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Injectivity of the inverse optimal control problem for control-affine systems

Frédéric Jean¹ and Sofya Maslovskaya²

Abstract—Given a control system and a set of optimal trajectories, is it possible to recover the cost for which the trajectories are minimizing? This question is called inverse optimal control problem, and the problem is said to be injective when it admits a unique solution. In this paper we present a general approach to address the issue of the cost uniqueness in the class of quadratic costs and in the case of dynamics given by a control-affine system. We then apply this method to characterize the non-uniqueness cases for a special subclass of control-affine systems.

I. INTRODUCTION

Inverse optimal control problem has attracted renewed interest in the last 10 years due to the growing number of its application, in particular, in modeling the human movements and in the field of humanoid robotics. Actually, one of the dominant hypothesis in physiology for describing human movements is that the realized movements of the mechanical system minimize some cost function under dynamics constraints (see [28], [27] for instance). However, even if it is known that a movement is optimal, the criteria being optimized is hidden. Thus, to model the human movements we should first solve an inverse optimal control problem: given the data of the realized movements and the dynamic constraints (i.e. the control system), find the cost function with respect to which the movements are optimal. This approach has already proved to be useful in the study of the human locomotion [8], [22] and the arm movements [4], [5]. In robotics, inverse optimal control is a tool to get the most adapted cost function to produce for instance biomimetic motions for humanoid robots [2], [17], or to predict human actions for autonomous robots [14], [15]. However, while in robotics we can be satisfied with numerical and approximate resolution methods, the modeling of human movement requires a more qualitative approach to understand the phenomena involved. It is therefore necessary to carry out a theoretical study of the problem.

From a mathematical point of view, inverse optimal control problems belong to the class of inverse problems where the first issue is whether the problem is well posed. Formally, given a dynamic and a class of candidate cost functions, for the corresponding direct optimal control problems we can define an operator which maps a cost function to the optimal synthesis, i.e. to the set of all optimal trajectories of the corresponding optimal control problem, for all realizable

initial and final points. The inverse optimal control problem is to inverse this operator. For such an inverse problem to be well posed, it should be surjective, injective and stable. In applications surjectivity is usually assumed to be satisfied, and the first issue is the injectivity, i.e. is there a one to one correspondence between the costs and the optimal syntheses?

In control theory, inverse optimal control was first introduced by Kalman in 1964 [12] for linear-quadratic regulators. Many authors have addressed this issue then (see [13] for a detailed review), and we finally made a complete study of the injectivity of the inverse optimal control in linear-quadratic case in [9]. Note that before the definition of Kalman, the inverse problem was already extensively studied in the domain of the calculus of variation (see [24] for a review), but these studies focus on the surjectivity of the inverse problem, not on the injectivity. After the paper of Kalman, the inverse problem was stated for different classes of optimal control problems: for nonlinear [26], [18], discrete [20], [25], stochastic (Markov decision processes [19], [23]) cases and others. Nevertheless, even in the continuous deterministic case, no general methodology emerges, mainly because the inverse problem with costs of general form is highly ill-posed, as noted in [21].

In view of these results, it is legitimate to reduce the class of candidate cost functions, typically by restricting to the class of integral costs that are quadratic with respect to control. For this class, we developed a new approach based on the concept of orbital diffeomorphism, which allowed us to make a very complete study of the case where the dynamics is a control-affine system without drift [10], [11] (see also [29]). The aim of the present paper is to extend this approach to the case of general control-affine systems.

Our approach (both in [10], [11] and in [9]) is based on one guideline and two technical steps. The guideline is that non injectivity of the inverse problem only appears when the control system admits a particular product structure. The first technical step is to show that the data of the inverse problem, the minimizers, can be transferred at the level of the Pontryagin Hamiltonian equations near appropriate trajectories, called ample geodesics. This idea allows to define the notion of orbital diffeomorphisms, which is our main tool to exhibit the product structure. The second step consists in replacing the control system by a nilpotent approximation when analyzing the equations of the orbital diffeomorphism.

For the study of control-affine systems, we keep the same guideline and we seek to use the same technical steps. Compared to the driftless case, however, we face new difficulties since there are no general results on ample geodesics and

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no good notion of nilpotent approximation. Thus we had to develop the corresponding tools, which is done in the PhD thesis [16]. Since these developments are long and technical, they are presented in this paper without proof or details, for which we refer to [16] (they will be the subject of a future paper). Then we show how to apply these tools in the first treatable case, the one with $n-1$ controls (n is the dimension of the state space).

