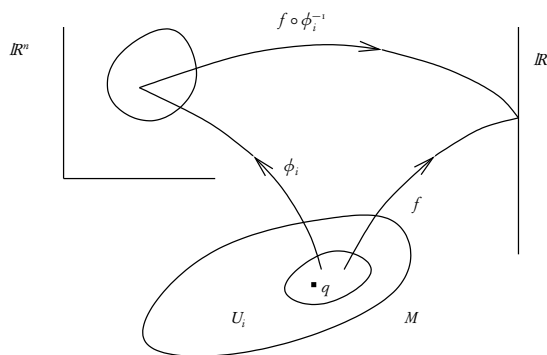


Figure 3.5: Composition of the map of a chart with a function.

in $\phi_i(U_i) \subset \mathbb{R}^n$) that are C^p in $\phi_i(U_i)$ and such that $f_i = f_j \circ \phi_j \circ \phi_i^{-1}$ in $\phi_i(U_i \cap U_j)$. This guarantees that the set of f_i determines a unique function $f \in C^p(M)$. The set of $f_i (= f \circ \phi_i^{-1})$ forms a **representation of f** in the atlas $\{(U_i, \phi_i)\}$. [See figure 3.6.]

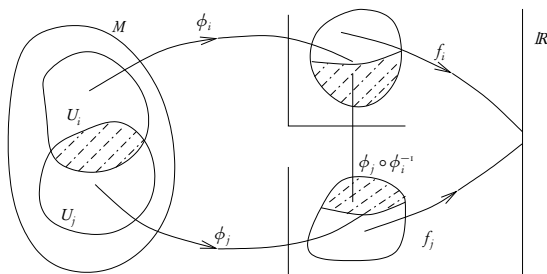


Figure 3.6: The relationship between the f_i .

Exercise: The circle, S^1 , can be thought of as the interval $[0, 1]$ with its ends identified. What are the functions in $C^2(S^1)$?

Using the previous construction, one can also define maps from M to \mathbb{R}^m that are p -times differentiable. Now we perform the inverse construction, that is, we will define the differentiability of a map from \mathbb{R}^m to M . We will do the case $\mathbb{R} \rightarrow M$, in which case the map thus obtained is called a curve. The general case is obvious.

3.3 Curves in M

Definition: A curve in M is a map between an interval $I \subset \mathbb{R}$ and M , $\gamma : I \rightarrow M$.

Note that the curve is the map and not its graph in M , that is, the set $\gamma[I]$. Thus, it is possible to have two different curves with the same graph. This is not a mathematical whim but a physical necessity: it is not the same for a car to travel the road Córdoba–Carlos Paz at 10 km/h as at 100 km/h, or to travel it in the opposite direction.

Definition: We will say that $\gamma \in C^p_{t_0}$ if given a chart (U_i, ϕ_i) such that $\gamma(t_0) \in U_i$ the map $\phi_i \circ \gamma(t) : I_{t_0} \subset I \rightarrow \mathbb{R}^n$ is p -times continuously differentiable at t_0 . [See figure 3.7.]

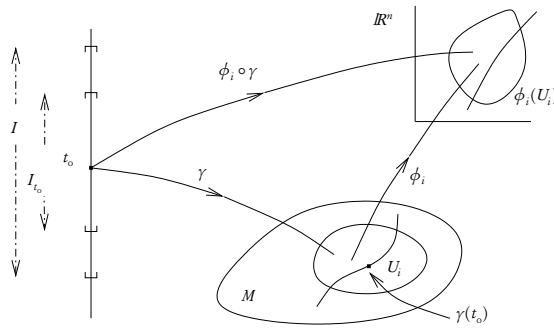


Figure 3.7: Differentiability of curves in M .

Exercise: Prove that the previous definition does not depend on the chart used.

This time we have used the concept of differentiability between maps from \mathbb{R} to \mathbb{R}^n .

Definition: A curve $\gamma(t) \in C^p(I)$ if $\gamma(t) \in C^p_t \forall t \in I$.

Exercise: How would you define the concept of differentiability of maps between two manifolds?

Of particular importance among these are the maps from M to itself $g : M \rightarrow M$ that are continuously differentiable and invertible. They are called **Diffeomorphisms**. From now on we will assume that all manifolds, curves, diffeomorphisms, and functions are smooth, that is, they are C^∞ .

3.4 Vectors

To define vectors at points of M we will use the concept of directional derivative at points of \mathbb{R}^n , that is, we will exploit the fact that in \mathbb{R}^n there is a one-to-one correspondence between vectors $(v^1, \dots, v^n)|_{x_0}$ and directional derivatives $v(f)|_{x_0} = v^i \frac{\partial}{\partial x^i} f \Big|_{x_0}$.

As we have defined differentiable functions on M we can define derivations, or directional derivatives, at its points and identify with them the tangent vectors.

Definition: A **tangent vector** v at $p \in M$ is a map

$$v : C^\infty(M) \rightarrow \mathbb{R}$$

satisfying: $\forall f, g \in C^\infty(M), a, b \in \mathbb{R}$

i) Linearity; $v(af + bg)|_p = a v(f)|_p + b v(g)|_p$.

ii) Leibniz; $v(fg)|_p = f(p)v(g)|_p + g(p)v(f)|_p$.

Note that if $h \in C^\infty(M)$ is the constant function, $h(q) = c \quad \forall q \in M$, then $v(h) = 0$. [i) $\implies v(h^2) = v(ch) = c v(h)$ while ii) $\implies v(h^2) = 2h(p)v(h) = 2c v(h)$]. These properties also show that $v(f)$ depends only on the behavior of f at p .

Exercise: Prove this last statement.

Let T_p be the set of all vectors at p . This set has the structure of a vector space and is called the **tangent space at the point** p . Indeed, we can define the sum of two vectors v_1, v_2 as the vector, that is the map, satisfying i) and ii), $(v_1 + v_2)(f) = v_1(f) + v_2(f)$ and the product of the vector v by the number a as the map $(av)(f) = a v(f)$.

As in \mathbb{R}^n , the dimension of the vector space T_p , (that is, the maximum number of linearly independent vectors), is n .

Theorem 3.1 $\dim T_p = \dim M$.

Proof: This will consist of finding a basis for T_p . Let $\dim M = n$ and (U, φ) such that $p \in U$ and $f \in C^\infty(M)$ any. For $i = 1, \dots, n$ we define the vectors $x_i : C^\infty(M) \rightarrow \mathbb{R}$ given by,

$$x_i(f) := \frac{\partial}{\partial x^i} (f \circ \varphi^{-1}) \Big|_{\varphi(p)}. \quad (3.1)$$

Note that these maps satisfy i) and ii) and therefore the x_i are really vectors. Note also that the right-hand side of 3.1 is well defined since we have the usual partial derivatives of maps between \mathbb{R}^n and \mathbb{R} . These x_i depend on the chart (U, φ) but this

does not matter in the proof since T_p does not depend on any chart. These vectors are linearly independent, that is if $x = \sum_{i=1}^n c^i x_i = 0$ then $c^i = 0 \forall i = 1, \dots, n$. This is easily seen by considering the functions (strictly defined only in U), $f^j := x^j \circ \varphi$, since $x_i(f^j) = \delta_i^j$ and therefore $0 = x(f^j) = c^j$. It only remains to show that any vector v can be expressed as a linear combination of the x_i . For this we will use the following result whose proof we leave as an exercise.

Lemma 3.1 *Let $F : \mathbb{R}^n \rightarrow \mathbb{R}$, $F \in C^\infty(\mathbb{R}^n)$ then for each $x_0 \in \mathbb{R}^n$ there exist functions $H_i : \mathbb{R}^n \rightarrow \mathbb{R} \in C^\infty(\mathbb{R}^n)$ such that $\forall x \in \mathbb{R}^n$ it holds*

$$F(x) = F(x_0) + \sum_{i=1}^n (x^i - x_0^i) H_i(x) \text{ and} \quad (3.2)$$

$$\text{also,} \quad \left. \frac{\partial F}{\partial x^i} \right|_{x=x_0} = H_i(x_0). \quad (3.3)$$

We now continue the proof of the previous Theorem. Let $F = f \circ \varphi^{-1}$ and $x_0 = \varphi(p)$, then $\forall q \in U$ we have

$$f(q) = f(p) + \sum_{i=1}^n (x^i \circ \varphi(q) - x^i \circ \varphi(p)) H_i \circ \varphi(q) \quad (3.4)$$

Using *i*) and *ii*) we obtain,

$$\begin{aligned} v(f) &= v(f(p)) + \sum_{i=1}^n (x^i \circ \varphi(q) - x^i \circ \varphi(p)) \Big|_{q=p} v(H_i \circ \varphi) \\ &\quad + \sum_{i=1}^n (H_i \circ \varphi) \Big|_p v(x^i \circ \varphi - x^i \circ \varphi(p)) \\ &= \sum_{i=1}^n (H_i \circ \varphi) \Big|_p v(x^i \circ \varphi) \\ &= \sum_{i=1}^n v^i x_i(f) \end{aligned} \quad (3.5)$$

where $v^i \equiv v(x^i \circ \varphi)$, and therefore we have expressed v as a linear combination of the x_i , thus concluding the proof ♠

The basis $\{x_i\}$ is called a **coordinate basis** and the $\{v^i\}$, the **components of v** in that basis.

Exercise: If $(\tilde{U}, \tilde{\varphi})$ is another chart such that $p \in \tilde{U}$, then it will define another coordinate basis $\{\tilde{x}_i\}$. Show that

$$x_j = \sum_{i=1}^n \frac{\partial \tilde{x}^i}{\partial x^j} \tilde{x}_i$$

where \tilde{x}^i is the i -th component of the map $\tilde{\varphi} \circ \varphi^{-1}$. Also show that the relationship between the components is $\tilde{v}^i = \sum_{j=1}^n \frac{\partial \tilde{x}^i}{\partial x^j} v^j$.

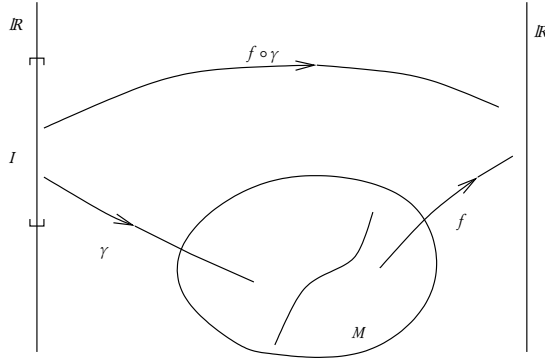


Figure 3.8: Definition of vector.

Example: Let $\gamma : I \rightarrow M$ be a curve in M . At each point $\gamma(t_0)$, $t_0 \in I$, of M we can define a vector as follows, [See figure 3.8.]

$$t(f) = \frac{d}{dt} (f \circ \gamma)|_{t=t_0}. \quad (3.6)$$

Its components in a coordinate basis are obtained through the functions

$$x^i(t) = x^i \circ \varphi \circ \gamma(t) \quad (3.7)$$

$$\begin{aligned} \frac{d}{dt}(f \circ \gamma) &= \frac{d}{dt}(f \circ \varphi^{-1} \circ \varphi \circ \gamma) \\ &= \frac{d}{dt}(f \circ \varphi^{-1}(x^i(t))) \\ &= \sum_{i=1}^n \left(\frac{\partial}{\partial x^i} (f \circ \varphi^{-1}) \right) \frac{dx^i}{dt} \\ &= \sum_{i=1}^n \frac{dx^i}{dt} x_i(f) \end{aligned} \quad (3.8)$$

3.5 Vector and Tensor Fields

If to each point q of M we assign a vector $v|_q \in T_q$ we will have a **vector field**. This will be in $C^\infty(M)$ if given any $f \in C^\infty(M)$ the function $v(f)$, which at each point p of M assigns the value $v|_p(f)$, is also in $C^\infty(M)$. We will denote the set of C^∞ vector fields by TM and it is obviously a vector space of infinite dimension.

3.5.1 The Lie Bracket

Now consider the operation in the set TM of vector fields, $[\cdot, \cdot]: TM \times TM \rightarrow TM$. This operation is called the **Lie bracket** and given two vector fields (C^∞) it gives us a third:

$$[x, y](f) := x(y(f)) - y(x(f)). \quad (3.9)$$

Exercise:

- 1) Show that $[x, y]$ is indeed a vector field.
- 2) See that the **Jacobi identity** is satisfied:

$$[[x, y], z] + [[z, x], y] + [[y, z], x] = 0 \quad (3.10)$$

3) Let x^i and x^j be two vector fields coming from a coordinate system, that is $x^i(f) = \frac{\partial f}{\partial x^i}$, etc. Show that $[x^i, x^j] = 0$.

4) Given the components of x and y in a coordinate basis, what are those of $[x, y]$?

With this operation TM acquires the character of an algebra, called **Lie Algebra**.

3.5.2 Diffeomorphisms and the Theory of Ordinary Differential Equations

Definition: A **one-parameter group of diffeomorphisms** g^t is a map $\mathbb{R} \times M \rightarrow M$ such that:

- 1) For each fixed t it is a diffeomorphism $^1 M \rightarrow M$
- 2) For any pair of real numbers, $t, s \in \mathbb{R}$ we have $g^t \circ g^s = g^{t+s}$ (in particular $g^0 = id$).

We can associate with g^t a vector field in the following way: For a fixed p , $g^t(p): \mathbb{R} \rightarrow M$ is a curve that at $t = 0$ passes through p and therefore defines a tangent vector at p , $v|_p$. Repeating the process for every point in M we have a vector field in M . Note that due to the group property satisfied by g^t , the tangent vector to the curve $g^t(p)$ is also tangent to the curve $g^s(g^t(p))$ at $s = 0$.

We can ask the inverse question: Given a smooth vector field v in M , does there exist a one-parameter group of diffeomorphisms that defines it? The answer to this question, which consists of finding all the integrable curves $g^t(p)$ that pass through each $p \in M$, is the theory of ordinary differential equations, –which will be the subject of our study in the following chapters– since it consists of solving the equations $\frac{dx^i}{dt} = v^i(x^j)$ with initial conditions $x^i(0) = \varphi^i(p) \quad \forall p \in M$. As we will see, the answer is affirmative but only locally, that is, we can only find g^t defined in $I(\subset \mathbb{R}) \times U(\subset M) \rightarrow M$.

¹That is, a smooth map with a smooth inverse.

Example: In \mathbb{R}^1 let the vector have the coordinate component x^2 , that is $v(x) = x^2 \frac{\partial}{\partial x}$. The ordinary differential equation associated with this vector is $\frac{dx}{dt} = x^2$, whose solution is

$$t - t_0 = \frac{-1}{x} + \frac{1}{x_0} \iff x(t) = \frac{-1}{t - \frac{1}{x_0}} \quad (3.11)$$

where we have taken $t_0 = 0$. That is, $g^t(x_0) = \frac{-1}{t - \frac{1}{x_0}}$. Note that for any t this map is

not defined for all \mathbb{R} and therefore is not a diffeomorphism. Also note that for any interval we take for its definition, the time interval of the solution's existence will be finite, either towards the future or the past.

Example: Let g^t be a linear diffeomorphism in \mathbb{R} , that is $g^t(x + \alpha y) = g^t(x) + \alpha g^t(y)$. Then it has the form $g^t(x) = f(t)x$. The group property implies $f(t) \cdot f(s) = f(t+s)$ or $f(t) = c e^{kt} = e^{kt}$, since $g^0 = id$. Therefore $g^t(x) = e^{kt} x$. The associated differential equation is: $x(t) = e^{kt} x_0 \implies \dot{x} = k e^{kt} x_0 = \boxed{kx = \dot{x}}$.

Exercise: Plot in a neighborhood of the origin in \mathbb{R}^2 the integral curves and therefore g^t of the following linear systems.

$$\begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & k \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \quad (3.12)$$

a) $k > 1$ b) $k = 1$ c) $0 < k < 1$, d) $k = 0$, e) $k < 0$

3.5.3 Covector and Tensor Fields

Just as we introduced the notion of a vector field, we can also introduce the notion of a covector field, that is, a smooth map from M to T_p^* . This will act on vector fields giving as a result functions on M . In the following example, we see how the field **differential of f** is defined.

Example: Let $f \in C_p^\infty$. A vector at $p \in M$ is a derivation on functions in C_p^∞ , $v(f) \in \mathbb{R}$. But given v_1 and $v_2 \in T_p$, $a \in \mathbb{R}$ and $f \in C_p^\infty$, $(v_1 + av_2)(f) = v_1(f) + av_2(f)$ and therefore each given f defines a linear functional $df|_p: T_p \rightarrow \mathbb{R}$, called the differential of f , that is an element of T_p^* ,

$$df(v) := v(f), \quad \forall v \in T_p.$$

In this way, the differential of a function, df , is a covector that when acting on a vector v gives us the number *the derivative of f at the point p in the direction of v* .

Let f be a smooth function on M , $a \in \mathbb{R}$ and consider the subset S_a of M such that $f(S_a) = a$. It can be seen that if $df \neq 0$ this will be a submanifold of M , that is, a surface embedded in M , of dimension $n - 1$. The condition $df|_p(v) = 0$ on

vectors of T_p with $p \in S_a$ means that these are actually tangent vectors to S_a , that is, elements of $T_p(S_a)$. On the contrary, if $df(v)|_p \neq 0$ then at that point v pierces S_a .

Example: The function $f(x, y, z) = x^2 + y^2 + z^2$ in \mathbb{R}^3 .

$S_a = \{(x, y, z) \in \mathbb{R}^3 | f(x, y, z) = a^2, a > 0\}$ is the sphere of radius a , and as we have already seen, a manifold. Let (v^x, v^y, v^z) be a vector at the point $(x, y, z) \in \mathbb{R}^3$, then the condition $df(v) = 2(xv^x + yv^y + zv^z) = 0$ implies that v is tangent to S . Indeed, we see that this is the condition that tells us that v is *perpendicular* to (x, y, z) when we are in the conventional Euclidean structure.

Given a coordinate system (chart) that covers a point $p \in M$, we have seen that we have a canonical basis of T_p associated with it given by the vectors,

$$x_i(f) := \frac{\partial f \circ \phi^{-1}}{\partial x^i} |_{\phi(p)}.$$

What will be the associated cobasis? Note that the coordinate system also gives us a set of n privileged functions, that is, the components of the map ϕ that defines the chart, $\{x^j\}$, $j = 1..n$, $x^i(p) := \text{value of the } i\text{-th coordinate assigned by } \phi \text{ to the point } p$. Note that $x^i \circ \phi^{-1}$ is then the identity map for the i -th coordinate. If we apply the basis vectors to these functions, we obtain,

$$x_i(x^j) := \frac{\partial x^j \circ \phi^{-1}}{\partial x^i} |_{\phi(p)} = \delta_i^j,$$

but then, since $dx^j(x_i) = x_i(x^j) = \delta_i^j$, we see that the differentials dx^j are the cobasis of the coordinate basis. In particular, we have that the coordinate components of a vector v are given by:

$$v^j = dx^j(v),$$

and the components of a covector $\omega \in T_p^*$ by,

$$\omega_i = \omega(x_i).$$

Similarly, we define **tensor fields** as the multilinear maps that when acting on vector and covector fields give functions from the manifold to the reals and that at each point of the manifold only depend on the vectors and covectors defined at that point. This last clarification is necessary because otherwise, we would include among the tensors, for example, line integrals over vector fields.

3.5.4 The Metric

Let M be an n -dimensional manifold. We have previously defined on M the notions of curves, vector fields, and covector fields, etc., but not a notion of distance between its points, that is, a function $d : M \times M \rightarrow \mathbb{R}$ that takes any two points, p and q of M and gives us a number $d(p, q)$ satisfying,

1. $d(p, q) \geq 0$.
2. $d(p, q) = 0 \leftrightarrow p = q$.
3. $d(p, q) = d(q, p)$.
4. $d(p, q) \leq d(p, r) + d(r, q)$.

This, and in some cases a notion of pseudo-distance [where 1) and 2) are not satisfied], is fundamental if we want to have a mathematical structure that is useful for the description of physical phenomena. For example, Hooke's law, which tells us that the force applied to a spring is proportional to its elongation (a distance), clearly needs this entity. Next, we will introduce a notion of infinitesimal distance, that is, between two infinitesimally separated points, which corresponds to the Euclidean notion of distance and allows us to develop a notion of global distance, that is, between any two points of M .

The idea is then to have a concept of distance (or pseudo-distance) between two *infinitesimally close* points, that is, two points connected by an *infinitesimal displacement*, that is, connected by a vector. The notion we need is then that of the norm of a vector. Since a manifold is locally like \mathbb{R}^n , in the sense that the space of tangent vectors at a point p , $T_p M$ is \mathbb{R}^n , it is reasonable to consider there the notion of Euclidean distance, that is, the distance between two points x_0 and $x_1 \in \mathbb{R}^n$ is the square root of the sum of the squares of the components (in some coordinate system) of the vector connecting these two points. The problem with this is that such a notion depends on the coordinate system being used and therefore there will be as many distances as coordinate systems covering the point p . This is just an indication that the structure we have so far does not contain a privileged or natural notion of distance. This must be introduced as an additional structure. One way to obtain infinitesimal distances independent of the coordinate system (that is, geometric) is by introducing at each point $p \in M$ a tensor of type $\binom{0}{2}$, symmetric [$\mathbf{g}(\mathbf{u}, \mathbf{v}) = \mathbf{g}(\mathbf{v}, \mathbf{u}) \quad \forall \mathbf{u}, \mathbf{v} \in T_p M$] and non-degenerate [$\mathbf{g}(\mathbf{u}, \mathbf{v}) = 0 \quad \forall \mathbf{v} \in T_p M \Rightarrow \mathbf{u} = 0$]. If we also require that this tensor be positive definite [$\mathbf{g}(\mathbf{u}, \mathbf{u}) \geq 0 \quad (= \Leftrightarrow \mathbf{u} = 0)$] it can be easily seen that this defines an inner product in $T_p M$ (or pseudo-inner product if $\mathbf{g}(\mathbf{u}, \mathbf{u}) = 0$ for some $\mathbf{u} \neq 0 \in T_p M$).² If we make a smooth choice for this tensor at each point of M we will obtain a smooth tensor field called the **metric** of M . This extra structure, a tensor field with certain properties, is what allows us to build the mathematical foundations to then construct much of physics on it.

Let \mathbf{g} be a metric on M , given any point p of M there exists a coordinate system in which its components are

$$g_{ij} = \delta_{ij}$$

and therefore gives rise to the Euclidean inner product, however, in general, this result cannot be extended to a neighborhood of the point and in general, its components will depend there on the coordinates. Note that this is what we wanted to do

²Later we will see that an inner product gives rise to a distance, correspondingly a pseudo-inner product gives rise to a pseudo-distance.

initially, but now by defining this norm via a vector we have given it an invariant character.

Restricting ourselves now to positive definite metrics, we define **the norm** of a vector $v \in T_p$ as $|v| = \sqrt{|g(v, v)|}$, that is, as the infinitesimal distance divided by ϵ between p and the point $\gamma(\epsilon)$ where $\gamma(t)$ is a curve such that $\gamma(0) = p$, $\frac{d\gamma(t)}{dt}|_{t=0} = v$. Similarly, we can define the length of a smooth curve $\gamma(t) : [0, 1] \rightarrow M$ by the formula,

$$L(\gamma) = \int_0^1 \sqrt{g(\dot{\gamma}, \dot{\gamma})} dt, \quad (3.13)$$

where $\dot{\gamma}(t) = \frac{d\gamma(t)}{dt}$. We see then that we define the length of a curve by measuring the infinitesimal lengths between nearby points on it and then integrating with respect to t .

