

Feedback Classification and Optimal Control with Applications to the Controlled Lotka-Volterra Model

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1 ARTICLE TEMPLATE

$_{\rm 2}$ $\,$ Feedback Classification and Optimal Control with Applications to the $\,$

3 Controlled Lotka–Volterra Model

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7 ARTICLE HISTORY

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35 36	Let <i>M</i> be a σ -compact C^{∞} manifold of dimension $n \ge 2$ and consider a single–input control system: $\dot{x}(t) = X(x(t)) + u(t) Y(x(t))$, where <i>X</i> , <i>Y</i> are C^{∞} vector fields on <i>M</i> . We prove					

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- that there exist an open set of pairs (X, Y) for the C^{∞} -Whitney topology such that they
- admit singular abnormal rays so that the spectrum of the projective singular Hamiltonian
- 39 dynamics is feedback invariant. It is applied to controlled Lotka–Volterra dynamics where
- 40 such rays are related to shifted equilibria of the free dynamics.
- 41 KEYWORDS
- 42 Feedback classification; Nonlinear systems; Lotka–Volterra model; Optimal control; Direct
- 43 numerical methods

44 1. Introduction

⁴⁵ Consider a single–input affine control system on a σ –compact C^{∞} manifold of dimension ⁴⁶ $n \ge 2$ defined by

$$\frac{\mathrm{d}x}{\mathrm{d}t}(t) = X(x(t)) + u(t) Y(x(t)), \tag{1}$$

- where *X*, *Y* are C^{∞} vector fields on *M* and the set of admissible controls \mathscr{U} is the set of bounded measurable mappings $u : [0, T(u)] \mapsto \mathbb{R}, T(u) > 0$.
- Let (X, Y) and (X', Y') be two such pairs. They are called *feedback equivalent* if there exist a C^{∞} diffeomorphism φ on M and a feedback $u = \alpha(x) + \beta(x)u'$, where α, β are C^{∞} mappings and β invertible so that

52 (i)
$$X' = \varphi * X + (\varphi * Y) \cdot \alpha$$
,

53 (ii) $Y' = (\varphi * Y) \cdot \beta$,

⁵⁴ where if Z is a C^{∞} vector field $\varphi * Z$ is the image of Z by φ given in local coordinates by

$$\varphi * Z = \left(\frac{\partial \varphi}{\partial x}\right)^{-1} \left(Z \circ \varphi\right)$$

⁵⁵ This action defines a group structure G_f on the set of triplets $\{g = (\varphi, \alpha, \beta)\}$ called the *feed-back group*.

⁵⁷ It is well known from geometric linear system theory [17] that, restricting to linear au-

tonomous controllable systems: $\dot{x}(t) = Ax(t) + u(t)b$, *A* being a constant matrix and *b* a constant vector, every such pairs (*A*, *b*) and (*A'*, *b'*) are feedback equivalent, restricting to linear diffeomorphisms and feedbacks: $u = \alpha \cdot x + \beta u'$, $\beta \neq 0$ constant.

The feedback equivalence for control affine systems was studied in the earliest reference 61 [3] in relation with the time-minimal geodesics corresponding to the so-called *singular* 62 *trajectories*. More precisely consider the pair (X, Y) and denote by $x(., x_0, u)$ the response 63 to $u(.) \in \mathcal{U}$ starting at time t = 0 from x_0 . Fixing t_f , the *fixed extremity mapping* at time 64 t_f is the map $E^{x_0, t_f} : \mathcal{U} \ni u \mapsto x(t_f, x_0, u) \in M$. If t_f is free, the extremity mapping is the 65 map $E^{x_0}: \mathcal{U} \ni u \mapsto x(., x_0, u)$. If we endow the set of controls with the L^{∞} -norm topology 66 both maps are Fréchet differentiable and they have singularities, that is the image of L^{∞} 67 by the derivative is not of full rank n. Such a pair (x(.), u(.)) is called *singular* for the fixed 68 time case and if the final time is not fixed such singular trajectories are called exceptional 69 or abnormal. 70

Singular trajectories can be parametrized thanks to the Maximum Principle [11] as projection of *singular extremal* pairs (z(.), u(.)), z = (x, p) where p is non vanishing adjoint vector solutions of the Hamiltonian dynamics:

$$\frac{\mathrm{d}z}{\mathrm{d}t}(t) = H_X(z(t)) + u(t) H_Y(z(t)),$$

$$H_Y(z(t)) = 0,$$
(2)

where $H_X = p \cdot X(x)$ and $H_Y = p \cdot Y(x)$ are respectively the Hamiltonian lifts of X, Y, H_X + 74 uH_Y is the pseudo Hamiltonian and additionally in the abnormal case, we have: 75

$$H_X(z(t)) = 0.$$

Given a pair (X, Y) the *collinearity set* is the set \mathscr{C} of points $x \in M$ such that X(x) and 76 Y(x) are linearly dependent. Clearly it is a feedback invariant. Taking a point x_e in \mathscr{C} there 77 exist a constant u_e so that $X(x_e) + u_e Y(x_e) = 0$ and the pair (x_e, u_e) is called a forced equili-78 brium. 79

The contribution of this article is to construct if $n \ge 2$ a set of pairs (X, Y) of codimension 80 *n* in the jets space such that there exist abnormal trajectories reduced to isolated point x_0 , 81 which can be lifted into lines ℓ in the projective space $PT_{x_0}^*M$ called *abnormal rays* and 82 the whole spectrum of the projectivized linearized dynamics defined by (2) is feedback 83 invariant. 84