The paper is organized as follows. Sect. II presents the main definitions and introduce the notion of equivalent costs, which allows us to study the injectivity of the inverse problem. We give in Sect. III the results on ample geodesics, orbital diffeomorphisms, and on the link between the latter concept and cost equivalence, and in Sect. IV the construction of the nilpotent approximation and its consequence on the analysis of ample geodesics and orbital diffeomorphism. Finally we study in Sect. V the case with $n-1$ controls.

II. INVERSE OPTIMAL CONTROL PROBLEMS

A. Inverse problem and injectivity

Fix a smooth n -dimensional manifold M and a control-affine system on M given by

$$\dot{q} = f_0(q) + \sum_{i=1}^m u_i f_i(q), \quad u = (u_1, \dots, u_m) \in \mathbb{R}^m, \quad (1)$$

where f_0, f_1, \dots, f_m are smooth vector fields on M . We denote by q_u a trajectory of this system. The choice of a criterion

$$J(q_u) = \int_0^T L(q_u(t), u(t)) dt,$$

defines the following class of direct optimal control problems: given a pair of points q_0, q_1 and a final time $T > 0$, determine

$$\inf\{J(q_u) : q_u \text{ traj. of (1), } q_u(0) = q_0, q_u(T) = q_1\}. \quad (2)$$

The corresponding inverse optimal control problem arises as follows: given the dynamics (1) and a set Γ of trajectories, find a cost L such that every $\gamma \in \Gamma$, $\gamma : [0, T] \rightarrow M$, is a solution of the optimal control problem (2) associated with $q_0 = \gamma(0)$ and $q_1 = \gamma(T)$.

We consider in this paper the question of the injectivity of the inverse problem: does the set of minimizers of (2) (for all $q_0, q_1 \in M$ and $T > 0$) determine in a unique way the cost L in a chosen class of costs \mathcal{L} ? The uniqueness is meant “up to a multiplicative constant”, since two constantly proportional costs cL and L trivially define the same set of minimizers.

We choose here to consider this problem in the class of quadratic costs (note however that most of our results can easily be extended to the class of Tonelli Lagrangians).

Definition 2.1: The class \mathcal{L} is the set of quadratic costs

$$L(q, u) = u^T R(q) u, \quad (3)$$

where, for any $q \in M$, $R(q)$ is a symmetric positive definite ($m \times m$) real valued matrix, and the mapping $q \mapsto R(q)$ is smooth.

We make moreover the following hypothesis on the control system.

Assumption 1: The vector fields f_0, f_1, \dots, f_m satisfy:

- the weak Hörmander condition at any $q \in M$, i.e.,

$$\text{Lie}_q(\{(ad f_0)^s f_i : s \geq 0, i = 1, \dots, m\}) = T_q M; \quad (4)$$

- the points q such that $\text{rank}\{f_0, f_1, \dots, f_m\}(q) = m+1$, called regular points, form an open dense subset of M .

The first condition ensures that the optimal control problem (2) has a non trivial set of solutions. The second one guarantees that the system can never be reduced locally to a driftless control system.

We will study the injectivity through the following equivalence relation on \mathcal{L} .

Definition 2.2: We say that two costs L and \tilde{L} in \mathcal{L} are equivalent via minimizers if they define the same set of minimizers of (2).

The existence of two non-proportional costs in \mathcal{L} which are equivalent via minimizers implies that the inverse optimal control problem may not have a unique solution, i.e. that the inverse problem is not injective.

B. Example of non-trivially equivalent costs

Let us construct an example of non proportional costs having the same minimizers. Fix positive integers N and $m_i \leq n_i$ for $i = 1, \dots, N$, and set $m = \sum_i m_i$, $n = \sum_i n_i$. We denote the coordinates on $\mathbb{R}^n = \mathbb{R}^{n_1} \times \dots \times \mathbb{R}^{n_N}$ by $x = (\bar{x}_1, \dots, \bar{x}_N)$, where $\bar{x}_j = (x_j^1, \dots, x_j^{n_j})$, and similarly the coordinates on \mathbb{R}^m by $u = (\bar{u}_1, \dots, \bar{u}_N)$. We consider an affine-control system on \mathbb{R}^n of the form:

$$\begin{cases} \dot{\bar{x}}_1 = f_0^1(\bar{x}_1) + \sum_{i=1}^{m_1} u_1^i f_i(\bar{x}_1), \\ \vdots \\ \dot{\bar{x}}_N = f_0^N(\bar{x}_N) + \sum_{i=1}^{m_N} u_N^i f_i(\bar{x}_N), \end{cases} \quad (5)$$

and a cost of the form

$$L(x, u) = \sum_{i=1}^N \bar{u}_i^T R_i(\bar{x}_i) \bar{u}_i. \quad (6)$$

The corresponding optimal control problem appears as a juxtaposition of N different problems that can be treated separately, and its solution is the product of the N corresponding solutions. Thus, for any N -tuple $\alpha_1, \dots, \alpha_N$ of positive real numbers, the cost

$$\tilde{L}(x, u) = \sum_{i=1}^N \alpha_i \bar{u}_i^T R_i(\bar{x}_i) \bar{u}_i$$

is equivalent via minimizers to L , and \tilde{L} is non-proportional to L if the α_i 's are not all equal to each other.