Exercise: Prove that the length $L(\gamma)$ is independent of the chosen parameter.

We define the distance between two points $p, q \in M$ as,

$$d_g(p, q) = \inf_{\{\gamma(t) : \gamma(0)=p, \gamma(1)=q\}} |L(\gamma)| \quad (3.14)$$

That is, as the infimum of the length of all curves connecting p with q .

Exercise: Find an example of a manifold with two points such that the infimum in the previous definition is not a minimum. That is, where there is no curve connecting the two points with the minimum distance between them.

Exercise: a) The Euclidean metric in \mathbb{R}^2 is $(dx)^2 + (dy)^2$, where $\{dx, dy\}$ is the cobasis associated with $\{\partial x, \partial y\}$. What is the distance between two points in this case?

Exercise: b) What is the form of the Euclidean metric in \mathbb{R}^3 in spherical coordinates? And in cylindrical coordinates?

Exercise: c) The metric of the sphere is $(d\theta)^2 + \sin^2 \theta (d\varphi)^2$. What is the distance in this case? For which points p, q are there multiple curves γ_i with $L(\gamma_i) = d(p, q)$?

Exercise: d) The metric $(dx)^2 + (dy)^2 + (dz)^2 - (dt)^2$ in \mathbb{R}^4 is the Minkowski metric of special relativity. What is the *distance* between the point with coordinates $(0, 0, 0, 0)$ and $(1, 0, 0, 1)$?

A metric gives us a privileged map between the space of tangent vectors at p , T_p , and its dual T_p^* for each p in M , that is, the map that assigns to each vector $v \in T_p$ the

covector $g(v, \cdot) \in T_p^*$. Since this is valid for each p , we thus obtain a map between vector and covector fields.

3.5.5 Diffeomorphisms and the Theory of Ordinary Differential Equations

Definition: A one-parameter group of diffeomorphisms g^t is a map $\mathbb{R} \times M \rightarrow M$ such that:

- 1) For each fixed t it is a diffeomorphism $^3 M \rightarrow M$
- 2) For any pair of real numbers, $t, s \in \mathbb{R}$ we have $g^t \circ g^s = g^{t+s}$ (in particular $g^0 = id$).

We can associate with g^t a vector field in the following way: For a fixed p , $g^t(p) : \mathbb{R} \rightarrow M$ is a curve that at $t = 0$ passes through p and therefore defines a tangent vector at p , $v|_p$. Repeating the process for every point in M we have a vector field in M . Note that due to the group property satisfied by g^t , the tangent vector to the curve $g^t(p)$ is also tangent to the curve $g^s(g^t(p))$ at $s = 0$.

We can ask the inverse question: Given a smooth vector field v in M , does there exist a one-parameter group of diffeomorphisms that defines it? The answer to this question, which consists of finding all the integrable curves $g^t(p)$ that pass through each $p \in M$, is the theory of ordinary differential equations, –which will be the subject of our study in the following chapters– since it consists of solving the equations $\frac{dx^i}{dt} = v^i(x^j)$ with initial conditions $x^i(0) = \varphi^i(p) \quad \forall p \in M$. As we will see, the answer is affirmative but only locally, that is, we can only find g^t defined in $I(\subset \mathbb{R}) \times U(\subset M) \rightarrow M$.

Example: In \mathbb{R}^1 let the vector have the coordinate component x^2 , that is $v(x) = x^2 \frac{\partial}{\partial x}$. The ordinary differential equation associated with this vector is $\frac{dx}{dt} = x^2$, whose solution is

$$t - t_0 = \frac{-1}{x} + \frac{1}{x_0} \quad \text{or} \quad x(t) = \frac{-1}{t - \frac{1}{x_0}} \quad (3.15)$$

where we have taken $t_0 = 0$. That is, $g^t(x_0) = \frac{-1}{t - \frac{1}{x_0}}$. Note that for any t this map is

not defined for all \mathbb{R} and therefore is not a diffeomorphism. Also note that for any interval we take for its definition, the time interval of the solution's existence will be finite, either towards the future or the past.

Example: Let g^t be a linear diffeomorphism in \mathbb{R} , that is $g^t(x + \alpha y) = g^t(x) + \alpha g^t(y)$. Then it has the form $g^t(x) = f(t)x$. The group property implies $f(t) \cdot$

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$f(s) = f(t+s)$ or $f(t) = c e^{kt} = e^{kt}$, since $g^0 = id$. Therefore $g^t(x) = e^{kt} x$. The associated differential equation is: $x(t) = e^{kt} x_0 \implies \dot{x} = k e^{kt} x_0 = \boxed{k x = \dot{x}}$.

Exercise: Plot in a neighborhood of the origin in \mathbb{R}^2 the integral curves and therefore g^t of the following linear systems.

$$\begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & k \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \quad (3.16)$$

a) $k > 1$ b) $k = 1$ c) $0 < k < 1$, d) $k = 0$, e) $k < 0$

3.5.6 Covector and Tensor Fields

Just as we introduced the notion of a vector field, we can also introduce the notion of a covector field, that is, a smooth map from M to T_p^* . This will act on vector fields giving as a result functions on M . In the following example, we see how the field **differential of f** is defined.

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but then, since $dx^j(x_i) = x_i(x^j) = \delta_i^j$, we see that the differentials dx^j are the cobasis of the coordinate basis. In particular, we have that the coordinate components of a vector v are given by:

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3.5.7 The Metric

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4. $d(p, q) \leq d(p, r) + d(r, q)$.

This, and in some cases a notion of pseudo-distance [where 1) and 2) are not satisfied], is fundamental if we want to have a mathematical structure that is useful for the description of physical phenomena. For example, Hooke's law, which tells us that the force applied to a spring is proportional to its elongation (a distance), clearly needs this entity. Next, we will introduce a notion of infinitesimal distance, that is, between two infinitesimally separated points, which corresponds to the Euclidean notion of distance and allows us to develop a notion of global distance, that is, between any two points of M .

The idea is then to have a concept of distance (or pseudo-distance) between two *infinitesimally close* points, that is, two points connected by an *infinitesimal displacement*, that is, connected by a vector. The notion we need is then that of the norm of a vector. Since a manifold is locally like \mathbb{R}^n , in the sense that the space of tangent vectors at a point p , $T_p M$ is \mathbb{R}^n , it is reasonable to consider there the notion of Euclidean distance, that is, the distance between two points x_0 and $x_1 \in \mathbb{R}^n$ is the square root of the sum of the squares of the components (in some coordinate system) of the vector connecting these two points. The problem with this is that such a notion depends on the coordinate system being used and therefore there will be as many distances as coordinate systems covering the point p . This is just an indication that the structure we have so far does not contain a privileged or natural notion of distance. This must be introduced as an additional structure. One way to obtain infinitesimal distances independent of the coordinate system (that is, geometric) is by introducing at each point $p \in M$ a tensor of type $\binom{0}{2}$, symmetric $[\mathbf{g}(\mathbf{u}, \mathbf{v}) = \mathbf{g}(\mathbf{v}, \mathbf{u}) \quad \forall \mathbf{u}, \mathbf{v} \in T_p M]$ and non-degenerate $[\mathbf{g}(\mathbf{u}, \mathbf{v}) = \mathbf{o} \quad \forall \mathbf{v} \in T_p M \Rightarrow \mathbf{u} = \mathbf{o}]$. If we also require that this tensor be positive definite $[\mathbf{g}(\mathbf{u}, \mathbf{u}) \geq \mathbf{o} \quad (= \Leftrightarrow \mathbf{u} = \mathbf{o})]$ it can be easily seen that this defines an inner product in $T_p M$ (or pseudo-inner product if $\mathbf{g}(\mathbf{u}, \mathbf{u}) = \mathbf{o}$ for some $\mathbf{u} \neq \mathbf{o} \in T_p M$).⁴ If we make a smooth choice for this tensor at each point of M we will obtain a smooth tensor field called the **metric** of M . This extra structure, a tensor field with certain properties, is what allows us to build the mathematical foundations to then construct much of physics on it.

Let \mathbf{g} be a metric on M , given any point p of M there exists a coordinate system in which its components are

$$g_{ij} = \delta_{ij}$$

and therefore gives rise to the Euclidean inner product, however, in general, this result cannot be extended to a neighborhood of the point and in general, its components will depend there on the coordinates. Note that this is what we wanted to do initially, but now by defining this norm via a vector we have given it an invariant character.

Restricting ourselves now to positive definite metrics, we define the **norm** of a vector $\mathbf{v} \in T_p$ as $|\mathbf{v}| = \sqrt{[\mathbf{g}(\mathbf{v}, \mathbf{v})]}$, that is, as the infinitesimal distance divided by ϵ between p and the point $\gamma(\epsilon)$ where $\gamma(t)$ is a curve such that $\gamma(0) = p$, $\frac{d\gamma(t)}{dt}|_{t=0} = \mathbf{v}$. Similarly, we can define the length of a smooth curve $\gamma(t) : [0, 1] \rightarrow M$ by the formula,

$$L(\gamma) = \int_0^1 \sqrt{\mathbf{g}(\dot{\gamma}(t), \dot{\gamma}(t))} dt, \quad (3.17)$$

where $\dot{\gamma}(t) = \frac{d\gamma(t)}{dt}$. We see then that we define the length of a curve by measuring the infinitesimal lengths between nearby points on it and then integrating with respect to t .

⁴Later we will see that an inner product gives rise to a distance, correspondingly a pseudo-inner product gives rise to a pseudo-distance.

Exercise: Prove that the length $L(\gamma)$ is independent of the chosen parameter.

We define the distance between two points $p, q \in M$ as,

$$d_g(p, q) = \inf_{\{\gamma(t) : \gamma(0)=p, \gamma(1)=q\}} |L(\gamma)| \quad (3.18)$$

That is, as the infimum of the length of all curves connecting p with q .

Exercise: Find an example of a manifold with two points such that the infimum in the previous definition is not a minimum. That is, where there is no curve connecting the two points with the minimum distance between them.

Exercise: a) The Euclidean metric in \mathbb{R}^2 is $(dx)^2 + (dy)^2$, where $\{dx, dy\}$ is the cobasis associated with $\{\partial x, \partial y\}$. What is the distance between two points in this case?

Exercise: b) What is the form of the Euclidean metric in \mathbb{R}^3 in spherical coordinates? And in cylindrical coordinates?

Exercise: c) The metric of the sphere is $(d\theta)^2 + \sin^2 \theta (d\varphi)^2$. What is the distance in this case? For which points p, q are there multiple curves γ_i with $L(\gamma_i) = d(p, q)$?

Exercise: d) The metric $(dx)^2 + (dy)^2 + (dz)^2 - (dt)^2$ in \mathbb{R}^4 is the Minkowski metric of special relativity. What is the *distance* between the point with coordinates $(0, 0, 0, 0)$ and $(1, 0, 0, 1)$?

A metric gives us a privileged map between the space of tangent vectors at p , T_p , and its dual T_p^* for each p in M , that is, the map that assigns to each vector $v \in T_p$ the covector $g(v, \cdot) \in T_p^*$. Since this is valid for each p , we thus obtain a map between vector and covector fields.

Since g is non-degenerate, this map is invertible, that is, there exists a symmetric tensor of type $\binom{2}{0}$, g^{-1} , such that

$$g(g^{-1}(\theta, \cdot), \cdot) = \theta \quad (3.19)$$

for any co-vector field θ . This indicates that when we have a manifold with a metric, it becomes irrelevant to distinguish between vectors and co-vectors or, for example, between tensors of type $\binom{0}{2}$, $\binom{2}{0}$, or $\binom{1}{1}$.

3.5.8 Abstract Index Notation

When working with tensorial objects, the notation used so far is not the most convenient because it is difficult to remember the type of each tensor, in which slot it "eats" other objects, etc. One solution is to introduce a coordinate system and work

with the components of the tensors, where having indices makes it easy to know what objects they are or to introduce general bases. In this way, for example, we represent the vector $l = l^i \frac{\partial}{\partial x^i}$ by its components $\{l^i\}$. A convenience of this notation is that "eating" becomes "contracting," since, for example, we represent the vector l "eating" a function f by the contraction of the coordinate components of the vector and the differential of f :

$$l(f) = \sum_{i=1}^n l^i \frac{\partial f}{\partial x^i}.$$

But a serious drawback of this representation is that it initially depends on the coordinate system and therefore all the expressions we construct with it have the potential danger of depending on such a system.

We will remedy this by introducing **abstract indices** (which will be Latin letters) that indicate where the coordinate indices would go but nothing more, that is, they do not depend on the coordinate system and do not even take numerical values, that is, l^a does not mean the n -tuple (l^1, l^2, \dots, l^n) as if they were indices. In this way, l^a will denote the vector l , θ_a the co-vector θ , and g_{ab} the metric g . A contraction such as $g(v, \cdot)$ will be denoted $g_{ab}v^a$ and we will denote this co-vector by v_b , that is, the action of g_{ab} is to lower the index of v^a and give the co-vector $v_b \equiv v^a g_{ab}$. Similarly, we will denote g^{-1} (the inverse of g) as g^{ab} , that is, g with the indices raised.

The symmetry of g is then equivalent to $g_{ab} = g_{ba}$.

Exercise: How would you denote an antisymmetric tensor of type $\binom{0}{2}$?

Using repeated indices for contraction, we see that $l(f)$ can be denoted by $l^a \nabla_a f$ where $\nabla_a f$ denotes the differential co-vector of f , while the vector $\nabla^a f := g^{ab} \nabla_b f$ is called the **gradient** of f and we see that it depends not only on f but also on g .

3.6 Covariant Derivative

We have seen that in M there is the notion of the derivative of a scalar field f , which is the differential co-vector of f that we denote $\nabla_a f$. Is there the notion of the derivative of a tensor field? For example, is there an extension of the operator ∇_a to vectors such that if l^a is a differentiable vector then $\nabla_a l^b$ is a tensor of type $\binom{1}{1}$? To fix ideas, let us define this extension of the differential ∇_a , called the covariant derivative, by requiring it to satisfy the following properties:⁵

i) Linearity: If $A_{b_1 \dots b_l}^{a_1 \dots a_k}, B_{b_1 \dots b_l}^{a_1 \dots a_k}$ are tensors of type $\binom{k}{l}$ and $\alpha \in \mathbb{R}$ then

$$\nabla_c (\alpha A_{b_1 \dots b_l}^{a_1 \dots a_k} + B_{b_1 \dots b_l}^{a_1 \dots a_k}) = \alpha \nabla_c A_{b_1 \dots b_l}^{a_1 \dots a_k} + \nabla_c B_{b_1 \dots b_l}^{a_1 \dots a_k} \quad (3.20)$$

⁵Note that they are an extension of those required to define derivations.

ii) Leibnitz:

$$\nabla_e \left(A_{b_1 \dots b_l}^{a_1 \dots a_k} B_{d_1 \dots d_n}^{c_1 \dots c_m} \right) = A_{b_1 \dots b_l}^{a_1 \dots a_k} \left(\nabla_e B_{d_1 \dots d_n}^{c_1 \dots c_m} \right) + \left(\nabla_e A_{b_1 \dots b_l}^{a_1 \dots a_k} \right) B_{d_1 \dots d_n}^{c_1 \dots c_m} \quad (3.21)$$

iii) Commutativity with contractions:

$$\nabla_e \left(\delta^c_d A_{b_1 \dots c, \dots, b_k}^{a_1, \dots, d, \dots, a_k} \right) = \delta^c_d \nabla_e A_{b_1 \dots c, \dots, b_k}^{a_1, \dots, d, \dots, a_k}, \quad (3.22)$$

where δ^c_d is the identity tensor. That is, if we first contract some indices of a tensor and then take its derivative, we obtain the same tensor as if we first take the derivative and then contract.

iv) Consistency with the differential: If l^a is a vector field and f a scalar field, then

$$l^a \nabla_a f = l(f) \quad (3.23)$$

v) Zero torsion: If f is a scalar field then

$$\nabla_a \nabla_b f = \nabla_b \nabla_a f \quad (3.24)$$

Example: Let $\{x^i\}$ be a global coordinate system in \mathbb{R}^n and let ∇_c be the operator that, when acting on $A_{b_1 \dots b_l}^{a_1 \dots a_k}$, generates the tensor field that in these coordinates has components

$$\partial_j A_{i_1 \dots i_l}^{i_1 \dots i_k} \quad (3.25)$$

By definition, it is a tensor and clearly satisfies all the conditions of the definition, since it satisfies them in this coordinate system, so it is a covariant derivative. If we take another coordinate system, we will obtain another covariant derivative, generally different from the previous one. For example, let us act ∇_c on the vector l^a , then

$$(\nabla_c l^a)_j^i = \partial_j l^i. \quad (3.26)$$

In another coordinate system $\{\bar{x}^i\}$ this tensor has components

$$(\nabla_c l^a)_k^l = \sum_{i,j=1}^n \frac{\partial \bar{x}^l}{\partial x^i} \frac{\partial x^j}{\partial \bar{x}^k} \frac{\partial l^i}{\partial x^j} \quad (3.27)$$

which are not, in general, the components of the covariant derivative $\bar{\nabla}_c$ that these new coordinates define, indeed

$$\begin{aligned} (\bar{\nabla}_c l^a)_k^l &\equiv \frac{\partial \bar{l}^l}{\partial \bar{x}^k} = \sum_{j=1}^n \left(\frac{\partial x^j}{\partial \bar{x}^k} \right) \frac{\partial}{\partial x^j} \sum_{i=1}^n \left(\frac{\partial \bar{x}^l}{\partial x^i} l^i \right) = \\ &= \sum_{i,j=1}^n \left(\frac{\partial \bar{x}^l}{\partial x^i} \right) \left(\frac{\partial x^j}{\partial \bar{x}^k} \right) \frac{\partial l^i}{\partial x^j} + \sum_{i,j=1}^n \frac{\partial x^j}{\partial \bar{x}^k} \left(\frac{\partial^2 \bar{x}^l}{\partial x^j \partial x^i} \right) l^i \\ &= (\nabla_c l^a)_j^i + \sum_{i,j=1}^n \frac{\partial x^j}{\partial \bar{x}^k} \left(\frac{\partial^2 \bar{x}^l}{\partial x^j \partial x^i} \right) l^i, \end{aligned} \quad (3.28)$$

which clearly shows that they are two different tensors and that their difference is a **tensor** that depends linearly and **not differentially** on l^a . Is this true in general? That is, given two connections, ∇_c and $\bar{\nabla}_c$, is their difference a tensor (and not a differential operator)? We will see that this is true.

Theorem 3.2 *The difference between two connections is a tensor.*

Proof: Note that by properties *iii*) and *iv*) of the definition, if we know how ∇_c acts on co-vectors, we know how it acts on vectors and thus by *i*) and *ii*) on any tensor. Indeed, if we know $\nabla_c w_a$ for any w_a , then $\nabla_c l^a$ is the tensor of type $\binom{1}{1}$ such that when contracted with an arbitrary w_a gives us the co-vector

$$(\nabla_c l^a) w_a = \nabla_c (w_a l^a) - l^a (\nabla_c w_a), \quad (3.29)$$

which we know since by *iv*) we also know how ∇_c acts on scalars. Therefore, it is sufficient to see that

$$(\bar{\nabla}_c - \nabla_c) w_a = C^b_{ca} w_b \quad (3.30)$$

for some tensor C^b_{ca} . First, let us prove that given any $p \in M$, $(\bar{\nabla}_c - \nabla_c) w_a|_p$ depends only on $w_a|_p$ and not on its derivative. Let w'_a be any other co-vector such that at p they coincide, that is, $(w_a - w'_a)|_p = 0$. Then given a smooth co-basis $\{\mu_a^i\}$ in a neighborhood of p , we will have that $w_a - w'_a = \sum_i f_i \mu_a^i$ with f_i smooth functions that vanish at p . At p we then have

$$\begin{aligned} \bar{\nabla}_c (w_a - w'_a) - \nabla_c (w_a - w'_a) &= \sum_i \bar{\nabla}_c (f_i \mu_a^i) - \sum_i \nabla_c (f_i \mu_a^i) \\ &= \sum_i \mu_a^i (\bar{\nabla}_c f_i - \nabla_c f_i) = 0 \end{aligned} \quad (3.31)$$

since by *iv*) $\bar{\nabla}_c$ and ∇_c must act in the same way on scalars –and in particular on the f_i –. This shows that

$$(\bar{\nabla}_c - \nabla_c) w'_a = (\bar{\nabla}_c - \nabla_c) w_a \quad (3.32)$$

and therefore that $(\bar{\nabla}_c - \nabla_c) w_a$ depends only on $w_a|_p$ and obviously in a linear way. But then $(\bar{\nabla}_c - \nabla_c)$ must be a tensor of type $\binom{1}{2}$ that is waiting to "eat" a co-vector to give us the tensor of type $\binom{0}{2}$, $(\bar{\nabla}_c - \nabla_c) w_a$. That is, $(\bar{\nabla}_c - \nabla_c) w_a = C^b_{ca} w_b$, which proves the theorem.

Note that condition *v*) tells us that $\nabla_a \nabla_b f = \nabla_b \nabla_a f$, taking $w_a = \nabla_a f$ we get

$$\begin{aligned} \bar{\nabla}_a \bar{\nabla}_b f &= \bar{\nabla}_a \nabla_b f \\ &= \bar{\nabla}_a w_b \\ &= \nabla_a w_b + C^c_{ab} \nabla_c f \\ &= \nabla_a \nabla_b f + C^c_{ab} \nabla_c f. \end{aligned} \quad (3.33)$$

Since $\nabla_c f|_p$ can be any co-vector, we see that the condition of no torsion implies that C^c_{ab} is symmetric in the lower indices, $C^c_{ab} = C^c_{ba}$.

Exercise: How does $(\bar{\nabla}_c - \nabla_c)$ act on vectors?

Exercise: Express the Lie bracket in terms of any connection and then explicitly prove that it does not depend on the connection used.

Exercise: Let $A_{b,\dots,z}$ be a totally antisymmetric tensor. Show that $\nabla_{[a} A_{b,\dots,z]}$, that is, the total antisymmetrization of $\nabla_a A_{b,\dots,z}$, does not depend on the covariant derivative used.

The difference between any connection ∇_c and one coming from a coordinate system $\{x^i\}$ is a tensor called the **Christoffel symbol** of ∇_c with respect to the coordinates $\{x^i\}$, Γ_{ca}^b ,

$$\nabla_c w_a = \partial_c w_a + \Gamma_{ca}^b w_b. \quad (3.34)$$

The knowledge of this tensor is very useful in practice, as it allows us to express ∇_c in terms of the corresponding coordinate connection, ∂_c .

As we have seen, in a manifold M there are infinite ways to *take the derivative of a tensor*. Is there any natural or privileged one? The answer is no, unless we add more structure to M . Intuitively, the reason for this is that in M we do not know how to compare $l^a|_p$ with $l^a|_q$ if p and q are two different points.⁶

Is the presence of a metric in M sufficient to make this comparison? The answer is yes!