This result can be illustrated by the following 2*d*-case, $x = (x_1, x_2)$, 85

$$\dot{x}_1 = \lambda x_1 + x_2^2$$

$$\dot{x}_2 = u$$
(3)

in which the singular line is the axis $x_2 = 0$ on which the singular dynamics is given by $\dot{x}_1 = \lambda x_1$ and the abnormal point is the equilibrium point 0. Denoting $p = (p_1, p_2)$ the 87

adjoint vector one has $p_2 = 0$ since $H_Y = 0$ and the adjoint vector is given by $\dot{p}_1 = -\lambda p_1$. Hence the ray at $T_0^* \mathbb{R}^2$ is $\ell : p_1(t) = e^{-\lambda t} p_1(0)$. The singular Hamiltonian dynamics reads 88

$$\dot{x}_1 = \lambda x_1, \qquad \dot{p}_1 = -\lambda p_1.$$

Clearly λ is feedback invariant and the projectivized dynamics is given by the dynamics of 90 $v = x_1/p_1$ with singular point at v = 0 with eigenvalue 2λ (the projectivized invariant). 91

Note that in the *n*-dimensional case, for each ray one gets 2n-3 eigenvalues to classify 92 the systems. 93

This result is applied to analyze vermin reduction for controlled Lotka-Volterra dyna-94

mics in the cone $M = \mathbb{R}_{>0}^n$ where the equations read 95

$$\frac{\mathrm{d}x}{\mathrm{d}t}(t) = (\mathrm{diag}x(t))\left(Ax(t) + r + u(t)\boldsymbol{\mathcal{E}}\right) \tag{4}$$

with $x := (x_1, \dots, x_n)^{\mathsf{T}}$ is the vector of interacting species, diagx denotes in short the diago-96 nal matrix with diagonal coefficients x_i , $A = (a_{ij})$ is the matrix of interaction coefficients 97 and $r = (r_1, ..., r_n)^{\mathsf{T}}$ is the individual growth vector without interaction. The constant vector 98 $\boldsymbol{\mathcal{E}} = (\varepsilon_1, \dots, \varepsilon_n)^{\mathsf{T}}$ describes the effect of probiotic or antibiotic agents to reduce the popula-99 tion of the infecting agent, which can be taken as the x_1 -population. The optimal control 100 problem is a *Mayer problem*: $\min_{u(.)} x_1(t_f)$, which can be formulated in the dual form as: 101 $\min_{u(.)} t_f$, $x_1(t_f) = d$, where d is a desired amount of population x_1 at final time t_f , taken as a parameter. More generally we shall consider any Mayer problem: $\min_{u(.)} \phi(x(t_f))$ with ϕ is a smooth mapping from M to \mathbb{R} .

The free motion was studied by the historical contributors Lotka and Volterra, genera-105 lizing the prey-predator model for populations dynamics [15]. The interest of the model is 106 to get easily computable equilibria using linear analysis only. In the regular case where A 107 is invertible, the *interior equilibrium* is $x_e := -A^{-1}r$ and the dynamics is related in parti-108 cular to interaction of x_e with the trivial boundary equilibrium 0. A large amount of liter-109 ature was devoted to analyze the dynamics in the 2d or 3d case starting from the seminal 110 complete analysis in the prey-predator model. In particular in the case of competitive sys-111 tems, the achievement in the 3d-case was the explicit analysis of the limit set of the origin 112 to determine the boundary of the basin of repulsion of this point. Related references are 113 [6, 16, 13] and in particular the May–Leonard model [10] will serve as case study. An im-114 portant very recent application of controlled Lotka-Volterra model comes from the con-115 trol of complex microbiota, see the work by Jones et al [7] based on the Stein et al study 116 [14] to model the C. difficile infection of the intestinal microbiote. It contains 11 interac-117 tion species and can admit up to 2^{11} equilibria. Hence the analysis is a huge task. More 118 modestly we shall concentrate our computation to specific 2d and 3d models to illustrate 119 our techniques. Besides the biological applications, the controlled Lotka–Volterra model 120 allows algebraic computations of the forced equilibria, which are by construction shifted 121 equilibria of the free equilibria and hence in fine one can compute in this case, the spec-122 trum of corresponding ray abnormal solutions. 123

This article is organized in four sections. In section 2, we make a recap of properties 124 of singular trajectories and their parametrization, thanks to the Maximum Principle as an 125 Hamiltonian dynamics in the projectivized cotangent bundle. We introduce the basic con-126 cepts of transversality theory in the jets space [9] to formulate in a neat geometric frame-127 work the main result of this article concerning feedback invariants. In section 3, we apply 128 this theorem to classify controlled Lotka-Volterra dynamics in dimension 2 or 3. We point 129 out some applications to controlled stability and time minimal syntheses. The final sec-130 tion 4 relates those calculations based on the Maximum Principle in the permanent case 131 to the computations of optimal solution of vermin reduction in the sampled-data control 132 frame. 133

134 2. A brief recap about singular trajectories and main result

This section is based on the results of [3] and [4], which are briefly presented to deduce the classification result. It combines the techniques from transversality theory in the jets space, see [9], and elementary geometric invariant theory [5].

138 **2.1.** Notations and general results

Let *M* be a σ -compact C^{∞} (smooth) manifold of dimension $n \ge 2$. We introduce the following notations:

- *TM* is the tangent space of *M* and T_xM is the tangent space at $x \in M$.
- T^*M is the cotangent space of M and T_x^*M is the cotangent space at $x \in M$. The null section of T^*M is denoted by 0 and $(T^*M)_0 = T^*M \setminus \{0\}$. We note PT^*M the projectivized cotangent space i.e. $PT^*M = T^*M/\mathbb{R}^*$ and [z] is the equivalence class of z in PT^*M .
- For any integer N, $J^N TM$ is the space of all N-jets of vector fields i.e. Taylor expan-

sions up to order N and $JTM = \bigcup_{N \ge 0} J^N TM$ is the jets space. 147 • VF(M) is the vector space of all smooth vector fields on M endowed with the Whit-148 ney topology. 149 • We denote by z = (x, p) the canonical coordinates on (T^*M, ω) , where ω is the Dar-150 boux form induced by the Liouville form. 151 • Take $(X, Y) \in VF(M)$, the Lie bracket is calculated in local coordinates with the con-152 vention:

$$[X, Y](x) \coloneqq \frac{\partial X}{\partial x}(x) Y(x) - \frac{\partial Y}{\partial x}(x) X(x).$$