This example suggest to introduce the following definition.

Definition 2.3: We say that an optimal control problem defined by (1) and (3) admits a product structure near $q_0 \in M$ if there exist local coordinates x in which the dynamics and the cost take the form (5) and (6) respectively.

We actually conjecture that, when a cost in \mathcal{L} admits a non proportional equivalent cost, then the corresponding problem has necessarily this product structure. The purpose of this paper is to move towards this conjecture.

C. Cost equivalence via geodesics

Pontryagin Maximum Principle gives a necessary conditions for a curve to be minimizing for the optimal control problem (2). Recall that an extremal of (2) is a curve $(q, p)(\cdot)$ on T^*M solution of the Hamiltonian equations arising from the maximization w.r.t. u of

$$H(q, p, u, p^0) = \langle p, f_0(q) + \sum_{i=1}^m u_i f_i(q) \rangle + p^0 L(q, u),$$

where $p^0 \leq 0$ is a scalar. There are two kinds of extremals, the normal and abnormal ones, corresponding to $p^0 = -\frac{1}{2}$ and $p^0 = 0$ respectively. The projection of a normal (resp. abnormal) extremal onto M is called a normal (resp. abnormal) geodesic. A geodesic can be both normal and abnormal, and it is called strictly normal when it is a projection only of normal extremals.

Definition 2.4: We say that two costs L and \tilde{L} are equivalent via geodesics if the corresponding classes of optimal control problems have the same geodesics.

Remark 2.5: Abnormal geodesics depend only on the control system and are independent of the cost. So, they are the same for all costs in our class and are irrelevant for the inverse problem. As a consequence, for two costs to be equivalent via geodesics we only need the costs to have the same strictly normal geodesics.

Let $h(q, p) = \max_u H(q, p, u, -\frac{1}{2})$ be the (maximized) normal Hamiltonian. Then the normal geodesics are the projections on M of the integral curves of the Hamiltonian flow, i.e. $\gamma(t) = \pi(e^{t\vec{h}}\lambda)$ for $\lambda \in T^*M$. Moreover, by the strong Legendre condition normal geodesics are minimizers for small time [7, Th. 1.6], thus the same technique as in the sub-Riemannian case in [10] allows to prove the following Lemma.

Lemma 2.6: If two costs in \mathcal{L} are equivalent via minimizers, then they are equivalent via geodesics. Lemma 2.6 permits to reduce the study of equivalence via minimizers to the study of equivalence via geodesics.

Corollary 2.7: If there do not exist pairs of non proportional costs in \mathcal{L} that are equivalent via geodesics, then the inverse optimal control problem is injective.

III. ORBITAL DIFFEOMORPHISM

We have seen that we can replace minimizers by geodesics in the study of equivalence of costs. We show now that the equivalence via geodesics extends to a relation between the normal Hamiltonians called orbital diffeomorphism. As stated in the introduction, the results are given without proof, for which we refer to [16, Chap. 1 & 4].

A. From geodesics to extremals

Let $q_u(\cdot)$ be a geodesic of (2) defined on $[0, T]$, and consider the linearized system around q_u in some local coordinates x ,

$$\dot{x}(t) = A(t)x(t) + B(t)v(t),$$

where the matrices $A(t)$ and $B(t)$ are defined as

$$A(t) = \frac{\partial f_0}{\partial x}(q_u(t), u(t)) + \sum_{i=1}^m u_i \frac{\partial f_i}{\partial x}(q_u(t), u(t)),$$

$$B(t) = (f_1(q_u(t)) \quad \cdots \quad f_m(q_u(t))).$$

Consider the sequence of matrix B_i , $i \in \mathbb{N}$, defined by

$$B_0(t) = B(t), \quad B_{i+1}(t) = A(t)B_i(t) - \frac{dB_i(t)}{dt}.$$

Definition 3.1: A geodesic q_u is ample at t if and only if there holds $\text{span}\{B_i(t)v : v \in \mathbb{R}^m, i \in \mathbb{N}\} = \mathbb{R}^n$. A covector $\lambda \in T^*M$ is ample if the geodesic $q_u(t) = \pi(e^{t\vec{h}}\lambda)$ is ample at $t = 0$.