Theorem 3.3 *Let g_{ab} be a (smooth) metric on M , then there exists a unique covariant derivative ∇_c such that $\nabla_c g_{ab} = 0$.*

Proof: Let $\bar{\nabla}_c$ be any connection and let ∇_c be such that $\nabla_c g_{ab} = 0$, it is sufficient to show that this condition uniquely determines the difference tensor, C_{ca}^d . But,

$$0 = \nabla_a g_{bc} = \bar{\nabla}_a g_{bc} - C_{ab}^d g_{dc} - C_{ac}^d g_{bd} \quad (3.35)$$

that is

$$C_{cab} + C_{bac} = \bar{\nabla}_a g_{bc} \quad (3.36)$$

but also

$$\begin{aligned} a \leftrightarrow b & \quad C_{cba} + C_{abc} = \bar{\nabla}_b g_{ac} \\ c \leftrightarrow b & \quad C_{bca} + C_{acb} = \bar{\nabla}_c g_{ab}. \end{aligned} \quad (3.37)$$

Adding the last two, subtracting the first and using the symmetry of the last two indices we get

$$2C_{abc} = \bar{\nabla}_b g_{ac} + \bar{\nabla}_c g_{ab} - \bar{\nabla}_a g_{bc} \quad (3.38)$$

⁶Note that one way to compare infinitesimally close vectors, given a vector field m^a , is with the Lie bracket of m^a with l^a , $[m, l]^a$. This is not appropriate since $[m, l]^a|_p$ depends on the derivative of m^a at p .

or

$$C^a{}_{bc} = \frac{1}{2} g^{ad} \{ \bar{\nabla}_b g_{dc} + \bar{\nabla}_c g_{db} - \bar{\nabla}_d g_{bc} \} \quad (3.39)$$

Note that the existence of a metric is not equivalent to the existence of a connection. There are connections $\bar{\nabla}_c$ for which there is no metric g_{ab} such that $\bar{\nabla}_a g_{bc} = 0$, that is, there are tensors $C^a{}_{bc}$ for which there is no g_{ab} that satisfies (3.39).

Exercise: If $\bar{\nabla}_c$ is a derivative corresponding to a coordinate system $\bar{\nabla}_c = \partial_c$ the corresponding Christoffel symbol is

$$\Gamma^a{}_{bc} = \frac{1}{2} g^{ad} \{ \partial_b g_{dc} + \partial_c g_{db} - \partial_d g_{bc} \}. \quad (3.40)$$

Show that its components in that coordinate system are given by,

$$\Gamma^i{}_{jk} = \frac{1}{2} g^{il} \{ \partial_j g_{lk} + \partial_k g_{lj} - \partial_l g_{jk} \}, \quad (3.41)$$

where g_{ij} are the components of the metric in the same coordinate system.

Exercise: The Euclidean metric of \mathbb{R}^3 in spherical coordinates is

$$ds^2 = (dr)^2 + r^2((d\theta)^2 + \sin^2 \theta (d\phi)^2).$$

Calculate the Laplacian $\Delta = g^{ab} \nabla_a \nabla_b$ in these coordinates.

*The Riemann Tensor

Given a covariant derivative on a manifold, is it possible to define tensor fields that depend only on it and therefore give us information about it?

The answer is yes! The following tensor is called the **Riemann or Curvature Tensor** and depends only on the connection:

$${}^2\nabla_{[a} \nabla_{b]} l^c := (\nabla_a \nabla_b - \nabla_b \nabla_a) l^c := R^c{}_{dab} l^d \quad \forall l^c \in TM.$$

Exercise: Show that the definition above makes sense, that is, that the left-hand side, evaluated at any point p in M , depends only on $l^c|_p$ and therefore we can write the right-hand side for some tensor $R^c{}_{dab}$.

Exercise: Let $\bar{\nabla}_a$ be another covariant derivative. Calculate the difference between the respective Riemann tensors in terms of the tensor that appears as the difference of the two connections.

Bibliography notes: I recommend the book by Wald [6], especially for its modern notation, see also [21]. The language of intuition is geometry, it is the tool that allows us to visualize problems, touch them, turn them around to our liking and then reduce them to algebra. Understanding geometry is the most efficient way to understand physics since it is only fully understood when translated into a geometric language. Do not abuse it, it is a very vast area and it is easy to get lost, learn the basics well and then only what is relevant to you.

ORDINARY DIFFERENTIAL EQUATIONS

4.1 Introduction

In this chapter we will begin the study of systems of ordinary differential equations – ODE from now on–. These systems constantly appear in physics and together with the systems of partial differential equations, –PDE from now on– that we will study in the second part of this course, form the mathematical skeleton of the exact sciences. This is because most physical phenomena can be described in terms of these equations. If the free parameter –or independent variable– has the interpretation of being a time, then we are dealing with **evolution equations**. These allow us, for each initial state of the system, to know its temporal evolution, that is, they give us the ability to make predictions from initial conditions, which is the characteristic aspect of any physical theory.¹

There are other physical phenomena described by ODE that do not have an evolutionary character. In general, in these cases, we are interested in equilibrium states of the system. For example, if we want to see what shape the clothesline at home describes. These cases will not be included in the theory of ODE that we will develop below, but their simplest cases will be treated as special (one-dimensional) cases of elliptic PDE.

Although quantum physics, or more precisely quantum field theory, shows us that a description of physical phenomena at that microscopic scale is not possible through the systems of differential equations mentioned above, we usually manage to find partial aspects of these phenomena, or certain approximations where a description in terms of these systems is possible. This is not due to a whim or stubbornness, but to the fact that our knowledge of ODE, and to a lesser extent of PDE, is quite deep, which allows us to handle them with relative ease and comfort. On the other hand, and it is important to emphasize this, our knowledge of the type of equations that appear in non-linear quantum field theories is, in comparison, practically nil.

¹In some cases, the independent variable is not time but some other parameter, but the equation can also be considered as describing evolution. For example, if we want to see how the density of a medium varies when the temperature increases.

4.2 The Case of a Single First-Order Ordinary Equation

4.2.1 First-Order Autonomous Equation

This has the form:

$$\dot{x} = \frac{dx}{dt} = f(x) \quad (4.1)$$

that is, an equation for a map $x(t) : I \subset \mathbb{R} \rightarrow U \subset \mathbb{R}$.

We will assume $f : U \rightarrow \mathbb{R}$ continuous. A solution of this equation will be a differentiable map $\varphi(t)$ defined for all I and with image in U . We will say that $\varphi(t)$ satisfies the **initial condition** x_0 at $t = t_0 \in I$ if

$$\varphi(t_0) = x_0. \quad (4.2)$$

As we will see below, under certain hypotheses, the initial condition that a solution satisfies determines it uniquely.

The equation 4.1 can be interpreted geometrically in the following way. At each point of the plane (x, t) we mark a "little line", that is, a field of directions such that the angle it forms with the axis $x = \text{cte}$ has a tangent equal to $f(x)$.

Example: $\dot{x} = x^{1/2}$. [See figure (4.1).]

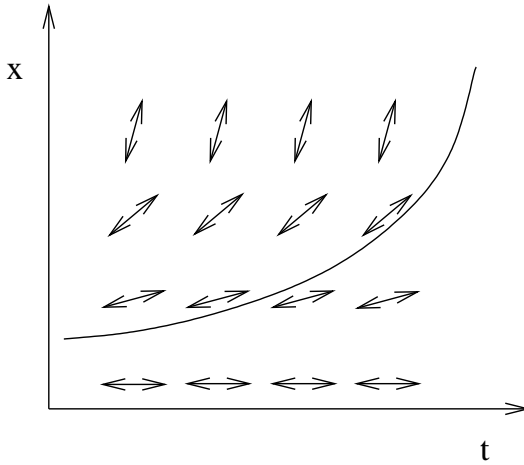


Figure 4.1: Geometric interpretation of the equation $f(x) = x^{1/2}$.

Since $\varphi(t)$ has a derivative $f(\varphi(t))$, at the point $(\varphi(t), t)$, $\varphi(t)$ will be tangent to the little line at that point. If we stand at some point (x_0, t_0) through which it passes, we can find the solution by following the little lines. Therefore, we see that given a point through which the solution passes, by following a little line, we will determine a unique solution. This, in particular, tells us that in this graph, most solutions do

not intersect. As we will see later, the uniqueness of the solution occurs if the little line at each point has a non-zero tangent. On the other hand, it is clear from the case in the previous figure, 4.1, that if we start with the point $(x_o = 0, t_o)$, we would have two options: either draw the already drawn line (t^2, t) or simply the line $(0, t)$.

Exercise: Extend this graphical intuition to the case $\dot{x} = f(x, t)$.

Example: In fact, the equation $\dot{x} = x^{1/2}$ has two solutions that pass through $(0, 0)$,

$$\varphi(t) = 0 \text{ y } \varphi = \frac{t^2}{4} \quad (4.3)$$

Indeed, suppose $x(t) \neq 0$ then $\frac{dx}{x^{1/2}} = dt$ or $2(x^{1/2} - x_o^{1/2}) = t - t_o$, taking $x_o = t_o = 0$

$$x = \left(\frac{t}{2}\right)^2 \quad (4.4)$$

The other is trivially a solution.

The uniqueness of the solutions is obtained, even in the case where $f(x)$ vanishes, if instead of asking that $f(x)$ be continuous, we ask a little more, and that is that, where it vanishes, say at x_o , it is fulfilled that

$$\lim_{\epsilon \rightarrow 0} \int_{x_o + \epsilon}^x \frac{dx}{f(x)} = \infty, \quad (4.5)$$

or alternatively that for $\epsilon > 0$ sufficiently small, there exists $k > 0$ such that,

$$|f(x)| < k |x - x_o| \text{ si } |x - x_o| < \epsilon. \quad (4.6)$$

That is, $f(x)$ goes to zero when $x \rightarrow x_o$ at most as quickly as a linear function. This condition is called the Lipschitz condition. From now on, we will assume that $f(x)$ is differentiable and therefore Lipschitz.

Exercise: See that this condition is weaker than asking that f be differentiable at x_o .

Solution: The function $f(x) = x \sin(\frac{1}{x})$ is Lipschitz at x_{+0} but not differentiable.

Thus, we arrive at our first theorem on ordinary equations.

Theorem 4.1 *Let $f(x) : U \rightarrow \mathbb{R}$ be continuous and Lipschitz at the points where it vanishes. Then*

i) *For each $t_o \in \mathbb{R}$, $x_o \in U$ there exists an interval $I_{t_o} \in \mathbb{R}$ and a solution $\varphi(t)$ of the equation 4.1 defined $\forall t \in I_{t_o}$ and the initial condition $\varphi(t_o) = x_o$;*

ii) *Any two solutions φ_1, φ_2 of 4.1 satisfying the same initial condition coincide in some neighborhood of $t = t_o$*

iii) the solution φ of 4.1 is such that

$$t - t_0 = \int_{x_0}^{\varphi(t)} \frac{dx}{f(x)} \text{ si } f(x) \neq 0 \quad (4.7)$$

Note: The solution whose existence is guaranteed by this theorem is valid in an open interval I_{t_0} , which the theorem does not determine. Therefore, uniqueness occurs only in the maximum interval that they have in common.

Proof:

If $f(x_0) = 0$ then $\varphi(t) = x_0$ is a solution.

If $f(x_0) \neq 0$ then by continuity there exists a neighborhood of x_0 where $f(x) \neq 0$ and therefore $1/f(x)$ is a continuous function, this implies that the integral in 4.29 is a differentiable function of its integration limits, which we will call $\psi(x) - \psi(x_0)$. Therefore, we have,

$$t - t_0 = \psi(x) - \psi(x_0) \quad (4.8)$$

That is, given x we obtain t and also this relationship is automatically fulfilled when $x = x_0$, $t = t_0$. We now want to solve for x from this relationship, that is, find $\phi(t)$. But

$$\left. \frac{d\psi}{dx} \right|_{x=\zeta} = \frac{1}{f(\zeta)} \neq 0, \quad (4.9)$$

Therefore, the inverse function theorem assures us that there exists an I_{t_0} and a *unique* $\varphi(t)$ differentiable, defined for all $t \in I_{t_0}$ and such that $\varphi(t_0) = x_0$ and $t - t_0 = \psi(\varphi(t)) - \psi(x_0)$. Taking its derivative, we obtain,

$$\frac{d\varphi}{dt} = \left. \frac{d\psi^{-1}}{dt} \right|_{x=\varphi(t)} = \left[\frac{1}{f(x)} \right]^{-1} \Big|_{x=\varphi(t)} = f(\varphi(t)) \quad (4.10)$$

which shows that it is a solution, necessarily unique of 4.1 with the appropriate initial condition.

It only remains to see that in the case where $f(x_0) = 0$ the solution is unique. For this, we will use the following Lemma.

Lemma 4.1 *Let f_1 and f_2 be real functions in $U \subset \mathbb{R}$ such that $f_1(x) < f_2(x) \forall x \in U$ and let φ_1 and φ_2 be the respective solutions to the differential equations that they generate, defined in an interval $I \in \mathbb{R}$ and such that $\varphi_1(t_1) = \varphi_2(t_1)$ for some $t_1 \in I$. Then*

$$\varphi_1(t) \leq \varphi_2(t) \quad \forall t > t_1 \in I \quad (4.11)$$

Proof: Trivial, the one who runs faster in each sector of the path wins! Let the set $S = \{t > t_1 \in I \mid \phi_1(t) > \phi_2(t)\}$. Let $T \geq t_1$, $T \in I$ be the greatest lower bound of S . At this moment $\phi_1(T) = \phi_2(T)$, but

$$\left. \frac{d\phi_1}{dt} \right|_{t=T} < \left. \frac{d\phi_2}{dt} \right|_{t=T} \quad (4.12)$$

Therefore, there exists $\varepsilon > 0$ such that for all $t \in [T, \varepsilon)$, $\phi_1(t) < \phi_2(t)$ which is a contradiction, unless it is the last value in I .

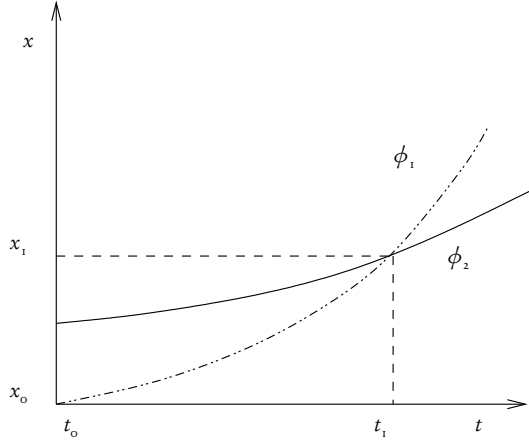


Figure 4.2: Proof of Lemma 4.1.

With the help of this Lemma, we will prove the uniqueness of the solution. Let $\varphi_1(t)$ be a solution with $\varphi_1(t_0) = x_0$ but different from the solution $\varphi(t) = x_0 \forall t$. Suppose then that there exists $t_1 > t_0$ such that $\varphi_1(t_1) = x_1 > x_0$, the other cases are treated similarly. [See figure 4.3.]

If we choose x_1 close enough to x_0 , it is fulfilled that

$$f(x) < k | x - x_0 | \quad \forall x_0 \leq x \leq x_1 \quad (4.13)$$

and therefore that $\varphi_2(t) < \varphi_1(t)$, $\forall t_0 \leq t \leq t_1$, where $\varphi_2(t)$ is the solution of the equation $\dot{x} = k(x - x_0)$ with the initial condition $\varphi_2(t_1) = x_1$. But $\varphi_2(t) = x_0 + (x_1 - x_0)e^{k(t-t_1)}$ which tends to x_0 only when $t \rightarrow -\infty$. Therefore, $\varphi_1(t)$ cannot take the value x_0 for any finite $t < t_1$ and therefore we have a contradiction ♠

The solution whose existence and uniqueness we have just demonstrated can not only be thought of as a map between I_{t_0} and U but also as a map g_x^t between $U_{x_0} \times I_{t_0}$ and U (U_{x_0} neighborhood of x_0). The map g_x^t takes (x, t) to the point $\varphi(t)$ where φ is the solution of 4.1 with the initial condition $\varphi(0) = x$.

These maps, or local diffeomorphisms, are called **local phase flows**, (since in general they cannot be defined in all $I_{t_0} \times U$) and satisfy the following important group property; if $t, s \in I_0$ then $g_x^{s+t} = g^s \circ g_x^t$.

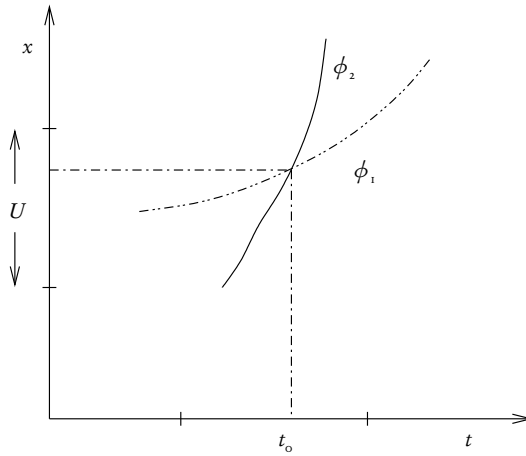


Figure 4.3: Proof of the uniqueness of the solution.

By the way the initial conditions appear in the previous Theorem, it is easy to see, using again the inverse function theorem, that if $f(x_0) \neq 0$ then g_x^t is also differentiable with respect to x .

Corollary 4.1 *if $f(x_0) \neq 0$ then $g_x^t : U_{x_0} \times I_{t_0} \rightarrow U$ is differentiable in both t and x .*

Note: The restriction $f(x_0) \neq 0$ is unnecessary as we will see later.

From the use of the inverse function theorem in Theorem 4.1, it also follows that if we consider the set of solutions distinguishing each one by its initial condition, $[\varphi(x_0, t_0, t)]$ is the solution that satisfies $\varphi(t_0) = x_0$, then the function $\varphi(x_0, t_0, t) : U \times I \times I \rightarrow U$ is differentiable with respect to all arguments. That is, if we slightly change the initial conditions, then the solution will only change slightly. Later we will give a proof of all the properties, including Theorem 4.1 for general systems, that is, equation 4.30.

Example: Let $x(t)$ be the number of bacteria in a culture vessel at time t . If the number is large enough, we can think of it as a smooth, differentiable function. If this number is not so large that its density in the vessel is low enough so that they do not have to compete for air and food, then their reproduction is such that $x(t)$ satisfies the equation,

$$\dot{x} = ax \quad a = \text{const.} \simeq 2.3 \frac{1}{\text{day}}, \quad (4.14)$$

that is, the growth rate is proportional to the number of bacteria present. The constant a is the difference between the reproduction rate and the extinction rate. Applying Theorem 4.1, we know that the equation allows us, given the number of bacteria

x_0 at time t_0 , to know the number of bacteria at any time. Indeed, the solution of 4.14 is

$$\frac{dx}{x} = a dt \quad (4.15)$$

Integrating both sides

$$\ln \frac{x}{x_0} = a(t - t_0) \quad \text{or}; \quad x(t) = x_0 e^{a(t-t_0)} \quad (4.16)$$

That is, the number of bacteria grows exponentially, unless of course $x_0 = 0$, and depends continuously on the initial number. It should be noted that although the dependence is continuous, the difference between solutions with different initial conditions grows exponentially.

The phase flow in this case is $g_x^t = x e^{at}$ and here it is global. With this growth, it is easy to realize that in a few days the number of bacteria and therefore their density will be so large that they will start to compete with each other for food, which will decrease the growth rate. It is experimentally observed that this fact can be taken into account by including in the 4.14 a term proportional to the square of the number of bacteria.

$$\dot{x} = ax - bx^2 \quad b = \frac{a}{375}. \quad (4.17)$$

If x is small, the first term dominates and x begins to grow, as we have seen, exponentially until the second term becomes important and the growth rate decreases, tending to zero when $x = x_s$ such that $ax_s - bx_s^2 = 0$ or $a - bx_s = 0$ or $x_s = a/b \simeq 375$ bacteria. Since $\varphi(t) = x_s$ is a solution, due to uniqueness, the solution that tends to x_s that we described above cannot reach the value x_s in finite time. If the number of bacteria present is greater than 375, the growth rate will be negative and the solution will again tend asymptotically to x_s .

The general solution of 4.17 can be obtained similarly to that of 4.14 and is,

$$x(t) = \frac{a/b}{1 + (\frac{a}{bx_0} - 1)e^{-a(t-t_0)}} \quad (4.18)$$

which confirms the previous analysis.

This example allows us to introduce two key concepts in the area. The first is that of a **stationary or equilibrium solution**. These are the constant solutions, that is, they are such that $\dot{x}_s = f(x_s) = 0$, that is, the roots of the function $f(x)$. For equation 4.14 $x_s = 0$, for equation 4.17 $x_s = 0$ and a/b .

Note that stationary solutions do not always exist, for example, the equation with $f(x) = 1$ does not have them, but when they do exist, the problem of finding them reduces to solving at most a transcendental equation. Due to the uniqueness of the solutions, this allows us to divide the plane (x, t) into strips such that if a solution has initial conditions in this strip, it must remain in it.

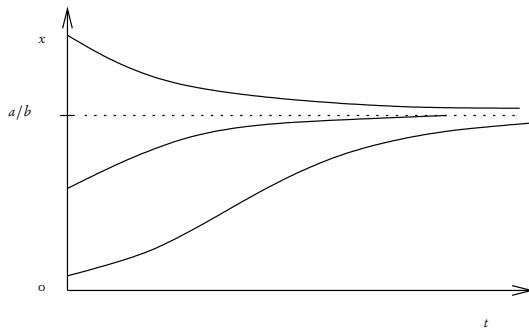


Figure 4.4: Different solutions of the bacterial growth equation: $\varphi_1(t_0) = 0$, $\varphi_2(t_0) = x_s = \frac{a}{b}$, $\varphi_3(t) = x_0 < x_s$ and $\varphi_4(t) = x_0 > x_s$

The second concept is the **stability** of these solutions. The solution $x_s = 0$ of 4.14 and 4.17 is not stable in the sense that if we choose as an initial condition a point in any neighborhood of this solution, the corresponding solution will initially move away exponentially from the previous one. On the contrary, the solution of 4.17 $x_s = a/b$ is stable in the sense that if we take initial values close to it, the corresponding solutions will asymptotically approach (in the future) the previous one.

Unstable solutions have no physical interest since the slightest perturbation of the system will generate a solution that quickly moves away from the unstable solution. This is not the case in the previous example since the number of bacteria is discrete and the equation only represents an approximation. If the culture vessel is sterilized and hermetically sealed, then there is no possibility of bacterial growth. Later we will analyze the problem of stability in detail.

4.2.2 Extending the Local Solution

If we forget about the physical model we described with 4.17 and take $z_0 < 0$, $t_0 = 0$, for example, there will be a maximum time, $t_d = \frac{1}{a} \ln \left[\frac{a/b - z_0}{-z_0} \right]$, finite for which the solution diverges, that is,

$$\lim_{t \rightarrow t_d} z(t) = \infty. \quad (4.19)$$

This shows that the previous theorem can only ensure the existence of an interval I for which the solution exists. The most we can conclude in general, that is, without giving more information about f , is the following Corollary of Theorem 4.1.