• Given any smooth function H defined on an open subset Ω of T^*M , \tilde{H} denotes the 154 Hamiltonian vector field defined by *H* on Ω , $\vec{H} := (\partial_p H, -\partial_x H)$. Given H_1, H_2 on *M*, 155 $\{H_1, H_2\}$ denotes the Poisson bracket: 156

$$\{H_1, H_2\}(z) := \mathrm{d}H_1(\vec{H}_2(z)) = \omega(H_1, H_2),$$

- If $X \in VF(M)$, H_X denotes the Hamiltonian lift: $H_X(x) := p \cdot X(x)$. Given $(X, Y) \in$ 157 VF(M), one has $\{H_X, H_Y\} = H_{[X,Y]}$. 158
- Finally, to each pair (X, Y) of vector fields on M, we associate the single-input con-159 trol affine system: 160

$$\frac{\mathrm{d}x}{\mathrm{d}t}(t) = X(x(t)) + u(t) Y(x(t)), \quad u(t) \in \mathbb{R}, \quad x \in M.$$
(5)

The study of time-minimal trajectory of (5) leads to introduce the extremal trajectories: 161 $(z, u): [0, T] \rightarrow T^*M \times \mathbb{R}, T > 0$, such that 162

- (1) *z* is absolutely continuous, *u* is measurable and bounded, 163
- (2) $z(t) \neq 0$ (0 being the null section) for all $t \in [0, T]$, 164
- (3) $\frac{dz}{dt}(t) = \vec{H}_X(z(t)) + u(t)\vec{H}_Y(z(t))$ for a.e. $t \in [0, T]$, 165
- (4) $H_X(z(t)) + u(t) H_Y(z(t)) = \max_{v \in \mathbb{R}} H_X(z(t)) + v H_Y(z(t))$ for a.e. $t \in [0, T]$ or equiva-166 lently 167

$$H_Y(z(t)) = 0, \text{ for all } t \in [0, T]$$
 (6)

since $z \mapsto H_Y(z)$ is continuous. 168

153

Definition 2.1. A curve $(z, u) : [0, T] \mapsto T^*M \times \mathbb{R}$ satisfying the above conditions (1)–(4) is 169 called a singular extremal and its projection $(\Pi_M(z, u), u) = (x, u)$ is a singular trajectory. 170 Denoting z = (x, p), p is then the adjoint (non zero) vector. 171

Proposition 2.2. Let (z, u) = (x, p, u) be a singular extremal on [0, T]. Then: 172

• The Fréchet derivative of the fixed time extremity mapping $E^{x(0),t_f}$ along (x(.), u(.)) is 173 given by the linear dynamics: 174

$$\dot{\delta_1 x}(t) = A(t)\delta_1 x(t) + u(t)b(t) \tag{7}$$

with $A = \frac{\partial X}{\partial x}(x(t)) + u(t)\frac{\partial Y}{\partial x}(x(t))$ and b(t) = Y(x(t)) with initial condition $\delta_1 x(0) = 0$. 175

• The adjoint vector p(.) is orthogonal to the image of the Fréchet derivative called the 176

first order Pontryagin space vector. 177

One can easily calculate with the Maximum Principle many singular trajectories (x, u). Indeed deriving twice with respect to *t* the equation (6) we get for a.e. $t \in [0, T]$:

$$H_Y(z(t)) = \{H_Y, H_X\}(z(t)) = 0,$$

$$\{\{H_Y, H_X\}, H_X\}(z(t)) + u(t)\{\{H_Y, H_X\}, H_Y\}(z(t)) = 0.$$
(8)

Definition 2.3. A singular extremal (z, u) on [0, T] is called of minimal order if $\mathscr{R} = \{t \in [0, T], \{\{H_X, H_Y\}, H_Y\}(z(t)) \neq 0\}$ is dense in [0, T].

Proposition 2.4. Let (z, u) be a singular extremal and \mathcal{R} be a non empty set, then:

183 • z restricted to \mathcal{R} is smooth,

• the set $\Sigma' = \{z, H_Y(z) = \{H_Y, H_X\}(z) = 0\}$ is invariant for the singular dynamics, which is given on Σ' by:

$$\frac{\mathrm{d}z}{\mathrm{d}t}(t) = \vec{H}_X(z(t)) + u_s(z(t))\vec{H}_Y(z(t))$$
(9)

where $u_s(z)$ is the singular dynamics feedback:

$$u_{s}(z) = \frac{\{\{H_{X}, H_{Y}\}, H_{X}\}(z)}{\{\{H_{Y}, H_{X}\}, H_{Y}\}(z)}.$$
(10)

Proposition 2.5. Let (X, Y) be a pair such that the open subset Ω of all $z \in (T^*M)_0$ such that $\{z, \{\{H_X, H_Y\}, H_Y\}(z) \neq 0\}$ is not empty. Let $H_s : \Omega \to \mathbb{R}$ be the true Hamiltonian:

$$H_{s} \coloneqq H_{X} + \frac{\{\{H_{X}, H_{Y}\}, H_{X}\}(z)}{\{\{H_{Y}, H_{X}\}, H_{Y}\}(z)} H_{Y}$$
(11)

restricted to the set Σ' : {z, $H_Y(z) = {H_Y, H_X}$ (z) = 0}. Then there exist an open set in $VF(M) \times VF(M)$ such that for any couple (X, Y) in this set, Ω is open and dense and the set of all $z \in \Sigma'$ is a codimension 2 symplectic manifold of Ω for the induced symplectic form: $\omega_{|\Sigma'}$.

¹⁹² Notations 2.6. Restricting to such pairs \mathscr{G} (called good pairs), denote:

• Σ : {z, $H_Y(z) = 0$ } the switching surface.

• H_s the Hamiltonian dynamics restricted to $\Sigma' \subset \Sigma$ with $\Sigma' : H_Y(z) = \{H_Y, H_X\}(z) = 0$.

• Let λ_s be the mapping $\mathscr{G} \ni (X, Y) \mapsto (\vec{H}_s, \Sigma)$.