Notice that if a geodesic is ample at $t = 0$, then it is strictly normal on any small enough interval $[0, \varepsilon]$ (see [1, Prop. 3.6]). Ample geodesics play a crucial role in the study of equivalence of costs because they are the geodesics characterized by their jets. Let us precise this fact. Fix a nonnegative integer k . For a given curve $\gamma : I \rightarrow M$, $I \subset \mathbb{R}$, denote by $j_{t_0}^k \gamma$ the k -jet of γ at the point t_0 . Given $q \in M$, we denote by $J_q^k(L)$ the space of k -jets at $t = 0$ of the normal geodesics of L issued from q . We set $J^k(L) = \bigsqcup_{q \in U} J_q^k(L)$

and we define the maps $P^k : T^*M \mapsto J^k(L)$, by

$$P^k(\lambda) = j_0^k \pi(e^{t\vec{h}}\lambda).$$

Lemma 3.2: Let $\lambda \in T^*M$ be ample. Then there exists an integer k_0 such that the map P^{k_0} is an immersion at λ .

Thus the existence of ample geodesics is crucial to recover the extremals from the geodesics. In the driftless case, i.e. when $f_0 \equiv 0$, the existence of ample geodesics was proven in [1]. For proper control-affine systems, the existence of an ample geodesic starting at a given point is still an open question except in the analytic case, where this existence can be deduced from [3, Theorem 2].

Theorem 3.3: Assume that the manifold M , the vector fields f_0, \dots, f_m , and the cost L are analytic, and that Assumption 1 is satisfied. Then, for any point $q \in M$ there exists an ample $\lambda \in T^*M$ such that $\pi(\lambda) = q$. Moreover the set of ample covectors is open and dense in T^*M .

B. Orbital diffeomorphism

Consider two costs L_1 and L_2 in \mathcal{L} . We denote by h_1 and h_2 the respective normal Hamiltonians of L_1 and L_2 .

Definition 3.4: We say that \vec{h}_1 and \vec{h}_2 are orbitally diffeomorphic on an open subset V_1 of T^*M if there exists an open subset V_2 of T^*M and a diffeomorphism $\Phi : V_1 \rightarrow V_2$ such that Φ is fiber-preserving, i.e. $\pi(\Phi(\lambda)) = \pi(\lambda)$, and Φ sends the integral curves of \vec{h}_1 to the integral curves of \vec{h}_2 , i.e. $\Phi(e^{t\vec{h}_1}\lambda) = e^{t\vec{h}_2}(\Phi(\lambda))$ for all $\lambda \in V$ and $t \in \mathbb{R}$ for which $e^{t\vec{h}_1}\lambda$ is well defined, or, equivalently

$$d\Phi \circ \vec{h}_1(\lambda) = \vec{h}_2(\Phi(\lambda)).$$

We say that Φ is an orbital diffeomorphism from \vec{h}_1 to \vec{h}_2 .

Proposition 3.5: If \vec{h}_1 and \vec{h}_2 are orbitally diffeomorphic on a neighborhood of $\pi^{-1}(q_0)$, then L_1, L_2 are equivalent via geodesics at q_0 .

The converse statement holds near ample geodesics.

Proposition 3.6: Assume that the Lagrangians L_1 and L_2 are equivalent via geodesics at q_0 . Then, for any $\lambda_1 \in \pi^{-1}(q_0)$ ample with respect to L_1 , \vec{h}_1 and \vec{h}_2 are orbitally diffeomorphic on a neighborhood V_1 of λ_1 in T^*M .

The idea behind this result is to construct the orbital diffeomorphism near ample covectors by using the maps $P_i^{k_0} : T^*M \mapsto J^{k_0}(L_i)$, which are locally diffeomorphic on their image, and the identification between the geodesics (and so between their jets) of two equivalent costs:

$$\begin{array}{ccc} V_1 & \xrightarrow{\Phi} & V_2 \\ P_1^{k_0} \downarrow & & \downarrow P_2^{k_0} \\ P_1^{k_0}(V_1) \subset J^{k_0}(L_1) & \xrightarrow{\text{Id}} & P_2^{k_0}(V_2) \subset J^{k_0}(L_2) \end{array}$$

C. Coordinates of an orbital diffeomorphism

From Proposition 3.6, the existence of an orbital diffeomorphism is a necessary condition for the equivalence between two costs. Let us give an expression of such a diffeomorphism in well chosen local coordinates.