Corollary 4.2 *Let $U \subset \mathbb{R}$ be the domain of definition of f , U_c a compact interval [that is, closed and bounded] of U , $z_0 \in U_c$. Then the solution $(\varphi(z_0, t_0, t), I)$ of equation 4.1 can be extended either indefinitely or up to the boundary of U_c . This extension is unique in the sense that any pair of solutions, $(\varphi_1(z_0, t_0, t), I_1)$, $\varphi_2(z_0, t_0, t), I_2$ coincide in the intersection of their definition intervals, $I = I_1 \cap I_2$.*

Proof: First, we prove the global version of the uniqueness of the solution. Let T be the smallest of the upper bounds of the set $\{\tau | \varphi_1(z_o, t) = \varphi_2(z_o, t) \text{ for all } t_o \leq t \leq \tau\}$. Suppose in contradiction that T is an interior point of $I_1 \cap I_2$. By continuity we have, $\varphi_1(z_o, T) = \varphi_2(z_o, T)$, but by the result of Theorem 4.1, using initial conditions $\varphi_1(z_o, T)$, at T we see that both solutions must coincide in a neighborhood of T , which contradicts that T is the maximum. Thus, T must be an endpoint of one of the definition intervals, so the solutions coincide in $I_1 \cap I_2 \cap \{t | t \geq t_o\}$. The case $t \leq t_o$ is treated similarly, and therefore the solutions coincide in all $I_1 \cap I_2$.

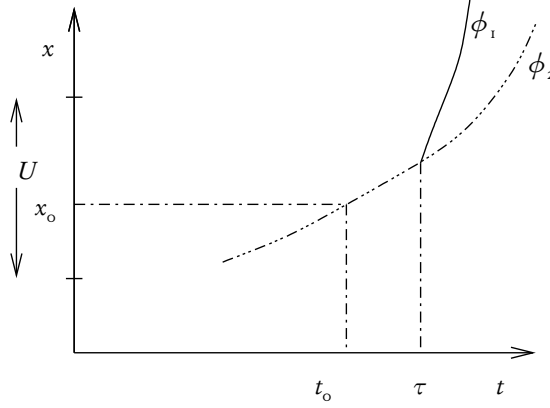


Figure 4.5: Global uniqueness.

Second, we construct the extension. The idea is that if two solutions coincide in $I_1 \cap I_2$, then we can combine them and form one in $I_1 \cup I_2$,

$$\varphi(t) = \begin{cases} \varphi_1(t) & \text{for all } t \in I_1 \\ \varphi_2(t) & \text{for all } t \in I_2. \end{cases} \quad (4.20)$$

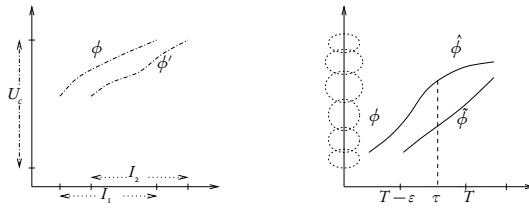


Figure 4.6: Extending the solution.

Let T be the smallest upper bound of $\{\tau | \varphi(t) \text{ exists and is contained in } U_c \text{ for all } t_o \leq t \leq \tau\}$. If $T = \infty$ there is nothing to prove, so suppose $T < \infty$. We will prove that there exists $\varphi(t)$ defined for all $t_o \leq t \leq T$ and such that $\varphi(T)$ is one of the endpoints

of U_c . Corollary 4.1 tells us that given any $\hat{z} \in U$, there exist $\epsilon_{\hat{z}} > 0$ and a neighborhood of \hat{z} , $V_{\hat{z}} \subset U$ such that for any $z \in V_{\hat{z}}$ there exists a solution with initial conditions $\varphi(t_o) = z$ defined for all t in $|t - t_o| < \epsilon_{\hat{z}}$. Since U_c is compact, we can choose a finite number of points \hat{z}_i such that $U_c \subset \bigcup V_{\hat{z}_i}$. Let $\epsilon > 0$ be the minimum of the $\epsilon_{\hat{z}_i}$. Since T is the smallest upper bound, there exists τ between $T - \epsilon$ and T such that $\varphi(t) \in U_c$ for all $t_o \leq t \leq \tau$, but since $\varphi(\tau) \in U_c$ is in one of the $V_{\hat{z}_i}$, there exists a solution $\hat{\varphi}(t)$ defined for all t in $|t - \tau| < \epsilon$ with initial condition $\hat{\varphi}(\tau) = \varphi(\tau)$. Using $\hat{\varphi}(t)$ we can now extend φ , whose extension we will call $\tilde{\varphi}$, to the interval $t_o \leq t < \tau + \epsilon$. But $\tilde{\varphi}(\theta) \in U_c$ for all $t_o < \theta < T$ since $\tilde{\varphi}(\theta) = \varphi(\theta) \in U_c$, and otherwise T would not be a smallest upper bound. By continuity $\tilde{\varphi}(T) \in U_c$ and as by definition of T for any open interval, I_T , containing T , $\tilde{\varphi}[I_T] \not\subset U_c$ we conclude that $\tilde{\varphi}(T)$ is an endpoint of U_c ♠

4.2.3 The Non-autonomous Case

This is the case where instead of 4.1 we have

$$\frac{dz}{dt} = f(z, t). \quad (4.21)$$

The geometric interpretation is the same, only now the *little lines* have an angle that also depends on the variable t , therefore we expect that if we start at a point (t_o, z_o) and trace a curve tangent to the little lines, we will also obtain a unique solution. This is indeed the case, but its discussion will be included in the general theory of first-order autonomous systems that we will state later.

Assuming certain conditions for $f(z, t)$ – for example, that it is Lipschitz with respect to both variables – it can be seen that local flows also exist, but these now also depend on t_o and not just on the difference between the initial and final time. We will denote these flows as $g_{t_o}^t(z)$ and the semigroup property they satisfy is $g_{t_i}^t \circ g_{t_o}^{t_i}(z) = g_{t_o}^t(z)$.

There is a special class of non-autonomous equations that can be reduced to the already studied case. This is the class where $f(z, t)$ can be written as the quotient of two functions, one of z and another of t ,

$$f(z, t) = \frac{g(z)}{h(t)}. \quad (4.22)$$

In this case, the geometric interpretation is that the parameter t is not the most suitable to describe this system. This is remedied by choosing another parameter τ given by, $\frac{d\tau}{dt} = \frac{1}{h(t)}$, or

$$\tau - \tau_o = \int_{t_o}^t \frac{dt}{h(t)}. \quad (4.23)$$

Once $t(\tau)$ is obtained [See Theorem 4.1], we must solve,

$$\frac{dz}{d\tau} = g(z(\tau)), \quad (4.24)$$

[Using Theorem 4.1 again], obtaining $\varphi(\tau)$. The solution with respect to the original parameter is obtained by defining $\varphi(t) = \varphi(\tau(t))$.

4.3 Reduction to First-Order Systems

Definition: The **general ordinary differential equation of order m** is an equation of the form:

$$F(x, x^{(1)}, \dots, x^{(m)}, t) = 0 \quad (4.25)$$

That is, an implicit equation of x , a map between $\tilde{I} \subset \mathbb{R}$ and $\tilde{U} \subset \mathbb{R}^{m \times n}$, (i.e., x is a vector of n components), and its derivatives (where $x^{(i)} \equiv \frac{d^i x}{dt^i}$) denotes the i -th derivative with respect to the independent parameter, t) up to order n^2 .

Definition: We will say that the m -times differentiable map $x(t) : I \rightarrow \mathbb{R}^n$ is a **solution** of the previous equation if F evaluated in this map is well-defined and identically zero over the entire interval I .

In this course, we will assume that 4.25 can be solved for $x^{(m)}$, that is, that 4.25 is equivalent to the following equation:

$$x^{(n)} = f(x, x^{(1)}, \dots, x^{(n-1)}, t) \quad (4.26)$$

in open sets $I \subset \tilde{I} \subset \mathbb{R}$ and $U \subset \tilde{U} \subset \mathbb{R}^m$

Example: The ODE $F(x, x^{(1)}, t) = ((x^{(1)})^2 - 1) = 0$ implies one of the following ODEs in our sense:

$$\begin{aligned} x^{(1)} &= 1 \\ \text{or} \\ x^{(1)} &= -1 \end{aligned} \quad (4.27)$$

If we know the values of x and its derivatives up to the $(m-1)$ -th order at a point t_0 , then equation 4.26 allows us to know the m -th derivative at that point. If we assume that f is continuous, then this allows us to know x and its derivatives up to the $(m-1)$ -th order at a point $t_0 + \varepsilon$ sufficiently close (the error incurred will be of the order of $\varepsilon \times \text{derivatives of } f$), but then we can use equation 4.26 again and thus approximately know the n -th derivative of x at $t_0 + \varepsilon$. Proceeding in this way, we can achieve an *approximate solution* in a given interval. At least intuitively, it would seem that by making the interval ε smaller and smaller, we should obtain a solution. This is indeed the case! And we will prove it later in the course. The important thing for now is the fact that to obtain a solution, we must give at a point t_0 the value of all the derivatives of the unknown variable of order less than the one that determines the system, that is, the one that appears on the left side of equation 4.26.

²In reality, to correctly define 4.25, the open sets \tilde{U}^i for the i -th derivative with respect to x where F is defined must be given.

Introducing $\tilde{u}^1 \equiv x$, $\tilde{u}^2 \equiv x^{(1)}$, \dots , $\tilde{u}^n \equiv x^{(m-1)}$, we can write 4.26 in the form:

$$\begin{aligned}\dot{\tilde{u}}^1 &= \tilde{u}^2 \\ \dot{\tilde{u}}^2 &= \tilde{u}^3 \\ &\vdots \\ \dot{\tilde{u}}^m &= f(\tilde{u}^1, \tilde{u}^2, \dots, \tilde{u}^m, t)\end{aligned}\tag{4.28}$$

where we have denoted with a dot over the function its derivative with respect to the parameter t . Therefore, 4.26 is equivalent to a first-order equation for the vector of vectors $\tilde{u} \subset U \subset \mathbb{R}^{n \times m}$,

$$\dot{\tilde{u}} = \tilde{V}(\tilde{u}, t)\tag{4.29}$$

where \tilde{V} is also a vector in some subset of $\mathbb{R}^{n \times m}$. If $\tilde{V}(\tilde{u}, t)$ depends explicitly on t , then we can add to \tilde{u} another component, the $(n \times m + 1)$ -th with $\tilde{u}^{n \times m + 1} = t$, and therefore we have that $\dot{\tilde{u}}^{n \times m + 1} = 1$. Equation 4.29, plus this last equation, take the form,

$$\dot{u} = V(u)\tag{4.30}$$

where $u = (\tilde{u}, \tilde{u}^{n \times m + 1})$ and

$$V^i(u) = \begin{cases} \tilde{V}^i(\tilde{u}, \tilde{u}^{n \times m + 1}) & \text{if } 1 \leq i \leq m \times n \\ 1 & \text{if } i = n \times m + 1 \end{cases},\tag{4.31}$$

a map between $U \subset \mathbb{R}^{n \times m + 1}$ and $\mathbb{R}^{n \times m + 1}$.

Thus, we arrive at the following theorem.

Theorem 4.2 (Reduction) *If the function f of the system 4.26 of m ordinary differential equations of order n is differentiable, then this system is equivalent to a system of $m \times n + 1$ first-order ordinary differential equations that do not explicitly depend on the independent variable. That is, for each solution of the system 4.26, we can construct a solution of the system 4.30 and vice versa.*

Proof: Clearly, given a sufficiently differentiable solution of the system 4.26, we can write with it a vector u and it will satisfy (because we constructed it this way) the corresponding system 4.30. Therefore, every solution of the original system is a solution of the associated first-order system.

To prove the inverse, that is, that every solution $u(t)$ of the system 4.30 gives rise to a solution of the system 4.26, it is only necessary to take repeated derivatives of the first components, $x(t) = u^1(t)$, and use the equations in the corresponding order until obtaining the original system satisfied by the n th derivative of $x(t)$ ♠

This theorem is important because it allows us to encompass the entire theory of ODEs by studying equation 4.30, which, as we will see, has a clear geometric meaning. However, it should be noted that many times in physics there appear very special subclasses of equations with particular properties of great physical importance that the general theory does not contemplate.

Example: Mathematical Pendulum

$$\ddot{x} = kx \quad u^1 = x, u^2 = \dot{x}$$
$$\frac{d}{dt} \begin{pmatrix} u^1 \\ u^2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ k & 0 \end{pmatrix} \begin{pmatrix} u^1 \\ u^2 \end{pmatrix} \quad (4.32)$$

Note that if $u = \gamma(t) : I \rightarrow U$ is a solution of 4.30 then $\forall s \in \mathbb{R}$ such that $t+s \in I$, $\gamma(t+s)$ is also a solution. Indeed,

$$\left. \frac{d}{dt} \gamma(t+s) \right|_{t=t_0} = \left. \frac{d}{dt} \gamma(t) \right|_{t=t_0+s} = V(\gamma(t_0+s)) = V(\gamma(t+s))|_{t=t_0}. \quad (4.33)$$

Due to this, from now on we will take t_0 , the point where we give the initial condition, equal to zero.

Exercise: If $u = \sin(t) : \mathbb{R} \rightarrow [0, 1]$, is a solution of 4.30 prove that $u = \cos(t)$ is also a solution.

4.4 ODE Systems

Do ODE systems have a geometric interpretation? As we have seen, they have the generic form:

$$\frac{dx^i}{dt} = v^i(x^j) \quad i = 1, \dots, n. \quad (4.34)$$

Let $\gamma(t)$ be a curve between an interval $I \in \mathbb{R}$ and M a manifold of n dimensions. Let $p = \gamma(t_0)$ be a point of M and $(U, \varphi = (x^1, \dots, x^n))$ a chart with $p \in U$. Then $\varphi \circ \gamma(t) = (x^1(t), \dots, x^n(t))$ is a map between I and an open set of \mathbb{R}^n , and $\frac{dx^i}{dt}$ are the components of the tangent vector to γ in the coordinate basis $\{x_i\}$ at the point $\gamma(t)$ of M . Therefore, $v^i(x^j)$ are the components of a vector field v in the coordinate basis $\{x_i\}$ at the point $p \in M$ corresponding to $\varphi(p) = (x^1, \dots, x^n)$. Thus, we see that ODE systems can be interpreted as a vector field v on a manifold M , and their solutions as curves $\gamma(t)$ such that their tangent at every point is v .

Example: The Physical Pendulum

$$\begin{cases} \ddot{\theta} &= -\sin \theta \\ \dot{\theta} &= z \\ \dot{z} &= -\sin \theta \end{cases} \quad (4.35)$$

Here the vector field is $z \frac{\partial}{\partial \theta} - \sin \theta \frac{\partial}{\partial z}$,

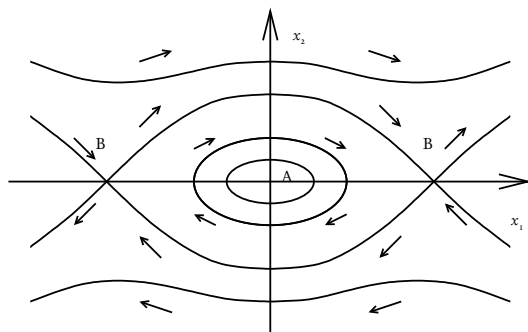


Figure 4.7: The Physical Pendulum.

note that the diagram in the plane repeats every 2π along the θ axis. In reality, the equation makes physical sense on a cylinder, with θ being the angular variable, so the manifold is a cylinder. The **integral curves**, i.e., the images of the maps $\gamma(t)$ that are solutions, are constructed by following the vectors so that they are always tangent to the curves.

From the example, it is clear that in general, if we give a point of M and follow the vectors from it, we can find a solution $\gamma(t)$ that passes through this point. If we have two solutions $\gamma_1(t), \gamma_2(t)$, they cannot cross each other – without being tangent – at points where the vector field is non-zero [since otherwise, their tangents would give two different vectors at the point]. If we assume that the vector field is differentiable, then it can be seen that different curves can never touch, and therefore the solutions are unique.

Of course, as seen for example in figure 4.8, the same curve can intersect itself (note that this can only happen if the system is autonomous, as otherwise, the inclusion of the temporal variable prevents any curve from intersecting itself). In such a case, it can be proven that this solution is **periodic**, meaning that there exists $T > 0$ such that $\gamma(t + T) = \gamma(t)$. Note in particular that this implies that γ is defined for all t in \mathbb{R} .

Problem 4.1 *If M is a compact manifold (example: a sphere or a torus), it can be proven using Corollary 4.2 that all its phase curves are defined for all t . However, there are some that are not closed, give an example.*

4.4.1 First Integrals

Given a vector field v on M , if there exists $f : M \rightarrow \mathbb{R}$ non-constant such that $v(f) = 0$ on M , then we will say that f is a **first integral or constant of motion** of the ODE system generated by v .

It can be shown that if n is the dimension of M , then locally, around points where

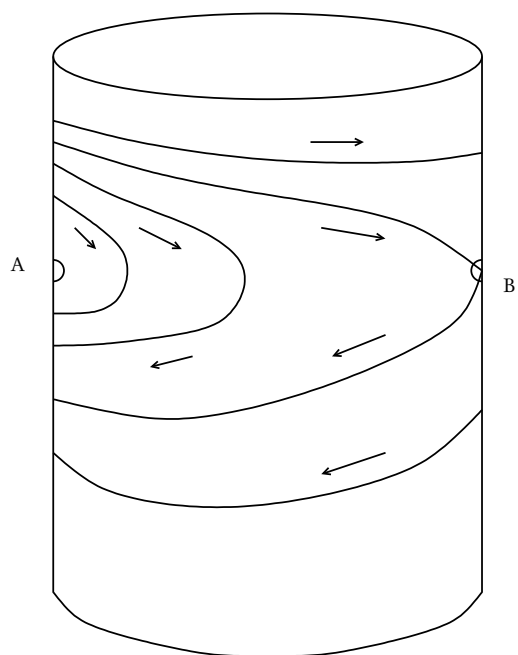


Figure 4.8: The Physical Pendulum Manifold.

$v \neq 0$, there exist $n - 1$ functionally independent first integrals. However, it is easy to find vector fields that do not have any first integrals.

Examples-Problems

Problem 4.2 *The system*

$$\begin{cases} \dot{x}_1 &= x_1 \\ \dot{x}_2 &= x_2 \end{cases}$$

has none, why? Hint: Draw the integral curves.

Problem 4.3 *Consider the system*

$$\begin{cases} \dot{x}_1 &= k_1 \\ \dot{x}_2 &= k_2 \end{cases}$$

on the torus (obtained by identifying the points (x_1, x_2) and $(x_1 + n, x_2 + m)$ of \mathbb{R}). For which values of k_1, k_2 does a first integral exist and for which does it not? Hint: same as 1).

Problem 4.4 *The Hamilton equations of classical mechanics form an ODE system,*

$$\dot{q}_i = \frac{\partial H}{\partial p_i} \tag{4.36}$$

$$\dot{p}_i = -\frac{\partial H}{\partial q_i} \tag{4.37}$$

$i = 1, \dots, n$, where $H : \mathbb{R}^{2n} \rightarrow \mathbb{R}$ is the Hamiltonian function or energy. Show that H is a first integral of this system.

The previous examples show that there are ODE systems that globally are not derivable from a variational principle.

Just as knowing the energy of a Hamiltonian system is an invaluable aid in understanding the quantitative form of its motion, knowing a first integral of a general ODE system also provides great help since we know that the phase curves are restricted to their level surfaces ($f = \text{constant}$), which allows us to effectively reduce the order of the equation by one: we only have to consider the problem on the manifold given by the points $f = \text{constant}$.

4.4.2 Fundamental Theorem of ODE Systems

Theorem 4.3 *Let v be a continuously differentiable (C^1) vector field in a neighborhood of $p \in M$. Then,*

i) There exists a neighborhood U_p of p and an open set $I \subset \mathbb{R}$, such that given any $q \in U_p$ ($t_o \in I$), there exists a differentiable curve $\gamma_q(t) : I \rightarrow M$ satisfying

$$\begin{aligned} \left. \frac{d\gamma_q}{dt} \right|_t &= \mathbf{v}|_{\gamma_q(t)} \\ \gamma_q(t_o) &= q. \end{aligned} \quad (4.38)$$

ii) If γ_q^1 and γ_q^2 satisfy condition i), then $\gamma_q^1 = \gamma_q^2$ in $I^1 \cap I^2$.

iii) $\gamma_q(t)$ is also C^1 when considered as a function from $I \times U_p \rightarrow M$. That is, it is continuously differentiable with respect to the initial condition.

iv) Let U be a compact (bounded) region of M , $\mathbf{v} \in C^1$, and $p \in U$, then $\gamma_p(t)$ can be extended (forward and/or backward) to the boundary of U . In particular, if M is compact and $\mathbf{v} \in C^1(M)$, then $\gamma_p(t)$ exists globally (for all $t \in \mathbb{R}$).

We will prove this theorem later, after introducing the necessary mathematical tools.

4.4.3 Parameter Dependence, Variation Equation

A very important consequence of the fundamental theorem in applications is the following:

Corollary 4.3 Let \mathbf{v}_λ be a differentiable vector field in $U \subset M$ that depends differentiably on a parameter $\lambda \in A \subset \mathbb{R}$. Then given $p \in U$, $t_o \in I$, and $\lambda_o \in A$, there exist neighborhoods U_p , I_{t_o} , and A_{λ_o} such that for any triplet $(q, t, \lambda) \in U_p \times I_{t_o} \times A_{\lambda_o}$, there exists a unique integral curve of \mathbf{v}_λ , $\gamma_\lambda(t) : I_{t_o} \times A_{\lambda_o} \rightarrow U_p$ with $\gamma_\lambda(o) = q$. This depends differentiably on q , t , and λ .

Proof: Consider the vector field $(\mathbf{v}_\lambda, o) : U \times A \rightarrow TM \times \mathbb{R}$. By hypothesis, this field is differentiable and therefore has integral curves $(\gamma_\lambda, \lambda)$ that depend differentiably on the initial conditions and therefore on λ ♠

Warning: The interval $I_{t_o} \times A_{\lambda_o}$ where the solution is differentiable is not necessarily the interval of definition of the solution. For example, it is not true that even when the solution is defined for all $t \in \mathbb{R}$, the limits $\lim_{T \rightarrow \infty} \gamma_\lambda(t)$ and $\lim_{\lambda \rightarrow \lambda_o} \gamma_\lambda(t)$ commute.

The practical importance of this corollary is that it allows us to obtain approximate solutions to already known ones. Indeed, suppose that for $\lambda = o$ we know some integral curve of $\mathbf{v}|_{\lambda=o}$, $\gamma_o(t)$ with $\gamma_o(o) = p$. Then, applying the Corollary and a Taylor series expansion, we have

$$\gamma_\lambda(t) = \gamma_o(t) + \lambda \gamma_i(t) + O(\lambda^2) \quad (4.39)$$

that is, $\gamma_o(t) + \lambda \gamma_i(t)$ is a good approximation to $\gamma_\lambda(t)$ for λ sufficiently small.

It remains to find the equation that $\gamma_1(t)$ satisfies. This is obtained by differentiating with respect to λ the differential equation at $\lambda = 0$,

$$\frac{d}{d\lambda} \left(\frac{d\gamma_\lambda(t)}{dt} = v_\lambda(\gamma_\lambda(t)) \right) \Big|_{\lambda=0}, \quad (4.40)$$

expanding and using coordinates

$$\begin{aligned} \frac{d}{dt} \gamma_1^j &= \frac{\partial v^j}{\partial x^i} \Big|_{(\gamma_0, \lambda=0)} \cdot \gamma_1^i + \frac{dv^j}{d\lambda} \Big|_{(\gamma_0, \lambda=0)}, \\ &= A^j_i(t) \gamma_1^i + b^j(t). \end{aligned} \quad (4.41)$$

that is, the equation for $\gamma_1(t)$ –usually called the variation equation– is a linear inhomogeneous non-autonomous equation. Since we took the initial condition $\gamma_\lambda(0) = p \quad \forall \lambda$, the initial condition we will consider for 4.4.3 is $\gamma_1(0) = 0$.