¹⁹⁶ 2.2. Action of the feedback group G_f on the set of good pairs

We shall briefly recall the results of [3]. First of all, borrowed from elementary geometric
invariant theory [5], we have.

Definition 2.7. Let *E*, *F* be two \mathbb{R} -vector spaces and let *G* be a group acting linearly on *E* and *F*. An homomorphism $\mathfrak{X} : G \to \mathbb{R}^*$ is called a character. Let \mathfrak{X} be a character. A semiinvariant of weight \mathfrak{X} is a map $\lambda : E \to \mathbb{R}$ such that $\forall g \in G, \forall x \in E, \lambda(g \cdot x) = \mathfrak{X}(g)\lambda(x)$. It is called an invariant if $\mathfrak{X} = 1$. A map $\lambda : E \to F$ is a semi-covariant of weight \mathfrak{X} if $\forall g \in G, \forall x \in$ *E*, $\lambda(g \cdot x) = \mathfrak{X}(g)g \cdot \lambda(x)$ and λ is called a covariant if $\mathfrak{X} = 1$.

The action of $(\varphi, \alpha, \beta) \in G_f$ can be lifted as the action of Matthieu symplectomorphisms $\vec{\varphi}$ on T^*M defined in canonical coordinates by $x = \varphi(y)$, $p = q \frac{\partial \varphi^{-1}}{\partial y}$. The action of (φ, α, β) on (\vec{H}_s, Σ) being reduced to the action of φ only. Also note that $\Sigma : H_X(z) = 0$ codes the distribution: $x \mapsto \operatorname{span} Y(x)$.

Proposition 2.8. Restricting to good pairs (X, Y), λ_s is a covariant, that is the following diagram is commutative:

$$\begin{array}{ccc} \mathscr{G} \ni (X,Y) & \xrightarrow{\lambda_s} & \lambda_s(X,Y) \\ & & & & \\ G_f \downarrow & & & \\ \mathscr{G} \ni (X',Y') & \xrightarrow{\lambda_s} & \lambda_s(X',Y') \end{array}$$

210

Definition 2.9. Let (z, u) be a singular extremal on [0, T] of minimal order. The singular extremal is called strict if the adjoint vector is unique up to a scalar (that is unique in PT^*M) on [0, T]. In the strict case, a singular trajectory $(x, u) = (\Pi_M(z), u)$ on [0, T] is called

- (1) *Abnormal* or exceptional if for every $t \in [0, T]$, $H_X(z(t)) = 0$,
- 215 (2) *Hyperbolic* if for every $t \in [0, T]$, $H_X(z(t)) \{\{H_Y, H_X\}, H_Y\}(z(t)) > 0$,

(3) *Elliptic* if for every $t \in [0, T]$, $H_X(z(t)) \{\{H_Y, H_X\}, H_Y\}(z(t)) < 0$.

According to the high order Maximum Principle [8], the hyperbolic trajectories are can didates to time-minimal control, while elliptic trajectories are candidates to time-maximal
 control, while abnormal (exceptional) can be both.

220 2.3. The 2d-case

The 2*d*-case is a specific situation but can be used to illustrate the general result avoiding technical difficulties. Without losing any generality, one can take $M = \mathbb{R}^2$ and we denote by $x = (x_1, x_2)$ the coordinates while $p = (p_1, p_2)$ denotes the non zero adjoint vector. We introduce the following determinantal sets called respectively the *singular locus* \mathscr{S} and the *collinear locus* \mathscr{C} :

$$\mathscr{S}: \{x, \det(Y, [Y, X])(x) = 0\}, \quad \mathscr{C}: \{x, \det(Y, X)(x) = 0\}.$$

One will assume that *Y* is not vanishing so that one can choose (local) coordinates with $Y = \frac{\partial}{\partial x_2}$. Moreover we assume that \mathscr{S} and \mathscr{C} are regular and intersect transversally at the point 0. If $D := \det(Y, [[Y, X], Y])$ is not vanishing when restricted to \mathscr{S} , the singular control is given by the feedback: $u_s(x) = -\frac{D'(x)}{D(x)}|_{\mathscr{S}}$, where $D' := \det(Y, [[Y, X], X])$. Moreover using $H_Y = 0$, the adjoint vector is such that $p_2 = 0$ identically.

We choose coordinates preserving *Y* so that \mathscr{S} coincides with the x_1 -axis. In a neighbourhood of 0, using the action of the feedback group, the system reads:

$$\dot{x}_1 = \lambda x_1 - x_2^2 + o_{x_1}(x_2^2),$$

$$\dot{x}_2 = u,$$
(12)

where $o_{x_1}(x_2^2)$ represents a term of order ≥ 3 in the jets space of (X, Y) along the singular lar line, identified to the x_1 -axis (the singular dynamics being feedback equivalent to the linear dynamics).

Assuming $\lambda \neq 0$ and restricting the dynamics to singular line identified to $x_2 = 0$,

straightforward computation gives us that the singular dynamics is $\dot{x}_1 = \lambda x_1$ and is foliated by the abnormal points $x_1 = 0$, the hyperbolic arc in $x_1 > 0$ and the elliptic arc in $x_1 < 0$.

The adjoint dynamics using $p_2 = 0$ is defined by the adjoint system $\dot{p}_1 = -\lambda p_1 + o_{x_1}(x_2)$. Hence linearized adjoint dynamics reads: $\dot{p}_1 = -\lambda p_1$.

In particular, the abnormal singular point x = 0 lifts into a ray in the projective bundle defined by $p_2 = 0$ and $p_1(t) = e^{-\lambda t} p_1(0)$. The linearized Hamiltonian dynamics takes the form

$$\dot{x}_1 = \lambda x_1, \quad \dot{p}_1 = -\lambda p_1.$$

²⁴⁴ Clearly λ is a feedback invariant and using the projective coordinates $v = x_1/p_1$, leads to ²⁴⁵ the dynamics $\dot{v} = 2\lambda v$ so that 2λ is a *projectivized feedback invariant*.

²⁴⁶ From this analysis one deduces the following.