Consider two costs in \mathcal{L} ,

$$L_1(q, u) = u^T R_1(q)u \quad \text{and} \quad L_2(q, u) = u^T R_2(q)u.$$

At each point $q \in M$ we define the transition matrix

$$S(q) = R_1(q)^{-1}R_2(q),$$

which is a $(m \times m)$ positive definite matrix. Thus its eigenvalues $\alpha_1(q), \dots, \alpha_m(q)$ are positive real numbers.

Definition 3.7: A point $q_0 \in M$ is called stable with respect to the pair (R_1, R_2) if q_0 is a regular point and if the number $N(q)$ of distinct eigenvalues of $S(q)$ is constant in some neighborhood of q_0 .

Notice that the set of stable points is open and dense in M .

Fix a point $q_0 \in M$ which is stable with respect to (R_1, R_2) and a neighborhood U of q_0 containing only stable points. For $q \in U$, the matrices $R_1(q)$ and $R_2(q)$ can be simultaneously diagonalized, i.e. there exists an orthogonal matrix $P(q)$, with $q \mapsto P(q)$ smooth, such that $P(q)^T R_1(q) P(q)$ is the identity matrix and $P(q)^T R_2(q) P(q)$ is a diagonal matrix with $\alpha_1(q), \dots, \alpha_m(q)$ on the diagonal. Thus, up to a feedback transformation $u \mapsto P(q)u$, the costs L_1, L_2 take the form

$$L_1(q, u) = \sum_{i=1}^m u_i^2 \quad \text{and} \quad L_2(q, u) = \sum_{i=1}^m \alpha_i(q) u_i^2. \quad (7)$$

Note that such feedback transformations are allowed since they preserve the set of geodesics.

Choose vector fields f_{m+2}, \dots, f_n whose values complete $f_1, \dots, f_m, f_{m+1} = f_0$ into a basis of the tangent space on U . For all $q \in U$, the functions $T_q^*M \ni p \mapsto \nu_i = \langle p, f_i(q) \rangle$, $i = 1, \dots, n$, define coordinates on T_q^*M . In

these coordinates the Hamiltonians h_1, h_2 corresponding to the costs L_1, L_2 respectively can be expressed as:

$$h_1 = \nu_{m+1} + \frac{1}{2} \sum_{i=1}^m \nu_i^2, \quad h_2 = \nu_{m+1} + \frac{1}{2} \sum_{i=1}^m \frac{\nu_i^2}{\alpha_i}. \quad (8)$$

Let Φ be an orbital diffeomorphism from \vec{h}_1 to \vec{h}_2 on some neighborhood $V \subset T^*U$ of λ_0 such that $\pi(\lambda_0) = q_0$. In coordinates on T^*U , Φ maps $\lambda = (q; \nu_1, \dots, \nu_n)$ to $(q; \Phi_1, \dots, \Phi_n)$ where $\Phi_i(\lambda) = \nu_i(\Phi(\lambda))$.

Proposition 3.8: If Φ is an orbital diffeomorphism from \vec{h}_1 to \vec{h}_2 with coordinates (Φ_1, \dots, Φ_n) , then:

- for $k = 1, \dots, m$, there holds $\Phi_k = \alpha_k \nu_k$;
- $\tilde{\Phi} = (\Phi_{m+1}, \dots, \Phi_n)$ satisfies a linear system of equations,

$$A(\nu)\tilde{\Phi} = b(\nu), \quad (9)$$

where $A(\nu)$ is a matrix with $(n-m)$ columns and an infinite number of rows, and $b(\nu)$ is a column vector with an infinite number of rows. The components of these matrices are polynomial functions in $\nu = (\nu_1, \dots, \nu_n)$ whose coefficients depend only on f_1, \dots, f_n for A , and only on f_1, \dots, f_n and $\alpha_1, \dots, \alpha_m$ for b .

We do not give here the general expression of A and b (see [16, Chap. 4]), we will do it only in the case $m = n - 1$ in Sect. V.

Remark 3.9: The fact that all the coefficients of the system (9) are polynomial in ν implies that, if there exists $\lambda_0 \in T_{q_0}^*M$ with coordinates ν_0 for which $A(\nu_0)$ has a nonzero minor of size $n - m$, then the corresponding minor of $A(\nu)$ is a nonzero polynomial of ν and the matrix is injective almost everywhere on $T_{q_0}^*M$. The injectivity of A implies the existence of a unique solution $\tilde{\Phi}$.