Problem 4.5 *What would be the initial condition for γ_1 if the condition for γ_λ were $\gamma_\lambda(0) = p(1 - \lambda) + \lambda q$? Consider only the one-dimensional case but think about the general case.*

Problem 4.6 *If we decided to consider an approximation up to the second order $\gamma_\lambda = \gamma_0 + \lambda \gamma_1 + \lambda^2 \gamma_2 + O(\lambda^3)$, what equations would we obtain for γ_2 ?*

Example: a) A body falls vertically in a low-viscosity medium, which depends on its position and velocity.

$$\ddot{x} = -g + \varepsilon F(x, \dot{x}) \quad \varepsilon \ll 1 \quad (4.42)$$

Calculate the effect of this resistance on the motion.

The solution, $x(t)$, will depend smoothly on ε , so we can expand it in a Taylor series with respect to ε .

$$x(t) = x_0(t) + \varepsilon x_1(t) + O(\varepsilon^2). \quad (4.43)$$

The solution for $\varepsilon = 0$ is clearly

$$x_0(t) = x_i + v_i t - g \frac{t^2}{2}, \quad (4.44)$$

where x_i and v_i are the initial position and velocity, respectively, while the variation equation is

$$\ddot{x}_1 = F(x_0, \dot{x}_0), \quad x_1(0) = 0, \quad \dot{x}_1(0) = 0. \quad (4.45)$$

Integrating, we obtain

$$x_1(t) = \int_0^t \int_0^\tau F(x_0(\tilde{\tau}), \dot{x}_0(\tilde{\tau})) d\tilde{\tau} d\tau. \quad (4.46)$$

b) The second example is self-oscillations.
Consider the system,

$$\begin{aligned}\dot{x}_1 &= x_2 + \varepsilon f_1(x_1, x_2) \\ \dot{x}_2 &= -x_1 + \varepsilon f_2(x_1, x_2), \quad \text{for } \varepsilon \ll 1.\end{aligned}\tag{4.47}$$

Without loss of generality, we can assume $f_i(0,0) = 0$. When $\varepsilon = 0$, we obtain the equations of the pendulum –in the small amplitude approximation–, that is, a conservative system. Its energy –first integral– is given by $E(x_1, x_2) = \frac{1}{2}(x_1^2 + x_2^2)$ and its phase curves are then circles of radius $\sqrt{2E}$.

When $\varepsilon > 0$, the phase curves are not necessarily closed, but due to the previous Corollary, they can at most be spirals with a distance of the order of ε between turns. These can tend towards some stationary point within the circle $\sqrt{2E_i}$, where $E_i =$ initial energy, that is, where $x_2 + \varepsilon f_1(x_1, x_2) = -x_1 + \varepsilon f_2(x_1, x_2) = 0$, or they can asymptotically approach some closed orbit or finally diverge. That these are the only options is the result of a classic theorem by **Poincaré–Bendixson**, which shows that in general these are the only possibilities in two dimensions.

To study which case it is among these three, we consider the variation of energy between two consecutive turns.

Using that,

$$\dot{E}(x_1, x_2) = \varepsilon (x_1 f_1 + x_2 f_2)\tag{4.48}$$

we obtain

$$\begin{aligned}\Delta E &= \varepsilon \oint (-f_1 dx_2 + f_2 dx_1) + O(\varepsilon^2) \\ &= \varepsilon F(E).\end{aligned}\tag{4.49}$$

where the integral is along a circle of radius $\sqrt{2E}$ in the direction of motion. That is, we have approximated the curve with $\varepsilon > 0$ by the curve with $\varepsilon = 0$.

If $F(E)$ is positive, the energy increases and the system undergoes growing oscillations.

If $F(E)$ is negative, the oscillations decrease in amplitude and the system eventually tends to an equilibrium point.

If $F(E)$ changes sign, say at $E = E_0$, it can be proven that near this circle there is a closed phase curve Γ_{E_0} .

If $F'(E)|_{E=E_0}$ is negative, Γ is a stable limit cycle –any nearby phase curve asymptotically approaches it–. If $F'(E)|_{E=E_0}$ is positive, Γ is unstable.

Problem 4.7 *Show that if there are two stable limit cycles, then there must also be at least one that is unstable.*

From the two previous examples, the practical importance of the Corollary is clear. The differentiable dependence on the initial conditions also leads to the variation equation, therefore this equation, which is linear and inhomogeneous, is of utmost importance and in the next chapter, we will study it in more detail.

4.5 Problems

Problem 4.8 (Kiseliöv) Find all the solutions of

$$\frac{dx}{dt} = \frac{3}{2}x^{2/3}. \quad (4.50)$$

Hint: there are infinitely many, and they are obtained from segments of some particular solutions. Graph some of these solutions.

Problem 4.9 (Kiseliöv) Apply the existence and uniqueness theorem to determine the regions where the following equations have a unique solution:

a) $\dot{x} = x + 3x^{1/3}$.

b) $\dot{x} = 1 - \cot x$.

c) $\dot{x} = \sqrt{1 - x^2}$.

Problem 4.10 (Kiseliöv) Solve the following equations, in all cases give the general solutions as a function of an initial value x_0 for the initial time t_0 .

a) $t\dot{x} + x = \cos(t)$. (Use the first integral of the homogeneous part.)

b) $\dot{x} + 2x = e^t$. (Use variation of constants.)

c) $(1 - t^2)\dot{x} + tx = 2t$. (First solve the homogeneous part and then add a particular inhomogeneous solution.)

d) $x\dot{x} = t - 2t^3$.

Problem 4.11 (Kiseliöv) Solve the following equations by making a change of variable in the independent variable (t).

a) $\dot{x}x = -t$. ($t = \cos(s)$.)

b) $\dot{x}t - x = 0$.

c) $\dot{x} + e^{\dot{x}} = t$.

Problem 4.12 (Kiseliöv) Graph the isoclines (lines of equal slope in the (t, x) plane) and then trace the solutions of the following equations:

a) $\dot{x} = 2t - x$.

b) $\dot{x} = \sin(x + t)$.

c) $\dot{x} = x - t^2 + 2t - 2$.

d) $\dot{x} = \frac{x-t}{x+t}$.

Problem 4.13 Solve the equation:

$$\dot{x} = A(t)x + B(t)x^n \quad \text{Hint: Use the change of variable } y = x^{1-n}.$$

Problem 4.14 Solve the equation

$$\frac{dx}{dt} = i\lambda x + Ae^{i\omega t} \quad \lambda, \omega \in \mathbb{R} \quad (4.51)$$

and see how its real part behaves. Examine the cases:

a) $A = 0$, b) $A \neq 0$, $\lambda \neq \omega$ and c) $A \neq 0$, $\lambda = \omega$.

Problem 4.15 *The equation of the physical pendulum.*

a) Graph the vector field of the equation

$$\frac{d^2\theta}{dt^2} = -\sin(\theta), \quad -\pi \leq \theta \leq \pi. \quad (4.52)$$

b) Graph some integral curves around $(\theta = 0, z = \frac{d\theta}{dt} = 0)$.

c) Graph the integral curves that pass through the point $\theta = \pm\pi, z = 0$. Infer that the time required to reach this point along these curves is infinite.

d) Graph the vector field along a line $z = z_0$. Infer from this that the solutions can never exceed the speed acquired when passing through the point $\theta = 0$. Hint: Conclude this first for the region $0 \leq \theta \leq \pi$, $0 \leq z \leq z_0$ and then use the symmetry of the solution.

e) Let $E(z, \theta) := \frac{z^2}{2} - \cos(\theta)$. See that $\frac{dE}{dt} = 0$. Use this quantity to analyze the behavior of the solutions with initial data $(z_0, \theta_0 = 0)$. In particular, determine which cross the $z = 0$ axis and which do not.

Problem 4.16 *Consider the equation:*

$$\frac{dz}{dt} = \begin{cases} f(z) & |z| < 1 \\ iz & |z| \geq 1 \end{cases} \quad (4.53)$$

Where $f(z)$ is continuous with iz at $|z| = 1$. Infer that no matter the form of $f(z)$, the solutions never escape the region $\{|z(t)| \leq \max\{|z(0)|, 1\}\}$.

Problem 4.17 *Consider a body affected by a central force, that is,*

$$m \frac{d^2 \vec{x}}{dt^2} = f(r) \vec{x} \quad r := |\vec{x}|. \quad (4.54)$$

a) Find an equivalent first-order system.

b) Verify that given any (constant) vector \vec{c} , then $F(\vec{x}, \vec{p}) := \vec{c} \cdot (\vec{x} \wedge \vec{p})$, where $\vec{p} := \frac{d\vec{x}}{dt}$, is a first integral.

Problem 4.18 *Consider the equation:*

$$\frac{d\vec{x}}{dt} = \vec{x} \wedge \vec{\omega}(\vec{x}). \quad (4.55)$$

a) See that $R := \vec{x} \cdot \vec{x}$ is a first integral.

b) Conclude that the solutions of this system exist for all time.

Problem 4.19 (Strichartz) *Show that*

$$J_k(t) = \sum_{j=0}^{\infty} \frac{(-1)^j (t/2)^{k+2j}}{j!(k+j)!}, \quad (4.56)$$

has an infinite radius of convergence and satisfies the Bessel equation, $x''(t) + (1/t)x'(t) + (1 - k^2/t^2)x(t) = 0$.

Problem 4.20 Solve the system:

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -x_2 + \varepsilon(x_1^2 + x_2^2) \\ x_1 \end{pmatrix} \quad (4.57)$$

For the initial data $(x_1(0) = 1, x_2(0) = 0)$. Now investigate the solution near $\varepsilon = 0$ by obtaining the coefficients in its Taylor series with respect to the parameter ε . Find which equations these coefficients satisfy.

Problem 4.21 (Arnold) Consider the equation of the physical pendulum:

$$\frac{d^2 x}{dt^2} = -\sin(x).$$

See how the frequency varies as a function of the amplitude for small solutions. *Hint: assume a solution in a Taylor series of the amplitude and solve until finding the first non-trivial contribution to the linearized solution. Find the period by locating the return points (zero velocity).*

Bibliography notes: For this and the next three chapters, I recommend the following books: [3], [7], and [8]. Especially the first one. Ordinary differential equations are the basis of classical mechanics, although in the latter there is also a particular geometric structure that is fundamental. Almost everything we have seen in this chapter is the *local* theory, the *global* theory, that is, the behavior of the solutions for large values of the independent variable (time in most cases) has only been understood in recent years, giving rise to new concepts such as chaos, attractors, fractal dimensions, etc. Unfortunately, these aspects have not yet been sufficiently synthesized, and therefore I cannot recommend any book on the subject. There are too many interesting ideas, but each of them requires a great deal of information.

5.1 Homogeneous Linear System

Theorem 5.1 *Let $A(t)$ be continuous on $I \subset \mathbb{R}$, closed. Then there exists a unique solution $x(t): I \rightarrow V$ of the equation,*

$$\frac{dx(t)}{dt} = A(t)x(t), \quad (5.1)$$

with initial condition $x(t_0) = x_0 \in V$, $t_0 \in I$.

Proof: Let $x, y \in V$, then

$$\|A(t)x - A(t)y\|_V \leq \|A(t)\|_{\mathcal{L}} \|x - y\|_V \quad (5.2)$$

and therefore $A(t)x$ is Lipschitz in V . The fundamental theorem guarantees local existence and uniqueness. The theorem will be proven if we show that $x(t)$ cannot become infinite in finite time. For this, we use,

$$\begin{aligned} \frac{d}{dt} \|x\|_V &= \lim_{t_1 \rightarrow t} \frac{\|x(t_1)\|_V - \|x(t)\|_V}{t_1 - t} \leq \lim_{t_1 \rightarrow t} \frac{\|x(t_1) - x(t)\|}{t_1 - t} \\ &= \left\| \frac{dx}{dt} \right\|_V = \|Ax\|_V \leq \|A\|_{\mathcal{L}} \|x\|_V. \end{aligned} \quad (5.3)$$

Integrating this inequality between t_0 and $t \in I$ we obtain,

$$\|x(t)\|_V \leq \|x(t_0)\|_V e^{\int_{t_0}^t \|A(\tilde{t})\|_{\mathcal{L}} d\tilde{t}}, \quad (5.4)$$

which shows that $x(t)$ cannot escape to infinity in finite time and therefore completes the proof.

Consider the set of solutions of 5.1 defined on a single closed interval $I \subset \mathbb{R}$, $Sol(A, I)$. This set has the structure of a vector space. Indeed, if $x(t)$ and $y(t)$ are two solutions of 5.1, then $x(t) + \alpha y(t)$, $\alpha \in \mathbb{R}$, is also a solution. What is its dimension? The following theorem answers this question and shows that any solution of 5.1 can be expressed as a linear combination of a finite number $(n) = \dim V$ of solutions.

Theorem 5.2 $\dim \text{Sol}(A, I) = \dim V$.

Proof: Let $\{\mathbf{u}_i^\circ\}$, $i = 1, \dots, n$ be a basis of V , $t_o \in I$ and $\{\mathbf{u}_i(t)\}$, $i = 1, \dots, n$ $t \in I$ the set of solutions of 5.1 with initial condition $\mathbf{u}_i(t_o) = \mathbf{u}_i^\circ$. We will show that $\{\mathbf{u}_i(t)\}$ form a basis of $\text{Sol}(A, I)$, and therefore $\dim \text{Sol}(A, I) = \dim V$. First, let's see that the solutions $\{\mathbf{u}_i(t)\}$ are linearly independent.

Suppose there are constants c^i such that $\mathbf{x}(t) = \sum_{i=1}^n c^i \mathbf{u}_i(t)$ is zero for some \tilde{t} in I . Since $\mathbf{x}(t)$ is a solution of 5.1 and these are unique when an initial condition is specified, taking in this case $\mathbf{x}(\tilde{t}) = \mathbf{o}$, we see that $\mathbf{x}(t) = \mathbf{o} \forall t \in I$. In particular, we have that $\mathbf{x}(t_o) = \sum_{i=1}^n c^i \mathbf{u}_i^\circ = \mathbf{o}$ and the independence of the set $\{\mathbf{u}_i^\circ(t)\}$ implies that $c^i = 0 \forall i = 1, \dots, n$, proving linear independence. Note that we have not only proven that $\{\mathbf{u}_i(t)\}$ are linearly independent as elements of $\text{Sol}(A, I)$ – for which we would only have had to prove that if $\sum_{i=1}^n c^i \mathbf{u}_i(t) = \mathbf{o} \forall t \in I$ then $c^i = 0 \forall i = 1, \dots, n$ which is trivial since $t_o \in I$ – but also that for each $t \in I$ the $\{\mathbf{u}_i(t)\}$ are linearly independent as elements of V , this result will be used later.

To complete the theorem, let's now see that any solution of 5.1 $\mathbf{x}(t) : I \rightarrow V$, that is, any element of $\text{Sol}(A, I)$ can be written as a linear combination of $\{\mathbf{u}_i(t)\}$. Let $\mathbf{x}(t) \in \text{Sol}(A, I)$, since $\{\mathbf{u}_i^\circ\}$ form a basis of V there will be constants c^i , $i = 1, \dots, n$ such that $\mathbf{x}(t_o) = \sum_{i=1}^n c^i \mathbf{u}_i^\circ$. Consider then the solution of 5.1 given by $\varphi(t) = \sum_{i=1}^n c^i \mathbf{u}_i(t)$. Since $\varphi(t_o) = \mathbf{x}(t_o)$ and the solutions are unique, we have,

$$\mathbf{x}(t) = \varphi(t) = \sum_{i=1}^n c^i \mathbf{u}_i(t) \forall t \in I \quad (5.5)$$

The previous theorem tells us that if we know n linearly independent solutions of A , $\{\mathbf{u}_i(t)\}$, we know all its solutions, since these will be linear combinations of $\{\mathbf{u}_i(t)\}$. The dependence on the initial data is also linear, that is, if $\mathbf{x}(t), \mathbf{y}(t) \in \text{Sol}(A, I)$ and $\mathbf{x}(t_o) = \mathbf{x}_o$ and $\mathbf{y}(t_o) = \mathbf{y}_o$ then $\mathbf{x}(t) + \alpha \mathbf{y}(t)$, $\alpha \in \mathbb{R}$, is the solution with initial data $\mathbf{x}_o + \alpha \mathbf{y}_o$ and therefore the map $g_{t_o}^t$ – which in this case we will call $X_{t_o}^t$ – that takes initial data given at $t = t_o$ to solutions with those data, at time t is a linear operator from V to V . Indeed, if $\{\theta^i\}$ is the co-basis of the basis $\{\mathbf{u}_i\}$, i.e. $\theta_o^j(\mathbf{u}_i^\circ) = \delta_o^j$. Then $X_{t_o}^t = \sum_{i=1}^n \mathbf{u}_i(t) \theta_o^i$. Due to the linear independence of $\{\mathbf{u}_i(t)\}$ as elements of V , it can be seen that the map $X_{t_o}^t : V \rightarrow V$ is injective and therefore invertible, for each $t \in I$. Its inverse, which we will denote $(X_{t_o}^t)^{-1}$, takes the solution at t to its initial data at $t = t_o$, that is, $(X_{t_o}^t)^{-1} = X_{t_o}^{t_o}$.

Exercise: Prove that $(X_{t_o}^t)^{-1} = X_{t_o}^{t_o}$.

Exercise: Prove that $X_{t_o}^t$ does not depend on the basis used.

The map $X_{t_o}^t$ is actually the one-parameter family of diffeomorphisms from V

to V generated by the vector field Ax . In this case, these are linear. Indeed, $X_{t_0}^t x_0$ is the curve that passes through the point $x_0 \in V$ at $t = t_0$ and whose tangent vector is $AX_{t_0}^t x_0$.

We have seen then that if we know a set of n linearly independent solutions $\{u_i(t)\}$ we can construct any solution using the operator $X_{t_0}^t$ applied to the initial data of our choice.

How do we know in practice that a set of n solutions $\{u_i(t)\}$ is linearly independent? As we have seen, it is sufficient to see that these are linearly independent, as elements of V , at any time t . That is, the scalar

$$w(t) = \varepsilon(u_1(t), u_2(t), \dots, u_n(t)) \quad (5.6)$$

is different from zero. This function is called the **Wronskian** of the system and satisfies a particularly simple equation whose solution, called the **Liouville formula**, is

$$w(t) = w(t_0) e^{\int_{t_0}^t t r(A(\tilde{t})) d\tilde{t}} \quad (5.7)$$

Proof:

$$\begin{aligned} \dot{w}(t) &= \varepsilon(\dot{u}_1, u_2, \dots, u_n) + \varepsilon(u_1, \dot{u}_2, \dots, u_n) + \dots + \varepsilon(u_1, u_2, \dots, \dot{u}_n) \\ &= \varepsilon(Au_1, u_2, \dots, u_n) + \varepsilon(u_1, Au_2, \dots, u_n) + \dots + \varepsilon(u_1, u_2, \dots, Au_n) \\ &= t r(A) w(t), \end{aligned} \quad (5.8)$$

whose solution is 5.7.

The Liouville formula is an independent demonstration of the result that $\{u_i(t)\}$ form a basis of V for each $t \in I$. Note that if $t r(A) \equiv 0$ the Wronskian is constant and can be useful to determine a solution in terms of others already known.

5.2 Inhomogeneous Linear System – Variation of Constants

Here we will deal with the system,

$$\frac{dx}{dt} = A(t)x + b(t), \quad (5.9)$$

where $b(t) : I \rightarrow V$ is integrable. We will see that if we know the operator $X_{t_0}^t$ corresponding to the homogeneous system

$$\frac{dx}{dt} = A(t)x, \quad (5.10)$$

Then we know all the solutions of system 5.9. The method we will use is called **variation of constants** and it is also useful for obtaining approximate solutions to

non-linear systems. The method consists of assuming that the solution of 5.9 will be of the form,

$$\varphi(t) = X_{t_0}^t c(t) \quad (5.11)$$

for some map $c(t) : I \rightarrow V$, differentiable. Note that if $c(t) = c \in V$, that is, a constant map, then $\varphi(t)$ satisfies 5.10. Substituting $\varphi(t)$ in 5.9 we get,

$$\begin{aligned} \frac{d}{dt}\varphi(t) &= \left(\frac{d}{dt}X_{t_0}^t\right)c(t) + X_{t_0}^t \frac{d}{dt}c(t) \\ &= A(t)X_{t_0}^t c(t) + X_{t_0}^t \frac{d}{dt}c(t) = A(t)X_{t_0}^t c(t) + b(t). \end{aligned} \quad (5.12)$$

From which we see that for $\varphi(t)$ to be a solution, $c(t)$ must satisfy the equation $X_{t_0}^t \frac{d}{dt}c(t) = b(t)$.

But since $X_{t_0}^t$ is invertible and its inverse is $X_t^{t_0}$ we get

$$\frac{d}{dt}c(t) = X_t^{t_0} b(t). \quad (5.13)$$

Integrating we get

$$c(t) = \int_{t_0}^t X_{\tilde{t}}^{t_0} b(\tilde{t}) d\tilde{t} + c(t_0) \quad (5.14)$$

or

$$\varphi(t) = X_{t_0}^t \left[\int_{t_0}^t X_{\tilde{t}}^{t_0} b(\tilde{t}) d\tilde{t} \right] + X_{t_0}^t \varphi(t_0), \quad (5.15)$$

where $\varphi(t_0)$ is any initial condition.

Due to the existence theorem of solutions of the homogeneous system we know that $X_{t_0}^t$ exists and is differentiable $\forall t \in I$ and therefore $\varphi(t)$ also exists and is differentiable $\forall t \in I$, since $b(\tilde{t})$ is integrable. Since the initial data for $\varphi(t)$ can be given arbitrarily, and the solutions of 5.9 are unique, we conclude that 5.15 is the general solution of system 5.9.

5.3 Homogeneous Linear Systems: Constant Coefficients

The equation we will deal with here is

$$\frac{dx}{dt} = Ax, \quad (5.16)$$

where $x : I \subset \mathbb{R} \rightarrow V$ is a curve in V and A is a linear operator of V , $A : V \rightarrow V$.

We already saw in an exercise that $x(t) = e^{At} x_0$, $x_0 \in V$ is a solution of 5.16 with initial condition $x(0) = x_0$. Note that since e^{At} is defined for all $t \in \mathbb{R}$, the solutions we have found are also defined in all \mathbb{R} .

Let $\tilde{x}(t)$ be any solution of 5.16 defined in an interval $\tilde{I} \subset \mathbb{R}$ and let $t_1 \in \tilde{I}$. Then the curve $x_o(t) = e^{A(t_1-t)} \tilde{x}(t_1)$ is also a solution of 5.16 and $x(t_1) = \tilde{x}(t_1)$, but by

the uniqueness of the solutions of 5.16, $x(t)$ and $\tilde{x}(t)$ must coincide in all \tilde{I} and $x(t)$ is the maximum extension of $\tilde{x}(t)$. Thus we see that through the exponential e^{At} we obtain all the solutions of 5.16. In fact, we have shown that the one-parameter family of linear diffeomorphisms $g^t = e^{At}$ is globally defined [$\forall t \in \mathbb{R}, \forall x_0 \in V$] and is the generator of the vector field $v(x) = Ax$. [With the definition given in 5.1 in this case $X_{t_0}^t = g^{t-t_0} = e^{A(t-t_0)}$.]