Theorem 2.10. In the 2*d*-case there exist a nonempty open set of pairs (X, Y) for the Whitney topology such that:

- (1) Equilibria of the singular dynamics are isolated and reduced to abnormal equilibria x_0 defining the singular lines in M with dynamics $\dot{x} = \lambda(x_0)x$.
- (2) Each such point defines a ray $z(t) = e^{-\lambda(x_0)t} z_0$, $z_0 \in \ell$ (ℓ being a line in $T_{x_0}^* M$).
- (3) The singular line is foliated into the abnormal equilibrium and hyperbolic, elliptic
 arcs.
- (4) The eigenvalue $2\lambda(x_0)$ is a projective feedback invariant.

Remark 1. Introducing the clock one form: $\alpha = p dx$ outside the collinear set, the singular lines are the zero of $d\alpha = dp \wedge dx$.

257 **2.4.** The 3*d*-case

The previous planar case can be generalized to the three dimensional case, which is a very rich situation and can be treated similarly, paving the road to the general case.

Using (8), we introduce the determinantal mappings:

$$D = \det(Y, [Y, X], [[Y, X], Y]), \quad D' = \det(Y, [Y, X], [[Y, X], X]),$$

$$D'' = \det(Y, [Y, X], X).$$
 (13)

Assume that D is non zero, one has the following proposition.

262 **Proposition 2.11.**

• Singular trajectories of minimal order are solutions of the dynamics:

$$\frac{\mathrm{d}x}{\mathrm{d}t}(t) = X_s(x) \coloneqq X(x) + u_s(x) Y(x), \tag{14}$$

where the singular controls u_s is given by the feedback:

$$u_{s}(x) = -\frac{D'(x)}{D(x)}.$$
(15)

• The sets D'' = 0, DD'' > 0 and DD'' < 0 are invariant for the singular dynamics and correspond respectively to abnormal, hyperbolic and elliptic trajectories.

• The adjoint vector in the projective space $PT^*\mathbb{R}^3$ is uniquely defined by the relations 267 $H_Y = \{H_Y, H_X\} = 0.$ 268

Introducing the clock form $\alpha = p \, dx$ defined by the relations: $H_X = 1$, $H_Y = \{H_Y, H_X\} = 0$, 269 outside the abnormal locus D'' = 0, the singular trajectories are the characteristics of $d\alpha$. 270 Clearly we have: 271

Lemma 2.12. The feedback group G_f acts on the singular dynamics by change of coordi-272 nates only. 273

From which we deduce: 274

Proposition 2.13. The singular points of (14) are abnormal equilibria and denoting by J 275 the Jacobian matrix of (14) at such points then the whole spectrum $\sigma(J)$ is feedback inva-276 riant. Each equilibrium point x_0 defines a ray solution in the projective cotangent space 277 given by $z(t) = e^{-\lambda t} z(0)$, where x_0 is the canonical projection of z(0), $\lambda \in \sigma(J)$. 278

2.5. The general case $n \ge 3$ 279

In this case, more technicality is necessary, but the result follows mainly from the proof of 280 Lemma 1 in the reference [4]. 281

One needs the following. 282

2.5.1. Ad-condition and the bad set of finite codimension 283

Given a pair (X, Y) of vector fields denote by adX the operator defined by: $ad^0X(Y) = Y$, 284 adX(Y) = [X, Y] and inductively: $ad^k X(Y) = [ad^{k-1}X \cdot Y, Y]$ for $k \ge 2$. Denote by adH_X the 285 induced operator on the Hamiltonians: $adH_X^{l} \cdot H_Y = \{H_X, H_Y^{l}\}$. 286

Definition 2.14. For *N* large enough we define the following subset of $J^N TM \times J^N TM$: 287

couples $(j_x^N X, j_x^N Y)$ all such (1) $B'_{\ell}(N)$ subset of is the that: 288 $\dim \text{span} \{X(x), Y(x), [X, Y](x)\} \le 1.$ 289

(2) \hat{B}''_{ℓ} is the subset of $J^N TM \times J^N TM \times \mathbb{R}$ of all triples $(j_x^N X, j_x^N Y, a)$ such that: 290

(i) $Y(x) \neq 0$ 291

(ii) X(x) = aY(x)292

293

(ii) $\operatorname{dim}\operatorname{span}\left\{\operatorname{ad}^{k}G_{a}(Y)(x), 0 \le k \le n-1, [[X, Y], Y](x)\right\} < n, \text{ where } G_{a} = X - aY.$ (3) Denote by $B_{\ell}''(N)$ the canonical projection of $\hat{B}_{\ell}''(N)$ on $J^{N}TM \times J^{N}TM$. Let $B_{\ell}(N) =$ 294 $B'_{\ell} \cup B''_{\ell}(N).$ 295

From the proof of Lemma 1 in [4], one has: 296

Proposition 2.15. Let (X, Y) be a pair in $VF(M) \times VF(M)$ such that for all $x \in M$, 297 $(j_x^N X, j_x^N Y) \notin B_\ell(N)$. Then: 298

- (1) Let (z, u) be a singular extremal on [0, T] such that dim span $\{X(x(t)), Y(x(t))\} \le 1$, 299 where x is the canonical projection of z. Then x(.) is constant and is an abnormal 300 singular arc reduced to a point x_0 . 301
- (2) The extremal z(.) is of minimal order and strict and there exist a line $\ell \in T^*_{x_0}M$ so that 302 $z(t) = e^{-\lambda t} z_0, z_0 \in \ell$ and the control u is constant a.e. Hence z(.) is a ray solution. 303
- (3) The point x_0 is a forced abnormal equilibrium contained in the collinear set and 304

- let (A, b) be the linearized dynamics (7) at (x_0, u) , then (A, b) are constant and span $\{b, Ab, ..., A^{n-1}b\}$ is of codimension one. The line ℓ is orthogonal to this space.
- Theorem 2.16 (Main theorem for $n \ge 3$). There exist an open set of pairs (X, Y) for the C^{∞} -Whitney topology such that:
- (1) Every singular extremal pair (z, u) on [0, T] is with minimal order and strict.
- (2) Every singular trajectory reduced to a point x_0 is abnormal and x_0 is a forced abnormal equilibrium associated to a ray solution $z(t) = e^{-\lambda t} z(0), z(0) \in \ell$, with ℓ a line in $PT^*_{x_0}M$.
- 313 (3) Every abnormal equilibrium is isolated.
- (4) The pair (A, b) is not controllable and span $\{b, Ab, ..., A^{n-1}b\} = n-1$. The uncontrollable mode satisfies the dynamics $\dot{x} = \lambda x$.
- (5) The whole spectrum of the linearized dynamics in $PT_{x_0}^*M$ is feedback invariant.