The matrix A is strongly related with the characterization of ample covectors.

Proposition 3.10: A covector $\lambda \in T^*M$ is ample with respect to L_1 if and only if $A(\nu(\lambda))$ is injective.

Note that, for two proportional costs L_1 and $L_2 = cL_1$ which are trivially equivalent via geodesics, all eigenvalues α_i are equal to the constant c . The corresponding orbital diffeomorphism is proportional to the identity, i.e. $\Phi_i(\nu) = c\nu_i$ for $i = 1, \dots, n$.

Corollary 3.11: If there exists an ample λ such that $A(\nu)\tilde{\Phi} = b(\nu)$ admits solutions near $\nu(\lambda)$ only when $\alpha_1 = \dots = \alpha_m = \text{const}$, then the inverse optimal control problem is injective.

This result shows the way to study the injectivity of the inverse problem through the analysis of the system (9). However the latter analysis requires to introduce the notion of nilpotent approximation, for which ample covectors exist (analytic system), and the system $A(\nu)\tilde{\Phi} = b(\nu)$ is simpler since the components of A and b are homogeneous polynomials.

IV. NILPOTENT APPROXIMATION

Let us briefly present the nilpotent approximation for control-affine systems defined in [6]. See [16, Chap. 4] for the details and the proofs of the lemmas.

Let f_0, \dots, f_m be vector fields satisfying Assumptions 1, $q_0 \in M$ be a regular point, and U be a neighborhood of q_0 with only regular points. The construction relies on a graded structure defined by assigning some weights to the vector fields f_i 's: we choose a weight 2 for f_0 and weights 1 for f_1, \dots, f_m so that the normal Hamiltonian (8) is homogeneous of weighted degree 2. These weights induce a filtration of the Lie algebra spanned by f_0, \dots, f_m in the following way. Let us define a sequence of subspaces of the set of vector fields by

$$\begin{aligned} D_1 &= \text{span}\{f_1, \dots, f_m\}; \\ D_2 &= \text{span}\{f_0, [f_i, f_j], f_i : i, j = 1, \dots, m\}; \\ D_3 &= \text{span}\{[f_0, f_i], [[f_0, f_i], f_k], f_0, [f_i, f_j], f_i : i, j, k\}; \\ &\vdots \end{aligned}$$

By the weak Hörmander condition (4), there exists $r \in \mathbb{N}$ such that $D_r(q) = T_q M$ for any $q \in U$. As a consequence, the following filtration of vector spaces is well defined for $q \in U$,

$$D_1(q) \subset D_2(q) \subset \dots \subset D_r(q) = T_q M. \quad (10)$$

The filtration defines a grading structure and weighted degrees of vector fields and functions. Moreover, by [6], there exist coordinates on a neighborhood of q_0 adapted to the grading. Based on these coordinates we can construct the vector fields $\hat{f}_0, \dots, \hat{f}_m$, each \hat{f}_i being the homogeneous part of the highest weighted degree in the Taylor expansion of f_i , and $\hat{f}_i(q_0) = f_i(q_0)$ for $i = 0, 1, \dots, m$. Therefore, \hat{f}_0 is of weighted degree 2 and \hat{f}_i is of weighted degree 1 for any $i = 1, \dots, m$. The vector fields $\hat{f}_0, \dots, \hat{f}_m$ are called a nilpotent approximation of f_0, \dots, f_m . We can construct a frame f_1, \dots, f_n of TM adapted to the filtration (10) and the corresponding frame of the nilpotent approximation $\hat{f}_1, \dots, \hat{f}_n$ which consist of homogeneous polynomial vector fields on \mathbb{R}^n , each \hat{f}_i being of weighted degree k .

Let A and b be the matrices from (9) corresponding to the adapted frame f_1, \dots, f_n and to some positive functions $\alpha_1, \dots, \alpha_m$. We denote by \hat{A} and \hat{b} the matrices corresponding to $\hat{f}_1, \dots, \hat{f}_n$ and to the constant functions $\alpha_1(q_0), \dots, \alpha_m(q_0)$.

Lemma 4.1: Any component \hat{A}_{ij} (resp. \hat{b}_j) of the matrix \hat{A} (resp. \hat{b}) is either zero or the homogeneous polynomial of highest weighted degree of the corresponding component A_{ij} of A (resp. b_j of b). Any minor of \hat{A} is either zero or the homogeneous polynomial of highest weighted degree of the corresponding minor of A . As a consequence the injectivity of \hat{A} implies the injectivity of A .