What is the general form of this family? Or in other words, what is the functional form of the exponential map? For this it is convenient to take a fixed basis $\{u_i\}$ (independent of t) in V and express the equation in terms of the n-tuple of components $\{x^i\}$ of x . In this way we obtain,

$$\frac{dx^i}{dt} = \sum_{j=1}^n A^i_j x^j, \quad (5.17)$$

that is, an equation for the n-tuple $\{x^i\}$. The advantage of this change is that we can take a convenient basis, in particular the orthogonal basis of Lemma 2.2 (of Schur) in which the matrix A^i_j is upper triangular. The price paid for the simplification is that, since the Schur basis is generally complex, both the n-tuple $\{x^i\}$ and the components of A, A^i_j , are complex and therefore we have moved to a problem in C^n , or abstractly in the complexification of V .

For that basis we have that the nth component satisfies a particularly simple equation!

$$\frac{dx^n}{dt} = A^n_n x^n \quad (5.18)$$

and therefore, $x^n(t) = e^{\lambda_n t} x^n_0$ (remember that the diagonal of an upper triangular matrix is composed of its eigenvalues), where x^n_0 is the nth component of the initial condition at $t = 0$. The equation for the (n-1)th component is,

$$\frac{dx^{n-1}}{dt} = A^{n-1}_{n-1} x^{n-1} + A^{n-1}_n x^n \quad (5.19)$$

$$= \lambda_{n-1} x^{n-1} + A^{n-1}_n e^{\lambda_n t} x^n_0. \quad (5.20)$$

Using the formula (5.15) for the one-dimensional case, (and A constant) we get,

$$x^{n-1}(t) = e^{\lambda_{n-1} t} x^{n-1}_0 + e^{\lambda_{n-1} t} \int_0^t e^{-\lambda_{n-1} s} (A^{n-1}_n e^{\lambda_n s} x^n_0) ds \quad (5.21)$$

$$= e^{\lambda_{n-1} t} x^{n-1}_0 + e^{\lambda_{n-1} t} \int_0^t e^{(\lambda_n - \lambda_{n-1})s} ds A^{n-1}_n x^n_0, \quad (5.22)$$

and we see that if $\lambda_n - \lambda_{n-1} \neq 0$ then the solution is a sum of exponentials, $x^{n-1}(t) = e^{\lambda_{n-1} t} x^{n-1}_0 + \frac{A^{n-1}_n x^n_0}{\lambda_n - \lambda_{n-1}} (e^{\lambda_n t} - e^{\lambda_{n-1} t})$. If the eigenvalues coincide ($\lambda_n - \lambda_{n-1} = 0$) then

a linear term in t appears, $x^{n-1}(t) = e^{\lambda_{n-1}t}(x_0^{n-1} + tA^{n-1}_n x_0^n)$. Knowing $x^{n-1}(t)$ we can now calculate, using again (5.15), $x^{n-2}(t)$ and so on for all the components of $x(t)$ in this basis. The final result will be that the components are all sums of exponentials by polynomials. Indeed note that the integral of a polynomial of degree m by an exponential is also a polynomial of the same degree by the same exponential plus an integration constant (unless the exponential is zero, in which case the result is a polynomial of degree $m + 1$).

Exercise: Prove this last statement by induction on the degree of the polynomial and the use of integration by parts, that is, use that $e^{at}P_m(t) = \frac{d}{dt}(\frac{e^{at}}{a}P_m(t)) - \frac{e^{at}}{a}\frac{d}{dt}P_m(t)$ and that $\frac{d}{dt}P_m(t)$ is a polynomial of order $m - 1$.

Rearranging terms we see that the general solution will be of the form:

$$x(t) = \sum_{i=1}^m e^{\lambda_i t} v_i(t) \quad (5.23)$$

where the sum is over the different eigenvalues and where the vectors $v_i(t)$ are polynomials in t .

An important property of the algebraic structure of this solution is manifested with the following lemma.

Lemma 5.1 *The vector spaces*

$$E_\lambda := \{f(t) : \mathbb{R} \mapsto V^C \mid f(t) = e^{\lambda t} P_m(t), \text{ with } P_m(t) \text{ a polynomial}\}$$

are linearly independent.

Proof: We must then prove that if we have a finite sum of elements of the $E_{\lambda_i}, \{u_i\}$ and this sum is zero, then each of these vectors must be identically zero. That is

$$\sum_{i=1}^m u_i = 0, \quad u_i \in E_{\lambda_i}, \quad \lambda_i \neq \lambda_j \Rightarrow u_i = 0. \quad (5.24)$$

We will prove it by induction on the number of elements in the sum. The case $m = 1$ is trivial. We will then assume that the statement is true for $m - 1$ and we will prove it for m . We will also assume by contradiction that there is a sum of m non-zero elements that is zero, that is:

$$\sum_{i=1}^m u_i = 0, \quad u_i \in E_{\lambda_i}, \quad \lambda_i \neq \lambda_j, \quad u_i \neq 0. \quad (5.25)$$

dividing by $e^{\lambda_1 t}$ we get

$$o = S(t) := e^{-\lambda_1 t} \sum_{i=1}^m u_i = P_{m_1}(t) + \sum_{i=2}^m e^{(\lambda_i - \lambda_1)t} P_{m_i}(t). \quad (5.26)$$

Taking $m_1 + 1$ derivatives of $S(t)$, where m_1 is the order of the polynomial of the first term we get:

$$o = \frac{d^{m_1+1}}{dt^{m_1+1}} S(t) = \sum_{i=2}^m e^{(\lambda_i - \lambda_1)t} \tilde{P}_{m_i}(t). \quad (5.27)$$

where it is easy to see that the polynomials $\tilde{P}_{m_i}(t)$ are of the same degree as the originals. We are thus under the inductive hypothesis and therefore since these polynomials are non-zero we have a contradiction ♠

Let us now see that in fact each of the terms in this sum is a solution of equation 5.16. Given any $\lambda \in C$ consider the vector space of functions from $\mathbb{R} \rightarrow V^C$ of the form $e^{\lambda t} w(t)$, where $w(t)$ is a polynomial¹ in t . This space is invariant under the action of $\frac{d}{dt}$, since taking its derivative we get a similar expression, that is, the product of the same exponential by another polynomial. On the other hand it is invariant under the action of A , since as A does not depend on t its action on any $e^{\lambda t} w(t)$ gives us another similar expression. Therefore, the action of $\frac{d}{dt} - A$ will also keep us in the same space. Thus we see that if we have a sum of terms with this structure and with different values of λ , as is the case in the previous equation 5.23, then if the application of $\frac{d}{dt} - A$ gives us zero this means that the application of $\frac{d}{dt} - A$ to each term of the sum must give zero. That is, each term is a solution of equation 5.16! Thus we conclude that each term in the sum 5.16 is of the form $e^{A t} v_o$ for some $v_o \in V^C$. Let us now consider the subset W_λ of V^C , such that if $v_o \in W_\lambda$ then $e^{A t} v_o = e^{\lambda t} v(t)$. It is clear that the only non-trivial subspaces will be those with $\lambda = \lambda_i$, where $\{\lambda_i\}$, $i = 1..d$ are the eigenvalues of A . Therefore we must consider only these subsets. Since this relationship is linear, W_λ is a subspace of V^C . Since $e^{A t} A v_o = A e^{A t} v_o = e^{\lambda t} A v(t) = e^{\lambda t} v'(t)$ and $v'(t)$ is also polynomial in t , we see that W_λ is an invariant subspace of A . While using a basis of W_λ in which the restriction to that space of A is upper triangular, we see that λ is the only eigenvalue that such a restriction can have. Finally, since every solution of 5.16 has the form 5.23, the existence theorem of solutions assures us that any initial data, that is, any element of V^C , can be expressed as a linear combination of elements in W_{λ_i} . Indeed, let v_o be any element of V^C , from the existence theorem of solutions, we then have a unique solution of 5.16 $x(t)$ with $x(o) = v_o$. But then

$$\begin{aligned} x(t) &= \sum_{i=1}^m e^{\lambda_i t} v_i(t) \\ &= \sum_{i=1}^m e^{A t} v_{oi}, \end{aligned} \quad (5.28)$$

¹That is, a linear combination of vectors in V^C with polynomial coefficients in t .

for some set of vectors $\{v_{oi}\} \in W_{\lambda_i}$. Evaluating this expression at $t = 0$, we obtain $v_o = \sum_{i=1}^m v_{oi}$ and we see that the W_{λ_i} generate V^C . On the other hand, the uniqueness of the solutions implies that no element of a given W_{λ_i} can be written as a linear combination of elements of the other W_{λ_j} [otherwise the same initial data would give rise to two different solutions (since their functional dependence would be different)]. Indeed, let $0 = v_{o1} + \dots + v_{os}$ be any linear combination of elements of W_{λ_i} , $i = 1..s$, we see that each of them must be zero. By the uniqueness of the solutions to the ordinary equations, we then have $0 = \sum_{i=1}^m e^{At} v_{oi} = \sum_{i=1}^m e^{\lambda_i t} v_i(t)$ as seen earlier, each element in the last sum must be zero and therefore, evaluating these at $t = 0$, we obtain $v_{oi} = 0 \forall i = 1..s$. Summarizing the above, we have,

Theorem 5.3 *Given an operator $A: V^C \rightarrow V^C$, there exists a set of invariant subspaces of A , $\{W_{\lambda_i}\}$, where λ_i are the eigenvalues of A such that:*

a) $V^C = W_{\lambda_1} \oplus W_{\lambda_2} \oplus \dots \oplus W_{\lambda_d}$

b) *The only eigenvalue of the restriction of A in $\{W_{\lambda_i}\}$ is λ_i .*

We conclude the study of the solutions of equation 5.16, that is, of the function e^{At} with a detailed description of their form obtained using the matrix representation of A in the basis where it acquires the Jordan canonical form. That is, for every invariant subspace W_{λ_i} the restriction of A in this subspace has the form,

$$A = \lambda_i I + \Delta_i, \quad (5.29)$$

where the numbers λ_i are the eigenvalues corresponding to A —generally complex, and therefore we are now considering A as an operator from C^n to C^n —and Δ_i is a matrix whose only non-zero components are ones and can only be in the immediate upper diagonal. The matrix Δ_i has the important property of being nilpotent, that is, there exists an integer m_i less than or equal to the multiplicity of λ_i , such that $\Delta_i^{m_i} = 0$.

Since $\lambda_i I$ and Δ_i commute, we have that

$$e^{t\lambda_i I + t\Delta_i} = e^{t\lambda_i I} e^{t\Delta_i}, \quad (5.30)$$

but $e^{t\lambda_i I} = e^{t\lambda_i} I$ and $e^{t\Delta_i} = \sum_{j=0}^{m_i-1} \frac{(t\Delta_i)^j}{j!}$, that is, a finite sum.

Since the exponential of A is a sum of powers of A , it follows that the invariant spaces of this are the same as those of A and therefore, $e^{tA}|_{B_i} = e^{tA_i}$. From this, we conclude that the matrix e^{tA} is of the form

$$e^{tA} = \begin{pmatrix} k_0 & & & \\ & k_1 & & \\ & & \ddots & \\ & & & k_p \end{pmatrix} \quad (5.31)$$

with each k_i a square sub-matrix, the $k_o = \text{diag}(e^{t\lambda_1}, \dots, e^{t\lambda_n})$ and if $i \neq o$

$$k_i = e^{t\lambda} \begin{bmatrix} 1 & t & t^2/2 & \cdots & \cdots & t^{n-1}/(n-1)! \\ & 1 & t & \cdots & \cdots & \vdots \\ & & 1 & \cdots & \cdots & \vdots \\ & & & \ddots & \cdots & t^2/2 \\ 0 & & & & \ddots & t \\ & & & & & 1 \end{bmatrix}, \quad (5.32)$$

where the number of rows or columns is the same as those of the J_i corresponding to A in the Jordan composition, this is less than or equal to the multiplicity of λ_i as a root of the characteristic polynomial.

The basis in which A has the Jordan canonical form is generally complex, it is actually a basis in C^n . If we wish to use a real basis, which is possible since we start with A as an operator from R^n to R^n , and we make the corresponding transformation, the matrix e^{At} will undergo the corresponding similarity transformation. Since this transformation is independent of time, although the components of e^{At} will not have the previous form, they will be sums of exponential terms by polynomials in t . Therefore, each component of the general solution of 5.16 will be of the form,

$$x^i(t) = \sum_{p=1}^q \left\{ (e^{t\lambda_p} + e^{t\bar{\lambda}_p}) P_p^i(t) + i(e^{t\lambda_p} - e^{t\bar{\lambda}_p}) Q_p^i(t) \right\} \quad (5.33)$$

where q is the number of distinct eigenvalues –counting complex pairs as one– and P_p^i , Q_p^i are polynomials in t whose degree is less than or equal to the multiplicity with which the eigenvalue λ_p appears as a root of the characteristic polynomial. This information is useful in two ways. Practically, because although the method of constructing the solution using a basis in which A has the Jordan canonical form is straightforward, for large dimension systems it becomes cumbersome. In some cases, it is convenient to calculate the eigenvalues and their multiplicity and then assume a solution of the form 5.33 and calculate the polynomials P_p^i , Q_p^i .

This method is also useful in the sense that it allows us to know the global behavior of the solutions. For example, we see that if the real part of the eigenvalues is not positive and those whose real part is zero do not appear repeated, then all solutions are bounded [There exists $C > 0$ such that $\|x(t)\| < C$.] If in addition, all have negative real part, then all solutions tend asymptotically to the trivial solution [$\lim_{t \rightarrow \infty} x(t) = 0$].

We will now analyze in detail the case where all eigenvalues are distinct. Note that although this is the generic case –in the sense that if a system has coincident eigenvalues, then there are arbitrarily small modifications to it that make the eigenvalues distinct– systems with coincident eigenvalues do appear in physics. This case also contemplates the situation of a general system where the initial data belongs to one of the one-dimensional subspaces B_i .

Since the operator A is real, its eigenvalues will be real or complex conjugates [if $\det(A - \lambda I) = 0$ then $\det(\overline{A - \lambda I}) = \det(A - \bar{\lambda} I) = 0$]. If λ_i is real, then its eigenvector can be chosen to be real. Indeed, if $(A - \lambda I)\mathbf{u}_i = 0$, then also $(A - \lambda I)\bar{\mathbf{u}}_i = 0$, but the roots are simple and therefore each λ_i has only one eigenvector—modulo a complex scalar—, that is $\bar{\mathbf{u}}_i = \alpha \mathbf{u}_i$ $\alpha \in \mathbb{C}$. Choosing $\mathbf{v}_i = \mathbf{u}_i + \bar{\mathbf{u}}_i = (1 + \alpha)\mathbf{u}_i$ we obtain a real eigenvector.

In this case, the component x_o^i of \mathbf{x}_o in the eigenbasis in the direction \mathbf{v}_i evolves as

$$x^i(t) = x_o^i e^{\lambda_i t}, \quad (5.34)$$

that is, it grows or decays exponentially with time according to the sign of λ_i .

If the eigenvalue λ_i is complex, then we can choose its eigenvector \mathbf{u}_i such that it is the complex conjugate of the chosen one corresponding to $\bar{\lambda}_i$. This pair of eigenvectors generates a 2-dimensional complex subspace. If \mathbf{x}_o belongs to this subspace and is real, then it will have the form $\mathbf{x}_o = a(\mathbf{u}_i + \bar{\mathbf{u}}_i) - ib(\mathbf{u}_i - \bar{\mathbf{u}}_i)$ with a and b real, that is $\mathbf{x}_1 = \mathbf{u}_i + \bar{\mathbf{u}}_i$ and $\mathbf{x}_2 = i(\mathbf{u}_i - \bar{\mathbf{u}}_i)$ form a real basis.

How do these vectors change if we apply the operator e^{At} to them? That is, what are the solutions of the equation $\dot{\mathbf{x}} = A\mathbf{x}$ with initial conditions \mathbf{x}_1 and \mathbf{x}_2 ? Calling these $\mathbf{x}_1(t)$ and $\mathbf{x}_2(t)$ respectively and using that $e^{At}\mathbf{u}_i = e^{\lambda_i t}\mathbf{u}_i$ we obtain,

$$\begin{aligned} \mathbf{x}_1(t) &= e^{\alpha_i t}(\mathbf{x}_1 \cos w_i t - \mathbf{x}_2 \sin w_i t) \\ \mathbf{x}_2(t) &= e^{\alpha_i t}(\mathbf{x}_2 \cos w_i t + \mathbf{x}_1 \sin w_i t), \end{aligned} \quad (5.35)$$

with $\lambda_i = \alpha_i + i w_i$.

We see that the action of the operator e^{At} in this case is to dilate the vectors by a factor of $e^{\alpha_i t}$ and to rotate them by an angle $w_i t$.

5.4 Problems

Problem 5.1 Given a set of functions $\{f_i(t)\}$, $i = 1, \dots, n$, vectors are defined as $\mathbf{u}_i(t) := (f_i(t), f_i^{(1)}(t), \dots, f_i^{(n-1)}(t))$ and the Wronskian of the system as $W(\{f_i\})(t) := \varepsilon(\mathbf{u}_1(t), \mathbf{u}_2(t), \dots, \mathbf{u}_n(t))$. If the Wronskian of a set of functions does not vanish, then the functions are linearly independent, that is, no non-trivial linear combination of them (with constant coefficients) vanishes. The converse is not true. Calculate the Wronskian of the following sets:

- a) $\{4, t\}$
- b) $\{t, 3t, t^2\}$
- c) $\{e^t, t e^t, t^2 e^t\}$
- d) $\{\sin(t), \cos(t), \cos(2t)\}$
- e) $\{1, \sin(t), \sin(2t)\}$

Problem 5.2 Decide if the following set of functions is linearly dependent or not. Then calculate the Wronskian.

$$f_1(t) = \begin{cases} 0, & 0 \leq x \leq 1/2 \\ (x - 1/2)^2, & 1/2 \leq x \leq 1 \end{cases} \quad (5.36)$$

$$f_2(t) = \begin{cases} (x - 1/2)^2, & 0 \leq x \leq 1/2 \\ 0, & 1/2 \leq x \leq 1 \end{cases} \quad (5.37)$$

Problem 5.3 Use the theory of ordinary differential equations to prove the following identities:

a)

$$e^s A e^t A = e^{(s+t)A}, \quad (5.38)$$

b)

$$e^A e^B = e^{A+B}, \text{ if and only if } [A, B] := AB - BA = 0. \quad (5.39)$$

Problem 5.4 Graph the vector fields corresponding to the following systems and some typical sets of solutions.

a)

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & \frac{1}{2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (5.40)$$

b)

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (5.41)$$

c)

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (5.42)$$

d)

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (5.43)$$

e)

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (5.44)$$

f) (compare the solutions with those of point b)

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (5.45)$$

g)

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} i & 0 \\ 0 & i \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (5.46)$$

h)

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1+i & 0 \\ 0 & 1-i \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (5.47)$$

i)

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1+i & 0 \\ 1 & 1-i \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad (5.48)$$

Problem 5.5 Reduce these equations to first-order systems and find the general solution of the equations:

a) $\frac{d^3x}{dt^3} - 2\frac{d^2x}{dt^2} - 3\frac{dx}{dt} = 0$

b) $\frac{d^3x}{dt^3} + 2\frac{d^2x}{dt^2} + \frac{dx}{dt} = 0$

c) $\frac{d^3x}{dt^3} + 4\frac{d^2x}{dt^2} + 13\frac{dx}{dt} = 0$

d) $\frac{d^4x}{dt^4} + 4\frac{d^3x}{dt^3} + 8\frac{d^2x}{dt^2} + 8\frac{dx}{dt} + 4 = 0$

Problem 5.6 (Newton's Law) Consider the equation $\frac{d^2x}{dt^2} + f(x) = 0$.

a) Prove that $\frac{1}{2}\dot{x}^2 + \int_{x_0}^x f(s)ds$ is a first integral.

b) Find the first integral of $\frac{d^2x}{dt^2} - x + x^2/2 = 0$.

c) Graph the corresponding vector field and some of its solutions. Find its stationary solutions (or equilibrium points) and study their neighborhoods by linearizing the equation at these points.

Problem 5.7 Study the following system

$$\begin{aligned}\dot{x}_1 &= x_1 - x_1x_2 - x_2^3 + x_3(x_1^2 + x_2^2 - 1 - x_1 + x_1x_2 + x_2^3) \\ \dot{x}_2 &= x_1 - x_3(x_1 - x_2 + x_1x_2) \\ \dot{x}_3 &= (x_3 - 1)(x_3 + 2x_3x_2^2 + x_3^3)\end{aligned}\tag{5.49}$$

a) Find the equilibrium points.

b) Show that the planes $x_3 = 0$ and $x_3 = 1$ are invariant sets, (that is, solutions that start in them never leave them).

c) Consider the invariant set $x_3 = 1$ and see if it has periodic solutions.

CHAPTER 6

STABILITY

Stationary or equilibrium solutions are very important in physics because many systems (especially those with dissipation) behave in such a way that the solution approaches these solutions in their evolution.

Another particularity that makes them important is the fact that they are simply given by the points where the vector field vanishes, and therefore at most one must solve an algebraic equation to find them, in contrast to the general case where we must solve a differential equation. In some cases, it is even possible to infer that the vector field must vanish, such as if the manifold is a two-dimensional sphere, since any continuous vector field defined on it is such that there are at least two points where it vanishes, and therefore in this case there will always be stationary solutions.

Exercise: Convince yourself that this is so.

From the above, it follows that it is important to study the behavior of solutions that originate from initial data close to a stationary solution. For this, we define the following concepts of stability.

Definition: Let v be a vector field in M and let $p \in M$ such that $v(p) = 0$. We will say that the stationary solution $\gamma(t) \equiv p$ is **stable** if given a neighborhood U_p of p there exists another neighborhood V_p of p such that any solution $\sigma(t)$ with $\sigma(0) \in V_p$ satisfies $\sigma(t) \in U_p \quad \forall t \geq 0$. [See figure 6.1.]

Definition: We will say that the previous solution is **asymptotically stable** if it is stable and also if $\sigma(0) \in V_p$ then $\lim_{t \rightarrow +\infty} \sigma(t) = p$.

If a stationary solution is not stable, then it has little physical interest since the slightest perturbation will *move it far* from it.

Examples:

a) The bacterial growth equation $\dot{x} = ax - bx^2$ has stationary solutions $x(t) \equiv 0$ and

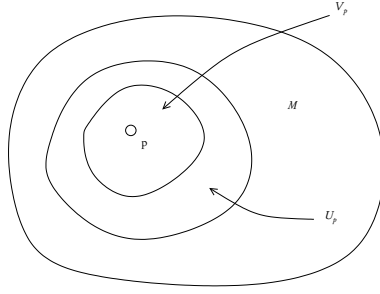


Figure 6.1: Stability.