Proof. The proof follows from [4] and standard linear geometric control theory. More precisely, from linear theory, if the pair (*A*, *b*) is controllable, the pole placement theorem [17] asserts that one can assign every spectrum using a linear gain control: u = kx. On the opposite, the non controllable modes cannot be modified. In the strict case, there exist a single uncontrollable mode.

³²² 2.5.2. Application to generic properties of pairs (X, Y, φ) .

- Practically one aim of biological models is to describe equilibria and their stability prop-323 erty. From control point of view, one can choose Y in a given class and the Mayer cost to 324 be maximized defines a family of terminal manifolds N(d) of codimension one given as 325 the level sets: $\{x, \varphi(x) = d\}$. Take a point $x \in N$, we denote by n(x) the normal vector to 326 N at x, and accessibility properties near the terminal point can be classified in a generic 327 context for the C^{∞} -Whitney topology on the triples: (X, Y, N) using the so-called transver-328 sality condition. This amounts to maximize the scalar product: $n \cdot \dot{x} = n \cdot (X + uY)$ for all u 329 in a feasible interval, which can be taken as [-1, +1]. 330
- This leads to stratify the final target N(d) into:
- The switching locus $\Sigma : \{x \in N, n(x) \cdot Y(x) = 0\}$
- The singular locus \mathscr{S} : { $x \in N$, $n(x) \cdot [Y, X](x) = 0$ }
- The exceptional locus $\mathscr{E} : \{x \in N, n(x) \cdot X(x) = 0\}.$
- In particular this leads, for fixed pairs (*X*, *Y*), to consider specific critical manifolds. In our study we consider manifolds *N* for which: $\Sigma \cap \mathscr{S} \cap \mathscr{E} \neq \emptyset$, to analyze
- time–minimal control syntheses,
- controlled stability.

This will be studied in the next sections using algebraic computations of equilibria on the Lotka–Volterra models.

341 3. Controlled Lotka–Volterra model

342 3.1. A brief recap about controlled Lotka–Volterra related to microbiote control

³⁴³ **Definition 3.1.** A controlled Lotka–Volterra dynamics is a control system of the form

$$\frac{\mathrm{d}x}{\mathrm{d}t}(t) = (\mathrm{diag}x(t))\left(Ax(t) + r + u(t)\boldsymbol{\mathcal{E}}\right)$$
(16)

with $x := (x_1, ..., x_n)^{\mathsf{T}} \in \mathbb{R}^n_{\geq 0}$, $A = (a_{ij})$, $r = (r_1, ..., r_n)^{\mathsf{T}}$ and $\mathcal{C} = (\varepsilon_1, ..., \varepsilon_n)^{\mathsf{T}}$ are constant vectors, u(t) represents the control intensity, which can be taken in [0, 1]. The free dynamics is called *regular* if the matrix *A* is invertible. One can extend the dynamics to the whole \mathbb{R}^n and the control intensity to the whole \mathbb{R} .

An *interior equilibrium* is a point x in $\mathbb{R}_{>0}^n$ such that Ax + r = 0 i.e. $x = A^{-1}r$. To each Lotka–Volterra dynamics one can assign in the regular case up to 2^n equilibria by considering all the induced Lotka–Volterra models with extinction of at most one species x_i (i.e. $x_i = 0$). For the controlled Lotka–Volterra model, the infecting agent population is denoted x_1 , hence vermin reduction aims to minimize the x_1 –population.

Lemma 3.2. Consider the dynamics (16) and let $\Omega = (K_1, ..., K_n)^T$ be an interior equilibrium. Then there exist coordinates such that the dynamics (16) can be written:

$$\dot{v}_i = -(v_i + 1) \left(\sum_{j=1}^n a_{ij}^* v_j + u \boldsymbol{\mathcal{C}} \right), \quad i = 1, \dots, n.$$
(17)

355

Proof. Let y_i be the dimensionless coordinates $y_i = x_i/K_i$, i = 1, ..., n so that the dynamics (16) with $A \leftarrow -A$ written as $\dot{x}_i = x_i \left(r_i - \sum_{j=1}^n a_{ij} x_j + u \epsilon_i \right)$, i = 1, ..., n, becomes

$$\dot{y}_i = y_i \left(r_i - \sum_{j=1}^n (a_{ij} K_i) y_j + u \epsilon_i \right), \quad i = 1, \dots, n.$$
 (18)

Denote $A^* = (a_{ij}^*) = (K_i a_{ij})$. By construction, the interior equilibrium is normalized to $\Omega = (1, ..., 1)$ so that: $r_i = \sum_{j=1}^n a_{ij}^*$. Hence (18) becomes

$$\dot{y}_i = y_i \left(\sum_{j=1}^n a_{ij}^* (1 - y_j) + u \epsilon_i \right), \quad i = 1, \dots, n.$$

³⁶⁰ Therefore if we set $v_i = y_i - 1$ the dynamics takes the form

$$\dot{\nu}_{i} = -(\nu_{i}+1) \left(\sum_{j=1}^{n} a_{ij}^{*} \nu_{j} - u \epsilon_{i} \right), \quad i = 1, \dots, n.$$
(19)

³⁶¹ It can be written shortly as:

$$\dot{v} = -(\operatorname{diag}(1+v))\left(A^*v - u\mathcal{E}\right)$$

³⁶² where the equilibrium is normalized to 0.

Therefore this triggers to consider controlled Lotka–Volterra model of the form $-(\operatorname{diag}(v+1))(Av-u\mathcal{E})$ for which we have the following Lemma.