Notice that the so-obtained nilpotent approximation satisfies

$$\text{Lie}_{q_0}(f_0, \dots, f_m) = \text{Lie}_{q_0}(\hat{f}_0, \dots, \hat{f}_m),$$

however the following inclusion,

$$\begin{aligned} &\text{Lie}_q \left(\{(\text{ad} \hat{f}_0)^s \hat{f}_i : s \geq 0, i = 1, \dots, m\} \right) \\ &\subset \text{Lie}_q \left(\{(\text{ad} f_0)^s f_i : s \geq 0, i = 1, \dots, m\} \right), \end{aligned}$$

may be strict, and in this case the weak Hörmander condition (4) does not hold for the nilpotent approximation. We can ensure this condition by introducing a supplementary assumption on the vector fields f_0, f_1, \dots, f_m .

Assumption 2: For q in an open and dense subset of M , there holds

$$f_0(q) \in \text{span}\{f_i, [f_i, f_j] : i, j = 1, \dots, m\}(q). \quad (11)$$

To simplify the notations, we modify the definition of a regular point by requiring also the condition (11) (the set of regular points is still an open and dense subset of M).

Lemma 4.2: If f_0, \dots, f_m satisfy Assumptions 1 and 2, then the nilpotent approximation $\hat{f}_0, \dots, \hat{f}_m$ satisfies Assumption 1.

This allows to prove the existence of ample geodesics, a result that has its own importance beyond the applications to inverse optimal control.

Theorem 4.3: Let f_0, \dots, f_m be vector fields satisfying Assumptions 1 and 2, and $L \in \mathcal{L}$. At any regular point $q \in M$ the control-affine system (1) admits a covector $\lambda \in T_q^* M$ ample with respect to L .

Proof: First, up to feedback equivalence, we can assume $L = u^T u$. Consider the nilpotent approximation at a regular point q with the cost L , which are both algebraic and thus analytic. From Lemma 4.2, $\hat{f}_0, \dots, \hat{f}_m$ satisfies Assumption 1, hence by Theorem 3.3 there exist $\lambda \in T_q^* M$ ample with respect to L for the nilpotent approximation. By Proposition 3.10, we get that $\hat{A}(\lambda)$ is injective, which implies that $A(\lambda)$ itself is injective by Lemma 4.1. We conclude by Proposition 3.10. \blacksquare

V. APPLICATION TO THE CASE $m = n - 1$

As a first application of the approach developed in this paper we study the case where $m = n - 1$. Let f_0, \dots, f_m , with $m = n - 1$, be vector fields satisfying Assumptions 1 and 2, and let L_1, L_2 be two costs in \mathcal{L} . Near a stable point q_0 , we can assume up to a feedback transformation that L_1, L_2 are of the form (7).

Proposition 5.1: Assume the costs L_1, L_2 to be non proportional and equivalent via geodesics in a neighborhood of a stable point q_0 . Then the control-affine problem corresponding to the nilpotent approximation $\hat{f}_0, \dots, \hat{f}_m$ at q_0 and to L_1 admits a product structure.

Proof: By Theorem 4.3, there exists an ample covector in $T_{q_0}^* M$, and thus by Prop. 3.6 there exists an orbital diffeomorphism Φ from \vec{h}_1 to \vec{h}_2 . As a consequence the linear system (9) admits a solution near the ample covectors, which implies that the rank of $\begin{pmatrix} A & b \end{pmatrix}$ is smaller than $n - m + 1$. By Lemma 4.1, the rank of $\begin{pmatrix} \hat{A} & \hat{b} \end{pmatrix}$ is smaller than $n - m + 1$ as well, and thus the system $\hat{A}\tilde{\Phi} = \hat{b}$ admits a solution.

Let us study this system. Since $n = m + 1$, it contains only one unknown variable $\Phi_{m+1} = \Phi_n$. Hence \hat{A} is a column vector, as \hat{b} , and $\hat{A}\tilde{\Phi} = \hat{b}$ writes as

$$\hat{A}_j \Phi_n = \hat{b}_j, \quad j \in \mathbb{N}.$$

The induction formula defining \hat{A} and \hat{b} is given in [16, Prop. 4.10]. Note that, with our assumptions, we can take $f_1, \dots, f_m, f_n = f_0$ as a frame adapted to the filtration (10) and we define coefficients \hat{c}_{ij}^k by

$$[\hat{f}_i, \hat{f}_j] = \sum_{k=1}^n \hat{c}_{ij}^k \hat{f}_k, \quad i, j = 1, \dots, m.$$