$x(t) \equiv \frac{a}{b}$. Since the general solution is,

$$x(t) = \frac{x(0)e^{at}}{1 + \frac{bx(0)}{a}(e^{at} - 1)}, x(0) \geq 0 \quad (6.1)$$

it is clear that $x(t) = 0$ is not a stable solution (we will call these unstable) and that $x(t) \equiv \frac{a}{b}$ is asymptotically stable. If we contaminate a culture container with a single bacterium, this is enough for it to reproduce¹ until it reaches (asymptotically) the concentration $\frac{a}{b}$. If we now remove or add some bacteria to change their concentration, the bacteria will reproduce or annihilate until they reach again *the stable concentration* $\frac{a}{b}$.

b) The equation $\dot{x} = Ax$ has among its stationary solutions the one given by $x(t) \equiv 0$. What are the others? These will be stable when the eigenvalues of A , λ_i , satisfy $\Re(\lambda_i) \leq 0$, $\lambda_i \neq \lambda_j$ or $\Re(\lambda_i) < 0$ if $\lambda_i = \lambda_j$, where \Re indicates the real part. This is clear since the general solution is $x(t) = e^{At}x(0)$ and the mentioned condition implies that $\|e^{At}\|_{\mathcal{L}} < C$, $C > 0 \forall t \geq 0$. In this case, we can take as $U_{p=0} = \{x \in \mathbb{R}^n \mid |x| < \epsilon\}$ and as $V_{p=0} = \{x \in \mathbb{R}^n \mid |x| < \frac{\epsilon}{C}\}$. If it is also fulfilled that $\Re(\lambda_i) < 0 \forall i = 1, \dots, n$ then $x(t) \equiv 0$ is asymptotically stable.

The following theorem provides a very practical tool to know when a stationary solution is stable or not.

Theorem 6.1 (Stability) *Let $\gamma(t) \equiv p$ be a stationary solution of $v \in TM$, that is $v(p) = 0$, and let $A : T_p M \rightarrow T_p M$ be defined by,*

$$Ax \equiv \frac{d}{ds} v(\sigma_x(s))|_{s=0}, \quad (6.2)$$

¹Note that in the real process we do not have a continuous variable and therefore in it the perturbations are at least one bacterium, which makes it possible to have sterile containers and therefore the mathematical instability does not manifest in reality.

where $\sigma_x(s)$ is a curve in M satisfying: $\sigma_x(o) = p$ and $\frac{d}{ds}\sigma_x|_{s=o} = x \in T_p M$, that is, it is any smooth curve that passes through p when $s = o$ and at that point has as tangent to $x \in T_p M$. If $\Re(\lambda_i) < o$, where λ_i are the eigenvalues of A , then $\gamma(t) \equiv p$ is asymptotically stable. If any λ_i has a positive real part then $\gamma(t) = p$ is unstable.

Exercises:

a) Show that A is really a linear operator.

b) Show that $Ax = [v, \tilde{x}]_p$, where \tilde{x} is any vector field such that $\tilde{x}|_p = x$. Hint: do not forget that $v|_p = o$.

Examples:

a) Consider the bacterial growth equation at $x = o$ and $x = \frac{a}{b}$. For the first case, we take $\sigma_{\delta x}(s) = \delta x s$ and obtain,

$$A\delta x = \frac{d}{ds}(a\delta x s - b(\delta x)^2 s^2)|_{s=o} = a\delta x, \quad (6.3)$$

which shows that $x = o$ is an unstable solution since $a > o$. For the second case, we take $\sigma_{\delta x} = \frac{a}{b} + \delta x s$ then

$$A\delta x = \frac{d}{ds}\left(\frac{a^2}{b} + a\delta x s - b\left(\frac{a}{b} + \delta x s\right)^2\right)|_{s=o} = a\delta x - 2a\delta x = -a\delta x, \quad (6.4)$$

which shows that it is asymptotically stable.

b) Physical pendulum with friction: $\ddot{\theta} = -\sin\theta - k\dot{\theta}$, $k > o$. The vector field in this case is given by,

$$\begin{aligned} \dot{\theta} &= z \\ \dot{z} &= -\sin\theta - kz, \end{aligned} \quad (6.5)$$

that is, the vector with components $(z, -\sin\theta - kz)$ in the phase space which in this case is a cylinder, $z \in [-\infty, +\infty]$, $\theta \in [o, 2\pi]$.

This vector vanishes only at $p_1 = (\theta = o, z = o)$ and $p_2 = (\theta = \pi, z = o)$. For the first stationary solution, using $\sigma(s)$ such that,

$$\phi \circ \sigma(s) = \begin{pmatrix} \delta\theta s \\ \delta z s \end{pmatrix}, \quad (6.6)$$

we obtain,

$$A \begin{pmatrix} \delta\theta \\ \delta z \end{pmatrix} = \begin{pmatrix} o & 1 \\ -1 & -k \end{pmatrix} \begin{pmatrix} \delta\theta \\ \delta z \end{pmatrix}. \quad (6.7)$$

In this case, the eigenvalue equation is $\lambda^2 + k\lambda + 1 = o$ from which we obtain, $\lambda_{\pm} = \frac{-k \pm \sqrt{k^2 - 4}}{2}$, which implies $\Re(\lambda_{\pm}) < o$ and thus stability.

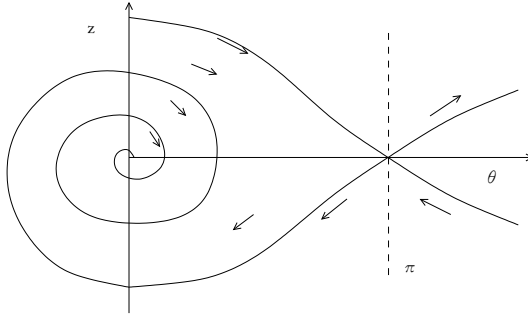


Figure 6.2: Physical pendulum with friction.

For the second stationary solution, we take $\sigma(s)$ such that,

$$\phi \circ \sigma(s) = \begin{pmatrix} \delta \theta s + \pi \\ \delta z \end{pmatrix} \quad (6.8)$$

and obtain

$$A \begin{pmatrix} \delta \theta \\ \delta z \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & -k \end{pmatrix} \begin{pmatrix} \delta \theta \\ \delta z \end{pmatrix}. \quad (6.9)$$

with eigenvalues $\lambda_{\pm} = \frac{-k \pm \sqrt{k^2 + 4}}{2}$, which implies $\Re(\lambda_{\pm}) > 0$ and thus instability.

From the examples, it is seen that this theorem has broad application. Since stability is a local notion, to facilitate the demonstration we will take $M = \mathbb{R}^n$ and a Cartesian coordinate system with origin at the stable solution to be considered. We will use several previous results that we prove below.

Theorem 6.2 (Lyapunov) *Let $A : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be a linear operator whose eigenvalues have a negative real part. Then there exists a tensor of type $(2,0)$, $\rho(\cdot, \cdot)$, symmetric and positive definite [$\rho(w, v) = \rho(v, w)$ and $\rho(w, w) \geq 0$ ($= 0$ if $w = 0$)] such that,*

$$Ax(\rho(x, x)) < 0 \quad \forall \quad x \neq 0,$$

that is, the derivative of $\rho(x, x)$ in the direction Ax is negative.

The geometric interpretation of this condition is that $\rho(x, x)$ defines a norm whose level surfaces are such that on each of these the vector Ax at $p = x$ points inward. See figure.

Proof: If all the eigenvalues of A are distinct, then there exists a Jordan basis $\{u_i\}$, $i = 1, \dots, n$ and the corresponding co-basis $\{\theta^i\}$, $i = 1, \dots, n$ (with $\theta^i(u_j) = \delta_j^i$) such that $A = \sum_{i=1}^n \lambda_i u_i \theta^i$. In this case, let $\rho = \sum_{i=1}^n \theta^i \bar{\theta}^i$. If $z = \sum_{i=1}^n z^i u_i$ and

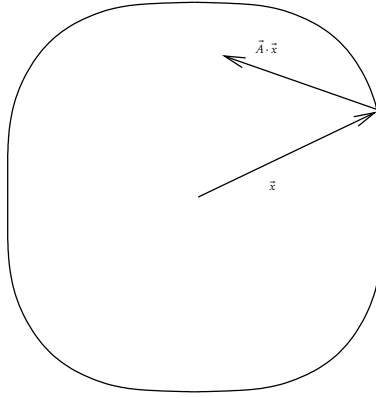


Figure 6.3: The norm $\rho(x, x)$.

$y = \sum_{i=1}^n y^i u_i \in C^n$, then $\rho(z, y) = \sum_{i=1}^{n'} z^i y^i + \sum_{n'+1}^{(n-n')/2} (z^i \bar{y}^i + \bar{z}^i y^i)$ where we have separated the sum into that of the real eigenvectors and that of the complex conjugates. It is easy to see that the $\rho(,)$ obtained by restricting z and y to \mathbb{R}^n is the usual Cartesian norm in the corresponding real basis. Now let's calculate $Ax(\rho(x, x))$.

$$\begin{aligned}
 Ax(\rho(x, x)) &= \lim_{\varepsilon \rightarrow 0} \frac{\rho(x + \varepsilon Ax, x + \varepsilon Ax) - \rho(x, x)}{\varepsilon} \\
 &= \rho(Ax, x) + \rho(x, Ax) \\
 &= 2 \sum_{i=1}^n (\Re \lambda_i) x^i x^i < 0.
 \end{aligned} \tag{6.10}$$

Thus, we have proved the theorem for the diagonalizable case. The case where A is not is more complicated, and for this, we will use the following lemmas.

Lemma 6.1 *Given $\epsilon > 0$ there exists a basis $\{u_i\}$ such that in that basis*

$$A = \text{diag}(\lambda_1, \dots, \lambda_n) + \epsilon \Delta,$$

with Δ a matrix with non-zero components (and equal to one) at most on the lower diagonal, that is,

$$A = \begin{pmatrix} \lambda_1 & 0 & 0 & \dots & \dots & \dots & 0 \\ \epsilon & \lambda_2 & 0 & \dots & \dots & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & \dots & \dots & \dots & \dots & \dots & 0 \\ 0 & \dots & \dots & \dots & \dots & \epsilon & \lambda_r \end{pmatrix}. \tag{6.11}$$

Proof: It is a simple rescaling of the Jordan basis of A . For example, in C^3 if $\{\tilde{u}_i\}$ is

such that in it,

$$A = \begin{pmatrix} \lambda & 0 & 0 \\ 1 & \lambda & 0 \\ 0 & 1 & \lambda \end{pmatrix}. \quad (6.12)$$

defining $u_1 = \frac{\tilde{u}_1}{\epsilon^2}$, $u_2 = \frac{\tilde{u}_2}{\epsilon}$, and $u_3 = \tilde{u}_3$, we see that in this new basis,

$$A = \begin{pmatrix} \lambda & 0 & 0 \\ \epsilon & \lambda & 0 \\ 0 & \epsilon & \lambda \end{pmatrix}. \quad (6.13)$$

$$[Au_1 = A \frac{\tilde{u}_1}{\epsilon^2} = \frac{\lambda \tilde{u}_1}{\epsilon^2} = \lambda u_1, Au_2 = A \frac{\tilde{u}_2}{\epsilon} = \frac{\tilde{u}_1 + \lambda \tilde{u}_2}{\epsilon} = \epsilon u_1 + \lambda u_2, \text{ etc.}]$$

Lemma 6.2 *The set of positive symmetric tensors is open in the set of symmetric forms, that is, if ρ_0 is symmetric and positive and ρ_1 is symmetric, then there exists $\epsilon > 0$ such that $\rho_0 + \epsilon \rho_1$ is also positive.*

Proof: Let B_1 be the unit sphere with respect to some norm in \mathbb{R}^n and consider $\rho_0(x, x) : B_1 \rightarrow \mathbb{R}^+$. Since B_1 is compact, ρ_0 reaches its minimum there, α_{\min} . Similarly, ρ_1 reaches its maximum there, which we will call β_{\max} . Taking $\epsilon < \frac{\alpha_{\min}}{\beta_{\max}}$ satisfies the requirement.

Example: In \mathbb{R}^2 let $\rho_0((x_1, y_1), (x_2, y_2)) = x_1 x_2 + y_1 y_2$ and $\rho_1((x_1, y_1), (x_2, y_2)) = x_1 y_2 + x_2 y_1$, then $\rho_0 + \epsilon \rho_1$ is positive if $|\epsilon| < 2$.

To complete the proof of Lyapunov's theorem we will take

$$\rho = \sum_i \theta^i \otimes \bar{\theta}^i,$$

where $\{\theta^i\}$ is the co-basis found in lemma 6.1 with $\epsilon > 0$ to be determined. Then,

$$\begin{aligned} -Ax(\rho(x, x)) &= -\rho(Ax, x) - \rho(x, Ax) \\ &= -2 \sum_{i=1}^n \Re(\lambda_i) x^i x^i - 2\epsilon \sum_{i=1}^{n-1} f_i x^i x^{i+1}, \end{aligned} \quad (6.14)$$

with $f_i = 1$ or 0 depending on whether there is ϵ or not in that place of the immediate upper diagonal. The first term is a positive definite form ρ_0 evaluated in x in both entries. The second is a symmetric form $\epsilon \rho_1$ evaluated in x in both entries. The second lemma tells us that taking ϵ small $\rho_0 + \epsilon \rho_1$ is also positive definite. This concludes the proof of the theorem.

Note now that not only the ρ we have found is symmetric and positive definite and therefore a norm in \mathbb{R}^n , but also $-Ax(\rho(x, x))$ is positive definite and defines a norm [since $-Ax(\rho(x, x)) = -\rho(Ax, x) - \rho(x, Ax)$]. But we have already seen that in \mathbb{R}^n all norms are equivalent and therefore there will exist $\gamma > 0$ such that

$$-\frac{1}{2\gamma} \rho(x, x) \leq Ax(\rho(x, x)) \leq -2\gamma \rho(x, x). \quad (6.15)$$

This is the result we will use for the proof of the stability theorem given below.

Proof of the Stability Theorem: We have seen that if $\Re(\lambda_i) < 0$ then there exists a constant $\gamma > 0$ and a symmetric positive definite form $\rho(\cdot, \cdot)$ such that

$$Ax(\rho(x, x)) \leq -2\gamma \rho(x, x). \quad (6.16)$$

Applying the vector field that defines the equation, $v(x)$ to $\rho(x, x)$ we get,

$$v(x)(\rho(x, x)) = Ax(\rho(x, x)) + O(\rho(x, x)^{3/2}), \quad (6.17)$$

where we have assumed v is differentiable and therefore $v(x) = Ax + O(|x|^2)$.

Example: $v(x) = (ax + bx^2)\frac{\partial}{\partial x}$ and $\rho(x, x) = \alpha x^2$ then,

$$\begin{aligned} v(x)(\rho(x, x)) &= ax \frac{\partial}{\partial x} \rho(x, x) + 2b\alpha x^3 \\ &= ax \frac{\partial}{\partial x} \rho(x, x) + \frac{2b}{\alpha^{1/2}} \rho(x, x)^{3/2} \end{aligned} \quad (6.18)$$

If x is sufficiently small, $[|x| < C \text{ for some } C > 0]$ it holds that $O(\rho(x, x)^{3/2}) < \gamma(\rho(x, x))$ and therefore we have that

$$v(x)(\rho(x, x)) \leq -2\gamma \rho(x, x) + \gamma \rho(x, x) \leq -\gamma \rho(x, x). \quad (6.19)$$

Let $\varphi(t)$ be a solution with initial data sufficiently close to $x = 0$, that is $\dot{\varphi}(t) = v(\varphi(t))$, $|\varphi(0)| < C$. Defining ²

$$r(t) = \ln \rho(\varphi(t), \varphi(t)) \quad (6.20)$$

and differentiating with respect to t we get,

$$\begin{aligned} \dot{r}(t) &= \frac{2 \rho(\varphi(t), \varphi(t))}{\rho(\varphi(t), \varphi(t))} = \frac{v(\varphi(t))(\rho(\varphi(t), \varphi(t)))}{\rho(\varphi(t), \varphi(t))} \\ &\leq -\gamma. \end{aligned} \quad (6.21)$$

²By uniqueness of the solution $\varphi(t) \neq 0$ and therefore $\rho(\varphi(t), \varphi(t)) > 0$ and the logarithm is well defined.

Therefore $r(t) \leq r(o) - \gamma t$, and integrating we conclude that

$$\rho(\varphi(t), \varphi(t)) \leq \rho(\varphi(o), \varphi(o))e^{-\gamma t}, \quad (6.22)$$

which implies that $\varphi(t) \xrightarrow[t \rightarrow \infty]{} o$ and concludes the proof of the theorem. ³

It is instructive to observe that the proof is based on the construction of a norm, $\rho(x, x)$, especially adapted to the problem in the sense that it assures us that on its level surfaces the vector $v(x)$ (if x is sufficiently small) points inward.

6.1 Problems

Problem 6.1 (The Volterra-Lotka Equation) *Consider the system:*

$$\begin{aligned} \dot{x}_1 &= (a - bx_2)x_1 \\ \dot{x}_2 &= -(c - fx_1)x_2, \end{aligned} \quad (6.23)$$

$a, b, c, f \geq 0$.

a) Perform a coordinate and time transformation to bring it to the form,

$$\begin{aligned} \dot{x}_1 &= (1 - x_2)x_1 \\ \dot{x}_2 &= -(e - x_1)x_2 \end{aligned} \quad (6.24)$$

b) Plot the vector field and see that there are no non-trivial first integrals in the quadrants where at least one of the coordinates is negative.

c) Find the equilibrium solutions and determine which are stable and which are not.

d) Examine the positive quadrant using the transformation: $x_1 = e^{q_1}$, $x_2 = e^{q_2}$ and see that the quantity $f(q_1, q_2) := e q_1 + q_2 - (e^{q_1} + e^{q_2})$ is an integral of motion. Use this information to infer that in this quadrant the trajectories remain in bounded regions.

e) Examine the linearized equations around the equilibrium solution in the positive quadrant and determine the frequency of oscillations of the variations near equilibrium that the system would have.

Note, this equation describes the population of two competing species. What we have just seen is that the species do not grow indefinitely nor disappear. This last point tells us that the approximation is not very good... Note that x_2 represents a predator that depends exclusively on the prey x_1 for its subsistence, since if the prey becomes extinct, the predator follows.

Problem 6.2 *If in the previous system the species have other alternative means of subsistence then the resulting equations are:*

$$\begin{aligned} \dot{x}_1 &= (1 - x_1 - ax_2)x_1 \\ \dot{x}_2 &= (1 - x_2 + bx_1)x_2 \end{aligned} \quad (6.25)$$

³In reality, we must also prove that if $|\varphi(o)| < C$ then $\varphi(t)$ exists for all t . But the last inequality proved tells us that $\varphi(t)$ cannot leave the compact region $\{x | \rho(x, x) \leq \rho(\varphi(o), \varphi(o))\}$ and therefore the extension theorem assures us that $\varphi(t)$ exists for all $t \in [0, +\infty)$.

where another parameterization has been used.

a) Find the equilibrium solutions for the cases i) $0 < a < 1$, and ii) $1 < a$.

b) It is observed that $b \approx 3a$ (a indicates how aggressive the predator is). What is the value of a if its instinct leads it to maximize the population of its species while maintaining a stable equilibrium?

c) Look at the system near its equilibrium solution and determine the frequency of the oscillations in the number of species that you would expect in that environment.

Problem 6.3 (Limit Cycle) Consider the system:

$$\begin{aligned}\dot{x}_1 &= x_2 + x_1(1 - (x_1^2 + x_2^2)) \\ \dot{x}_2 &= -x_1 + x_2(1 - (x_1^2 + x_2^2)).\end{aligned}\quad (6.26)$$

a) Transform this system into a pair of decoupled equations,

$$\begin{aligned}\dot{r} &= f(r) \\ \dot{\theta} &= -1.\end{aligned}\quad (6.27)$$

b) Study the equilibrium points of the first equation and their stability. What is the solution of the original system corresponding to this equilibrium point?

c) Plot solutions near these points, in the (r, θ) plane and in the (x_1, x_2) plane.

d) Use the method described at the end of chapter ?? to corroborate the stability found in point b).

Problem 6.4 (Verhulst) Find the equilibrium solutions (and analyze their stability) of the system:

$$\begin{aligned}\dot{x} &= y \\ \dot{y} &= x - 2x^3\end{aligned}\quad (6.28)$$

Plot some solutions.

Problem 6.5 (Verhulst) Find the equilibrium solutions (and analyze their stability) of the system:

$$\begin{aligned}\dot{x} &= x^2 - y^3 \\ \dot{y} &= 2x(x^2 - y)\end{aligned}\quad (6.29)$$

Plot some solutions.

Problem 6.6 (Verhulst) Find the equilibrium solutions (and analyze their stability) of the system:

$$\begin{aligned}\dot{x} &= -x \\ \dot{y} &= 1 - (x^2 + y^2)\end{aligned}\quad (6.30)$$

Plot some solutions.

PROOF OF THE FUNDAMENTAL THEOREM

We will only prove points *i*) and *ii*), for which we will use the method of successive approximations by Picard, which is important in applications. The proof of point *iii*) is of a technical nature and does not provide any additional significant insight. Point *iv*) follows from points *i*) and *ii*), and its proof is identical to that of Corollary 1.2 of the analogous Theorem for the case of an ODE.

To prove these points, we will need to develop some ideas and results from the mathematical theory of infinite-dimensional vector spaces. We will restrict ourselves to the minimum required by the theorem since these topics will be developed more extensively in the second part of this course.

Definition: We will say that a vector space, V , is of **infinite dimension** if it has an infinite number of linearly independent vectors.

Examples:

a) Let V be the set of all sequences $\{x_i\}$, $i = 1, \dots, \infty$ of real numbers. This is a vector space if we define the **sum** and the **product** of sequences by the following formula,

$$\{x_i\} + \alpha\{y_i\} = \{x_i + \alpha y_i\}. \quad (7.1)$$

These vectors can also be written as $\{x_i\} = (x_1, x_2, x_3, \dots)$, which shows that it is the extension of \mathbb{R}^n with $n \rightarrow \infty$. Clearly, the vectors $u_1 = (1, 0, 0, \dots)$; $u_2 = (0, 1, 0, \dots)$; $u_3 = (0, 0, 1, \dots)$; are linearly independent and infinite in number.

b) Let V be the set of continuous functions on the interval $[0, 1]$. This is a vector space since if f and g are continuous on $[0, 1]$, then $h = f + \alpha g$, $\alpha \in \mathbb{R}$, is also continuous on $[0, 1]$.

The set of vectors $u_n = x^n$, $n \in \mathbb{N}$, is linearly independent and infinite ($\sum_{n=0}^M c^n u_n = \sum_{n=0}^M c^n x^n = 0 \implies c^n = 0 \quad \forall n \leq M$) since a polynomial of degree M has at most M roots.

Infinite-dimensional spaces can also be assigned norms, but in this case, these norms are not equivalent, and therefore one must be careful not to confuse the resulting structures. To avoid confusion, we will assign different names to spaces with different norms.

Examples: a) The normed vector space l^2 is the space of infinite sequences with the norm $\|\{x_i\}\|_2 = \sqrt{\sum_{i=1}^{\infty} x_i^2} < \infty$.

b) The normed vector space l^∞ is the space of infinite sequences with the norm $\|\{x_i\}\|_\infty = \sup_i \{|x_i|\}$. That is, the space of all bounded sequences.

c) The normed vector space $C[a, b]$ is the space of continuous functions with the norm $\|f\|_c = \sup_{x \in [a, b]} \{|f(x)|\}$. That is, the space of bounded continuous functions on $[a, b]$.

Exercises:

1) Show that the norms defined in the previous example are indeed norms.
 2) Show that there are sequences in l^∞ that do not belong to l^2 . Hint: Find one of them.

3) Show that $\|\{x_i\}\|_n := \sqrt[n]{\sum_{i=1}^{\infty} |x_i|^n} \xrightarrow{n \rightarrow \infty} \sup_i \{|x_i|\}$.