Lemma 3.3. Consider the controlled Lotka–Volterra model $\dot{x} = -(\operatorname{diag}(x+1))(Ax - u\mathcal{E})$ with interior equilibrium x = 0. Denote $X(x) = -(\operatorname{diag}(x+1))A$, $Y(x) = (\operatorname{diag}(x+1))\mathcal{E}$ so that $-A = \frac{\partial X}{\partial x}(0)$ and $Y(0) = \mathcal{E}$. Hence span $\left\{\operatorname{ad}^{k} X \cdot Y(0), k = 0, ..., n - 1\right\} =$ span $\{\mathcal{E}, A\mathcal{E}, ..., A^{n-1}\mathcal{E}\}$. Therefore (x, u) = (0, 0) is an abnormal strict singular point if and only if the rank of the Kalman matrix $K = [\mathcal{E}, A\mathcal{E}, ..., A^{n-1}\mathcal{E}]$ is strictly less than n - 1.

Corollary 3.4. The pair (A, b) is controllable if and only if $\boldsymbol{\epsilon}$ is a cyclic vector of A.

This leads to easily characterize every pair (A, \mathcal{E}) using a Jordan decomposition of *A* such that (0,0) is an (abnormal) singular trajectory.

Example 3.5. Take a matrix *A* with Jordan blocks with equal eigenvalues e.g. the diagonal matrix $A = \text{diag}(\lambda_1, \lambda_1, \lambda_2)$. Then for every vector $\boldsymbol{\mathcal{C}}$ the pair $(A, \boldsymbol{\mathcal{C}})$ is not controllable.

375 3.2. Some examples of case studies

- 376 3.2.1. The prey-predator model
- This historical model [15], using the (x, y) coordinates, takes the form

$$\frac{\mathrm{d}x}{\mathrm{d}t} = x(\lambda_1 + \mu_1 y) + u\epsilon_1 y, \qquad \frac{\mathrm{d}y}{\mathrm{d}t} = y(\lambda_2 + \mu_2 x) + u\epsilon_2 y, \tag{20}$$

³⁷⁸ which covers both the elliptic and hyperbolic case.

The point $\Omega = (-\lambda_2/\mu_2, -\lambda_1/\mu_1) := (K_1, K_2)$ is the interior equilibrium in the quadrant x, y > 0 provided that $\lambda_1 \mu_1, \lambda_2 \mu_2 < 0$. Introduce the dimensionless coordinates $x \leftarrow x/(-\lambda_2/\mu_2)$ and $y \leftarrow y(-\lambda_1/\mu_1)$ we obtain the system:

$$\dot{x} = \lambda_1 x(1-y) + ux\epsilon_1, \qquad \dot{y} = \lambda_2 y(1-x) + ux\epsilon_2,$$

where the interior equilibrium is normalized to $\Omega = (1, 1)$.

³⁸³ The Jacobian matrix at Ω is

$$J = \begin{pmatrix} 0 & -\lambda_1 \\ -\lambda_2 & 0 \end{pmatrix}$$

and is elliptic if $\lambda_1 \lambda_2 < 0$ and hyperbolic if $\lambda_1 \lambda_2 > 0$.

The collinearity locus is \mathscr{C} : $xy [\epsilon_2 \lambda_1 (1-y) - \epsilon_1 \lambda_2 (1-x)] = 0$ and contains $\Omega = (1,1)$ by construction.

The singular locus is $\mathscr{S} : xy [\epsilon_1^2 \lambda_2 x - \epsilon_2^2 \lambda_1 y] = 0$ and defines the interior line for $xy \neq 0$ given by $y = \frac{\epsilon_1^2 \lambda_2}{\epsilon_2^2 \lambda_1} x$. Hence we have:

• *Elliptic case*: $\lambda_1 \lambda_2 < 0$. The singular line is not contained in the cone *x*, *y* > 0.

• *Hyperbolic case*: $\lambda_1 \lambda_2 > 0$. The singular line is contained in the cone of positive population *x*, *y* > 0.

The crucial point of the classification is to compute abnormal points, intersection of such two interior lines. At such point, the analysis of 3.2 applies. A richer situation making interacting the elliptic and hyperbolic cases is to consider the model:

$$\frac{dx}{dt} = (1 - u)X(x) + uY(x), \quad u \in [0, 1],$$

396 with

$$X = \lambda_1 x (1 - y) \frac{\partial}{\partial x} + \lambda_2 y (1 - x) \frac{\partial}{\partial y}, \quad Y = x (\lambda_1' + \mu_1' y) \frac{\partial}{\partial x} + y (\lambda_2' + \mu_2' x) \frac{\partial}{\partial y},$$

which leads to the system: $\dot{x} = X + u(Y - X)$ connecting the elliptic case to the hyperbolic case. It is an extension of models (16). Straightforward computations lead to an interior collinearity locus, which is defined by a quadratric mapping vs. a linear map in the preypredator model, while the interior singular locus is defined by a cubic mapping vs. a linear map. More generally automatic computations lead to analyze all the 2*d*-cases.

402 3.2.2. 3*d*-case studies

The 3d-case is a very complicated situation due to the complexity of the classification of the singular dynamics. This is not surprising in the context of dynamical systems due to the phenomenon of chaos. For Lotka–Volterra models the situation can be tamed if we consider the case of *competitive* 3d-models. Restricting to this case, we shall make a brief recap of this theory based on the presentation of [2] introducing the concept of carrying simplicia and the May–Leonard model [10] as a case study.

Definition 3.6. If the Lotka–Volterra dynamics takes the form: $\dot{x} = (\text{diag}x)(r - Ax)$, where $a_{ij}, r_i > 0$ for all i, j = 1, ..., n the system is called competitive.

⁴¹¹ **Property 3.7.** *For a competitive Lotka–Volterra system, we have:*

• The equilibrium
$$x_e = 0$$
 is an unstable node.

• Every positive trajectory in the $\mathbb{R}^n_{\geq 0}$ -space is bounded.