With these notations, the first components of \hat{A} and \hat{b} are

$$\hat{A}_j = \sum_{i=1}^m \hat{c}_{ij}^k \nu_i, \quad \hat{b}_j = \hat{\alpha}_j \hat{A}_j \nu_n, \quad j = 1, \dots, m,$$

where $\hat{\alpha}_j = \alpha_j(q_0)$. Then

$$\Phi_n = \hat{\alpha}_j \nu_n$$

for any j such that $\hat{A}_j \neq 0$. From Assumptions 1 and 2, at least one coefficient \hat{c}_{ij}^{m+1} is nonzero. Therefore there exists j^* such that $\hat{A}_{j^*} \neq 0$ and $\Phi_n = \hat{\alpha}_{j^*} \nu_n$. This implies that, for any $j = 1, \dots, m$ such that $\hat{\alpha}_j \neq \hat{\alpha}_{j^*}$, we have $\hat{A}_j = 0$ and thus $\hat{c}_{ij}^n = 0$ for any $i = 1, \dots, m$. On the other hand, the nilpotent approximation being homogeneous, $k < n$ implies $\hat{c}_{ij}^k = 0$. As a consequence, if $\hat{\alpha}_i \neq \hat{\alpha}_{j^*}$ then $[\hat{f}_i, \hat{f}_j] = 0$ for any $j = 1, \dots, n$.

Let us define local coordinates $x = (x_1, \dots, x_n)$ as the inverse of the local diffeomorphism

$$x \mapsto \exp\left(x_n \hat{f}_0 + \sum_{\{i: \hat{\alpha}_i = \hat{\alpha}_{j^*}\}} x_i \hat{f}_i\right) \circ \exp\left(\sum_{\{i: \hat{\alpha}_i \neq \hat{\alpha}_{j^*}\}} x_i \hat{f}_i\right),$$

and decompose x as $x = (\bar{x}_1, \bar{x}_2)$, where $\bar{x}_1 = (x_n, \{x_i : \hat{\alpha}_i = \hat{\alpha}_{j^*}\})$ and $\bar{x}_2 = (\{x_i : \hat{\alpha}_i \neq \hat{\alpha}_{j^*}\})$. Note that the $\hat{\alpha}_i$'s are not all equal to each other since L_1 and L_2 are assumed to be non proportional. In these coordinates, the control-affine system defined by the nilpotent approximation takes the form

$$\begin{cases} \dot{\bar{x}}_1 = \hat{f}_0(\bar{x}_1) + \sum_{\{i: \hat{\alpha}_i = \hat{\alpha}_{j^*}\}} u_i \hat{f}_i(\bar{x}_1), \\ \dot{\bar{x}}_2 = \sum_{\{i: \hat{\alpha}_i \neq \hat{\alpha}_{j^*}\}} u_i \hat{f}_i(\bar{x}_2). \end{cases}$$

This dynamics is of the form (5) and together with the cost

$$L = \sum_{\{i: \hat{\alpha}_i = \hat{\alpha}_{j^*}\}} u_i^2 + \sum_{\{i: \hat{\alpha}_i \neq \hat{\alpha}_{j^*}\}} u_i^2$$

they form a control-affine problem with a product structure, which ends the proof. ■

This theorem shows that, if for a given control-affine system there exists non proportional equivalent costs, then necessarily the nilpotent approximation must admit a product structure, which is a very specific condition. For instance, generically an element (f_0, \dots, f_m) in the set $\text{VF}(M)^n$ of n -tuple of vector fields satisfy Assumptions 1 and 2. The product structure of the nilpotent approximation implies that the corank distribution spanned by f_1, \dots, f_m admits as well a nilpotent approximation with product structure. Using the characterization of such distributions in [11] we get the following result.

Corollary 5.2: Assume $n = m + 1$ is odd. Then for (f_0, f_1, \dots, f_m) in an open and dense subset of $\text{VF}(M)^n$, the associated inverse optimal control problem is injective.

VI. CONCLUSIONS AND FUTURE WORKS

In this paper we present a general approach to study the injectivity of the inverse optimal control problems applicable to a wide class of problems. We apply this approach to the special class of optimal control problems with $m = n - 1$ controls. In this case we show that a non-injective cases have the spacial structure at the level of the nilpotent approximation. We show as well that a generic control-affine problem on odd-dimensional manifold is injective.

Still in that case $m = n - 1$, we plan to extend the results on the product structure obtained for the nilpotent approximation to the affine-control problem itself. A more general idea is to relate the product structure of an optimal control problem with non-injectivity of the inverse problem via decomposition of the orbital diffeomorphism in the independent components.

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