Unlike the finite-dimensional case, an infinite-dimensional normed space is not necessarily complete. To illustrate this, consider the normed vector space l_0^∞ , which is a subspace of l^∞ consisting of all bounded sequences with only a finite number of non-zero terms. Each sequence $\{x_i\}_n = (1, 1/2, 1/3, \dots, 1/n, 0, 0, \dots)$ is in l_0^∞ , the sequence of sequences $\{\{x_i\}_n\}$ is Cauchy

$$\left(\|\{x_i\}_m - \{x_i\}_n\|_\infty = \frac{1}{n+1} \xrightarrow{n \rightarrow \infty} 0 \right)$$

and converges to the sequence $\{x_i\}_\infty = (1, 1/2, 1/3, \dots) \in l^\infty$ which does not belong to l_0^∞ . Therefore, l_0^∞ is not complete.

Definition: We will say that a normed vector space $(V, \|\cdot\|)$ is a **Banach space** if it is complete.

Examples: \mathbb{R}^n with any of its norms, l^2 and l^∞ are Banach spaces.

An important result, crucial in the proof of the Fundamental Theorem, is the following theorem.

Theorem 7.1 *The vector space of bounded continuous functions, $C[a, b]$, is complete.*

Proof: Let $\{f_n(x)\}$ be a Cauchy sequence in $C[a, b]$, that is, each $f_n(x)$ is a continuous and bounded function on $[a, b]$ and it holds that given $\varepsilon > 0$ there exists $N > 0$ such that $\forall m, n > N \quad \|f_n - f_m\|_c = \sup_{x \in [a, b]} |f_n(x) - f_m(x)| < \varepsilon$. But then for each $x \in [a, b]$ the sequence of real numbers $\{f_n(x)\}$ is Cauchy. But the real numbers are complete and therefore for each $x \in [a, b]$ $\{f_n(x)\}$ converges to a number that we will call $f(x)$. But then given $\varepsilon > 0$ for all N such that $m, n \geq N$ implies $\|f_n - f_m\|_c < \varepsilon$, we have that

$$\begin{aligned}
\sup_{x \in [a, b]} |f(x) - f_N(x)| &= \sup_{x \in [a, b]} \lim_{n \rightarrow \infty} |f_n(x) - f_N(x)| \\
&\leq \sup_{n \geq N} \sup_{x \in [a, b]} |f_n(x) - f_N(x)| \\
&= \sup_{n \geq N} \|f_n - f_N\|_c < \varepsilon.
\end{aligned} \tag{7.2}$$

Therefore, if we could prove that $f \in C[a, b]$, then we would have that $\|f - f_n\|_c \rightarrow 0$, $n \rightarrow \infty$ and therefore that $\{f_n\} \rightarrow f$ in $C[a, b]$ and the theorem would be proven.

Let $x \in [a, b]$ be any point, we will prove that f is continuous at x and thus on all $[a, b]$. Let $\varepsilon > 0$, we want to find a δ such that $|x - y| < \delta$ implies $|f(x) - f(y)| < \varepsilon$. Take N such that $\|f - f_N\|_c < \varepsilon/3$ and δ such that $|x - y| < \delta$ implies $|f_N(x) - f_N(y)| < \varepsilon/3$ [This is possible since the $f_N(x)$ are continuous on $[a, b]$]. Then $|x - y| < \delta$ implies

$$\begin{aligned}
|f(x) - f(y)| &\leq |f(x) - f_N(x)| + |f_N(x) - f_N(y)| + |f_N(y) - f(y)| \\
&< \frac{1}{3}\varepsilon + \frac{1}{3}\varepsilon + \frac{1}{3}\varepsilon = \varepsilon.
\end{aligned} \tag{7.3}$$

and therefore the continuity of $f(x)$. Since $[a, b]$ is a compact set (closed and bounded), f is bounded on $[a, b]$ and therefore belongs to $C[a, b]$ ♠

Definition: Let $T : V \rightarrow V$ be a map from a Banach space V to itself. We will say that T is a **contraction** if there exists $\lambda < 1$ such that:

$$\|T(x) - T(y)\|_V \leq \lambda \|x - y\|_V. \tag{7.4}$$

Examples:

- a) The linear map in l^2 ; $A\{x_i\} = \left\{\frac{x_i}{i+1}\right\}$.
- b) The map from \mathbb{R}^2 to \mathbb{R}^2 that sends the point (x, y) to the point $(x_0 + x/2, y/2)$.
- c) Any Lipschitz function from \mathbb{R} to \mathbb{R} with a modulus of continuity (k) less than one.

The important property of these maps is the following theorem

Theorem 7.2 *Let $T : V \rightarrow V$ be a contraction. Then there exists a unique $v \in V$ such that $T(v) = v$, and the sequence $T^n(u)$ converges to v for any u .*

Proof: Let $\|T(u) - u\| = d$, then $\|T^{n+1}(u) - T^n(u)\| \leq \lambda^n d$ and if $m > n$

$$\begin{aligned}
&\|T^m(u) - T^n(u)\| \\
&= \|T^m(u) - T^{m-1}(u) + T^{m-1}(u) - T^{m-2}(u) + \dots - T^n(u)\| \\
&\leq \|T^m(u) - T^{m-1}(u)\| + \|T^{m-1}(u) - T^{m-2}(u)\| + \dots \\
&\leq d \sum_{n=0}^m \lambda^n.
\end{aligned}$$

Since $\sum_{n=0}^{\infty} \lambda^n$ converges, $\{T^n(u)\}$ is a Cauchy sequence. But V is complete and therefore there exists $v \in V$ such that $\lim_{n \rightarrow \infty} T^n u = v$.

Since every contraction is continuous, we have that

$$T(v) = T(\lim_{n \rightarrow \infty} T^n(u)) = \lim_{n \rightarrow \infty} T^{n+1}(u) = v.$$

It only remains to prove that v is unique. Suppose by contradiction that there exists $w \in V$ different from v and such that $Tw = w$. But then $\|w - v\|_V = \|T(w) - T(v)\|_V \leq \lambda \|w - v\|_V$ which is a contradiction since $\lambda \neq 1$ ♠

Exercise: Let $T : B_{R, x_0} \rightarrow B_{R, x_0}$, $B_{R, x_0} \in V$ be a closed ball of radius R around x_0 , a contraction ¹. Prove for this case the same statements as in the previous theorem.

Proof of points i) and ii) of the fundamental theorem. Since we only want to see local existence and uniqueness, that is, only in a neighborhood of a point p of M , it is sufficient to consider the system in \mathbb{R}^n . This will allow us to use the Euclidean norm present there. To see this, take a chart (U, φ) with $p \in U$ and for simplicity $\varphi(p) = 0 \in \mathbb{R}^n$. Using this map, we can translate the vector field v in M to a vector field \tilde{v} defined in a neighborhood of zero in \mathbb{R}^n . There we will treat the problem of finding its integral curves $g(t, x)$ that pass through the point $x \in \varphi(U)$ at time $t = 0$. Then, through the map φ^{-1} , we will obtain in M the one-parameter families of diffeomorphisms $g^t(q)$, $q \in U$, which will be tangent at every point to the vector field v .

With this in mind, it only remains to see that for $R > 0$ and $\varepsilon > 0$ sufficiently small there exist integral curves, $g(t, x) : [0, \varepsilon] \times B_R = \{x \in \mathbb{R}^n \mid \|x\|_V < R\} \rightarrow \mathbb{R}^n$, of the vector \tilde{v} , that is, maps satisfying

$$\frac{dg(t, x)}{dt} = \tilde{v}(g(t, x)) \quad , \quad g(0, x) = x, \quad (7.5)$$

with $g(t, x)$ continuous with respect to the second argument, that is, with respect to the initial condition. By the assumption that v is Lipschitz ², we have that there exists $k > 0$ such that $\forall x, y \in B_R$

$$\|\tilde{v}(x) - \tilde{v}(y)\|_V < k \|x - y\|_V. \quad (7.6)$$

Now consider the Banach space $C([0, \varepsilon] \times B_R)$, which consists of all continuous maps (in t and x) from $[0, \varepsilon] \times B_R$ to \mathbb{R}^n with the norm

$$\|h\|_C = \sup_{\substack{x \in B_R \\ t \in [0, \varepsilon]}} \|h(t, x)\|_V, \quad (7.7)$$

Let the map of the ball of radius R in $C([0, \varepsilon] \times B_R)$ into itself be given by,

$$T(h) = \int_0^t \tilde{v}(x + h(\tau, x)) d\tau. \quad (7.8)$$

¹ Also adjust the definition of a contraction for this case.

² To prove the local existence and uniqueness of solutions, it is only necessary to assume that v is Lipschitz.

For this map to be well-defined, we will assume R is sufficiently small so that \tilde{v} is defined and satisfies the Lipschitz condition in B_{2R} and ε is less than R/C , where $C = \max_{x \in B_{2R}} |\tilde{v}(x)|$, so that if $\|b\|_C < R$ then $\|x + b(\tau, x)\|_V < 2R \quad \forall \tau \in [0, \varepsilon]$ and therefore $\|T(b)\|_C < R$ in all $C([0, \varepsilon] \times B_R)$. [See figure 7.1.]

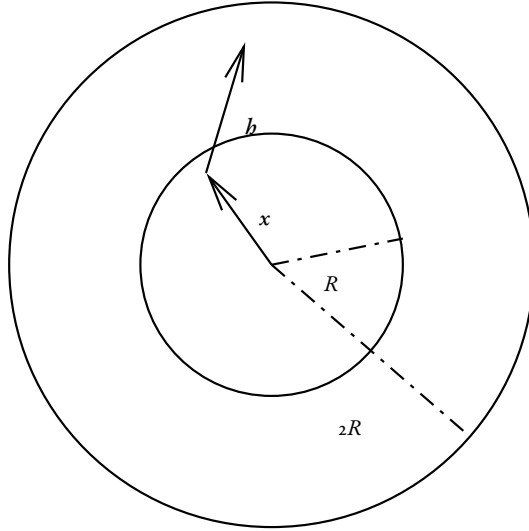


Figure 7.1: Neighborhoods used in the proof of the Fundamental Theorem.

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Infinite-dimensional spaces can also be assigned norms, but in this case, these norms are not equivalent, and therefore one must be careful not to confuse the resulting structures. To avoid confusion, we will assign different names to spaces with different norms.

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b) The normed vector space l^∞ is the space of infinite sequences with the norm $\|\{x_i\}\|_\infty = \sup_i \{|x_i|\}$. That is, the space of all bounded sequences.

c) The normed vector space $C[a, b]$ is the space of continuous functions with the norm $\|f\|_c = \sup_{x \in [a, b]} \{|f(x)|\}$. That is, the space of bounded continuous functions on $[a, b]$.

Exercises:

- 1) Show that the norms defined in the previous example are indeed norms.
- 2) Show that there are sequences in l^∞ that do not belong to l^2 . Hint: Find one of them.
- 3) Show that $\|\{x_i\}\|_n := \sqrt[n]{\sum_{i=1}^{\infty} |x_i|^n} \xrightarrow{n \rightarrow \infty} \sup_i \{|x_i|\}$.

Unlike the finite-dimensional case, an infinite-dimensional normed space is not necessarily complete. To illustrate this, consider the normed vector space l_0^∞ , which is a subspace of l^∞ consisting of all bounded sequences with only a finite number of non-zero terms. Each sequence $\{x_i\}_n = (1, 1/2, 1/3, \dots, 1/n, 0, 0, \dots)$ is in l_0^∞ , the sequence of sequences $\{\{x_i\}_n\}$ is Cauchy

$$\left(\|\{x_i\}_m - \{x_i\}_n\|_\infty = \frac{1}{n+1} \xrightarrow{n \rightarrow \infty} 0 \right)$$

and converges to the sequence $\{x_i\}_\infty = (1, 1/2, 1/3, \dots) \in l^\infty$ which does not belong to l_0^∞ . Therefore, l_0^∞ is not complete.

Definition: We will say that a normed vector space $(V, \|\cdot\|)$ is a **Banach space** if it is complete.

Examples: \mathbb{R}^n with any of its norms, l^2 and l^∞ are Banach spaces.

An important result, crucial in the proof of the Fundamental Theorem, is the following theorem.

Theorem 7.3 *The vector space of bounded continuous functions, $C[a, b]$, is complete.*

Proof: Let $\{f_n(x)\}$ be a Cauchy sequence in $C[a, b]$, that is, each $f_n(x)$ is a continuous and bounded function on $[a, b]$ and it holds that given $\varepsilon > 0$ there exists $N > 0$ such that $\forall m, n > N \quad \|f_n - f_m\|_c = \sup_{x \in [a, b]} |f_n(x) - f_m(x)| < \varepsilon$. But then for each $x \in [a, b]$ the sequence of real numbers $\{f_n(x)\}$ is Cauchy. But the real numbers are complete and therefore for each $x \in [a, b]$ $\{f_n(x)\}$ converges to a number that we will call $f(x)$. But then given $\varepsilon > 0$ for all N such that $m, n \geq N$ implies $\|f_n - f_m\|_c < \varepsilon$, we have that

$$\begin{aligned} \sup_{x \in [a, b]} |f(x) - f_N(x)| &= \sup_{x \in [a, b]} \lim_{n \rightarrow \infty} |f_n(x) - f_N(x)| \\ &\leq \sup_{n \geq N} \sup_{x \in [a, b]} |f_n(x) - f_N(x)| \\ &= \sup_{n \geq N} \|f_n - f_N\|_c < \varepsilon. \end{aligned} \quad (7.10)$$

Therefore, if we could prove that $f \in C[a, b]$, then we would have that $\|f - f_n\|_c \rightarrow 0, n \rightarrow \infty$ and therefore that $\{f_n\} \rightarrow f$ in $C[a, b]$ and the theorem would be proven.

Let $x \in [a, b]$ be any point, we will prove that f is continuous at x and thus on all $[a, b]$. Let $\varepsilon > 0$, we want to find a δ such that $|x - y| < \delta$ implies $|f(x) - f(y)| < \varepsilon$. Take N such that $\|f - f_N\|_c < \varepsilon/3$ and δ such that $|x - y| < \delta$ implies $|f_N(x) - f_N(y)| < \varepsilon/3$ [This is possible since the $f_N(x)$ are continuous on $[a, b]$]. Then $|x - y| < \delta$ implies

$$\begin{aligned} |f(x) - f(y)| &\leq |f(x) - f_N(x)| + |f_N(x) - f_N(y)| + |f_N(y) - f(y)| \\ &< \frac{1}{3}\varepsilon + \frac{1}{3}\varepsilon + \frac{1}{3}\varepsilon = \varepsilon. \end{aligned} \quad (7.11)$$

and therefore the continuity of $f(x)$. Since $[a, b]$ is a compact set (closed and bounded), f is bounded on $[a, b]$ and therefore belongs to $C[a, b]$ ♠

Definition: Let $T : V \rightarrow V$ be a map from a Banach space V to itself. We will say that T is a **contraction** if there exists $\lambda < 1$ such that:

$$\|T(x) - T(y)\|_V \leq \lambda \|x - y\|_V. \quad (7.12)$$

Examples:

- The linear map in l^2 ; $A\{x_i\} = \left\{ \frac{x_i}{i+1} \right\}$.
- The map from \mathbb{R}^2 to \mathbb{R}^2 that sends the point (x, y) to the point $(x_0 + x/2, y/2)$.
- Any Lipschitz function from \mathbb{R} to \mathbb{R} with a modulus of continuity (k) less than one.

The important property of these maps is the following theorem

Theorem 7.4 Let $T : V \rightarrow V$ be a contraction. Then there exists a unique $v \in V$ such that $T(v) = v$, and the sequence $T^n(u)$ converges to v for any u .

Proof: Let $\|T(u) - u\| = d$, then $\|T^{n+1}(u) - T^n(u)\| \leq \lambda^n d$ and if $m > n$

$$\begin{aligned} & \|T^m(u) - T^n(u)\| \\ &= \|T^m(u) - T^{m-1}(u) + T^{m-1}(u) - T^{m-2}(u) + \dots + T^n(u)\| \\ &\leq \|T^m(u) - T^{m-1}(u)\| + \|T^{m-1}(u) - T^{m-2}(u)\| + \dots \\ &\leq d \sum_{n=0}^{m-n} \lambda^n. \end{aligned}$$

Since $\sum_{n=0}^{\infty} \lambda^n$ converges, $\{T^n(u)\}$ is a Cauchy sequence. But V is complete and therefore there exists $v \in V$ such that $\lim_{n \rightarrow \infty} T^n u = v$.

Since every contraction is continuous, we have that

$$T(v) = T(\lim_{n \rightarrow \infty} T^n(u)) = \lim_{n \rightarrow \infty} T^{n+1}(u) = v.$$

It only remains to prove that v is unique. Suppose by contradiction that there exists $w \in V$ different from v and such that $Tw = w$. But then $\|w - v\|_V = \|T(w) - T(v)\|_V \leq \lambda \|w - v\|_V$ which is a contradiction since $\lambda \neq 1$ ♠

Exercise: Let $T : B_{R, x_0} \rightarrow B_{R, x_0}$, $B_{R, x_0} \in V$ be a closed ball of radius R around x_0 , a contraction ³. Prove for this case the same statements as in the previous theorem.

Proof of points i) and ii) of the fundamental theorem. Since we only want to see local existence and uniqueness, that is, only in a neighborhood of a point p of M , it is sufficient to consider the system in \mathbb{R}^n . This will allow us to use the Euclidean norm present there. To see this, take a chart (U, φ) with $p \in U$ and for simplicity $\varphi(p) = o \in \mathbb{R}^n$. Using this map, we can translate the vector field v in M to a vector field \tilde{v} defined in a neighborhood of zero in \mathbb{R}^n . There we will treat the problem of finding its integral curves $g(t, x)$ that pass through the point $x \in \varphi(U)$ at time $t = o$. Then, through the map φ^{-1} , we will obtain in M the one-parameter families of diffeomorphisms $g^t(q)$, $q \in U$, which will be tangent at every point to the vector field v .

With this in mind, it only remains to see that for $R > 0$ and $\varepsilon > 0$ sufficiently small there exist integral curves, $g(t, x) : [o, \varepsilon] \times B_R = \{x \in \mathbb{R}^n \mid \|x\|_V < R\} \rightarrow \mathbb{R}^n$, of the vector \tilde{v} , that is, maps satisfying

$$\frac{dg(t, x)}{dt} = \tilde{v}(g(t, x)) \quad , \quad g(o, x) = x, \quad (7.13)$$

with $g(t, x)$ continuous with respect to the second argument, that is, with respect to the initial condition. By the assumption that v is Lipschitz ⁴, we have that there

³Also adjust the definition of a contraction for this case.

⁴To prove the local existence and uniqueness of solutions, it is only necessary to assume that v is Lipschitz.

exists $k > 0$ such that $\forall x, y \in B_R$

$$\|\tilde{v}(x) - \tilde{v}(y)\|_V < k \|x - y\|_V. \quad (7.14)$$

Now consider the Banach space $C([0, \varepsilon] \times B_R)$, which consists of all continuous maps (in t and x) from $[0, \varepsilon] \times B_R$ to \mathbb{R}^n with the norm

$$\|h\|_C = \sup_{\substack{x \in B_R \\ t \in [0, \varepsilon]}} \|h(t, x)\|_V, \quad (7.15)$$

Let the map of the ball of radius R in $C([0, \varepsilon] \times B_R)$ into itself be given by,

$$T(h) = \int_0^t \tilde{v}(x + h(\tau, x)) d\tau. \quad (7.16)$$

For this map to be well-defined, we will assume R is sufficiently small so that \tilde{v} is defined and satisfies the Lipschitz condition in B_{2R} and ε is less than R/C , where $C = \max_{x \in B_{2R}} |\tilde{v}(x)|$, so that if $\|h\|_C < R$ then $\|x + h(\tau, x)\|_V < 2R \quad \forall \tau \in [0, \varepsilon]$ and therefore $\|T(h)\|_C < R$ in all $C([0, \varepsilon] \times B_R)$. [See figure 7.1.]

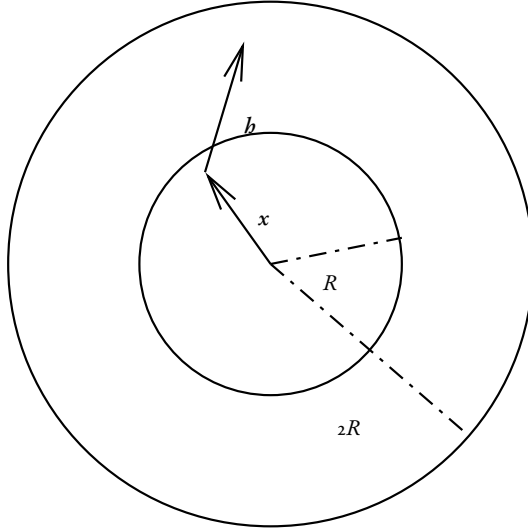


Figure 7.2: Environments used in the proof of the Fundamental Theorem.

Lemma 7.1 *If ε is sufficiently small then T is a contraction.*

Proof:

$$\begin{aligned}
 \|T(h_1) - T(h_2)\|_V &= \int_0^t \|\tilde{v}(x + h_1(\tau, x)) - \tilde{v}(x + h_2(\tau, x))\|_V d\tau \\
 &\leq \int_0^t k \|h_1(\tau, x) - h_2(\tau, x)\|_V d\tau \\
 &\leq k\varepsilon \|h_1 - h_2\|_C
 \end{aligned}$$

and therefore

$$\|T(h_1) - T(h_2)\|_C \leq k\varepsilon \|h_1 - h_2\|_C \quad \forall h_1, h_2 \in C([0, \varepsilon] \times B_R).$$

Taking $\varepsilon < 1/k$ we complete the proof of the Lemma.

This Lemma and Theorem 7.4 ensure that there exists a unique map $h(t, x)$ –the fixed point of T – satisfying

$$h(t, x) = T(h(t, x)) = \int_0^t \tilde{v}(x + h(\tau, x)) d\tau. \quad (7.17)$$

Let $g(t, x) \equiv x + h(t, x)$, this function is continuous in both arguments –since $h(t, x) \in C([0, \varepsilon] \times B_R)$ – and by construction continuously differentiable in t –since it satisfies (7.17)–. Differentiating (7.17) with respect to t we see that $g(t, x)$ satisfies the equation (7.13) and its initial condition, which completes the proof of points $i)$ and $ii)$ of the Fundamental Theorem. ♠

7.1 Problems

Problem 7.1 See that l^2 the unit ball is not compact. Hint: find an infinite sequence in the unit ball that has no accumulation point.

Problem 7.2 Prove that the condition

$$\|A(x) - A(y)\| < \|x - y\| \quad (7.18)$$

is not sufficient to guarantee the existence of a fixed point. Hint: construct a counterexample. There are some very simple ones using functions from the line to itself.

Problem 7.3 Let $f : [a, b] \rightarrow [a, b]$ be a Lipschitz function with a constant less than one throughout the interval $[a, b]$. Prove using the fixed point theorem for contractions that the equation $f(x) = x$ always has a solution. Plot in a diagram $(y = f(x), x)$ the sequence given by $x_i = f(x_{i-1})$ for the case of f with positive slope and less than one at every point. What happens when the slope becomes greater than one in some interval?

Problem 7.4 Let $g : [a, b] \rightarrow \mathbb{R}$ be a continuously differentiable function such that $g(a) < 0$, $g(b) > 0$ and $0 < c_1 < g'(x)$. Use the fixed point theorem for contractions to prove that there is a unique root $g(x) = 0$ in the interval. Hint: define the function $f(x) = x - \lambda g(x)$ for a conveniently chosen constant λ and find a fixed point: $f(x) = x$. Note that the approximating sequence is that of the Newton method for finding roots.

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