The interesting case, in the non trivial case, is when the dynamics admits an unique equilibrium in the cone $\mathbb{R}_{>0}^n$. Without loss of generality, it can be taken as $\Omega = (1, ..., 1)$ using Lemma 3.2.

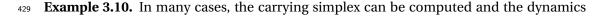
⁴¹⁷ Denoting in short by X(x) the vector field (diag x)(r - Ax) and by φ_t be the (local) pa-⁴¹⁸ rameter group $\{\exp tX\}$, one can assume that X is complete and let $\Lambda^+(x)$ denotes the ⁴¹⁹ ω -limit set of a point x. Recall that $p \in \Lambda^+(x)$ if there exist a sequence $t_k \to +\infty$ such that ⁴²⁰ $\exp t_k X(x) \to p$ when $k \to +\infty$. One can take [12] as a reference for the properties of such ⁴²¹ set in relation with stability analysis.

Definition 3.8. Since every positive trajectory is bounded in $\mathbb{R}^n_{\geq 0}$ the basin of repulsion of the unstable node $x_e = 0$ is bounded and its boundary is called the *carrying simplex* and is denoted by Π .

⁴²⁵ The following result holds [6].

Theorem 3.9. In the competitive case, every positive trajectory of the dynamics in $\mathbb{R}^n_{\geq 0} \setminus \{0\}$ is asymptotic to one trajectory in Π and Π is a Lipschitz submanifold transverse to all strictly

positive direction and homeomorphic to the probability simplex: $\Delta_1 : \sum_{i=1}^n x_i = 1$.



on Π describes the asymptotic behaviours of positive trajectories. The simplest example where it differs from Δ_1 is given by the 2*d*-system:

$$\dot{x} = x(1 - x - y/2), \quad \dot{y} = y(1 - 3x - y).$$

The only equilibria are the origin and the non interior equilibria: (1,0) and (0,1). The carrying simplex is the graph of the function $y(x) = (1 - x)^2$ on [0,1] and is a trajectory joining the saddle (0,1) to the attracting node (1,0).

⁴³⁵ Clearly the computation of this set leads to solve the global stability problem, see [16]. ⁴³⁶ This opens the road to the concept of *controlled stability* of a chosen point in Π using the ⁴³⁷ controlled Lotka–Volterra model (see Section 4.2).

438 3.2.3. The May–Leonard model [10]

It is a basic model where the carrying simplex coincides for some parameters of the model
with the probability simplex and where the dynamics can be investigated using *bifurcation analysis*. We follow the presentation of [2], see also [1].

442 **Definition 3.11.** The *May–Leonard model* is the dynamics

$$\dot{x} = x(1 - x - \alpha y - \beta z)$$

$$\dot{y} = y(1 - \beta x - y - \alpha z)$$

$$\dot{z} = z(1 - \alpha x - \beta y - z)$$
(21)

⁴⁴³ and we denote by $A = \begin{pmatrix} 1 & \alpha & \beta \\ \beta & 1 & \alpha \\ \alpha & \beta & 1 \end{pmatrix}$ the associated (circular) matrix.

444 **Property 3.12.** For α , $\beta > 0$ and $\alpha + \beta = 2$, the carrying simplex Π coincides with the proba-445 bility simplex x + y + z = 1 and eliminating in (21) the z-variable leads to the dynamics on 446 Π:

$$\dot{x} = \frac{\alpha - \beta}{2} x(1 - x - 2y)$$

$$\dot{y} = -\frac{\alpha - \beta}{2} y(1 - 2x - y)$$
(22)

⁴⁴⁷ with $\alpha + \beta = 2$. The system is conservative (i.e. with zero divergence) and we have a canonical ⁴⁴⁸ Hamiltonian system associated to

$$H(x, y) = \frac{\alpha - \beta}{2} (1 - x - y) x y$$

so that in the open triangle \mathcal{T} : {(x, y) > 0, x + y < 1} all the solutions are periodic.

More generally for arbitrary parameters α , β the eigenvalues of A can be determined, see [10] for a discussion.

The equilibria points are the origin, three single population of the form (1,0,0) and three two populations solutions of the form $(1 - \alpha, 1 - \beta, 0)/(1 - \alpha\beta)$. The interior equilibrium is $\Omega = (1, 1, 1)/(1 + \alpha + \beta)$, where the linearized dynamics is determined by the eigenvalues of *A*. Therefore the model is a good case study to apply our analysis in particular taking the control direction *Y* tangent to the 2*d*-plane. This allows to get a complete classification of all cases with respect to the parameters (ϵ_1, ϵ_2) and computations associated to this example are given in Section 4.2

460 4. Applications and numerical results

461 **4.1.** Discussion about the nonemptyness property of the good set in Theorem **2.16**

462 4.1.1. Nonemptyness in the controlled Lotka–Volterra model

Using example 3.5 from Section 3.1, consider the controlled Lotka–Volterra model: $-\operatorname{diag}(x+1) [Ax - u\mathcal{E}]$ in \mathbb{R}^3 with:

• $A = \operatorname{diag}(\lambda_1, \lambda_2, \lambda_3),$

466 • $\boldsymbol{\mathcal{E}} = (\boldsymbol{\varepsilon}_1, \boldsymbol{\varepsilon}_2, \boldsymbol{\varepsilon}_3)^{\mathsf{T}}, \, \boldsymbol{\varepsilon}_1 \neq \boldsymbol{\varepsilon}_2.$

Hence since the matrix *A* contains two identical eigenvalues, one deduces that for every *C*, the equilibrium point $x_e = 0$ is abnormal with $u_e = 0$ being the abnormal control.

Symbolic computations allow to determine the determinantal mapping D, D' defined in Section 2.4 and the singular control being $u_s(x) = -D'(x)/D(x)$. By construction at x = 0, one has $u_s(0) = u_e = 0$.

But symbolic calculation yields the following spectrum of the linearized singular dynamics at 0 :

$$(\lambda_3 - \lambda_1, -\lambda_3, -\lambda_1).$$

Hence the singular point 0 is isolated, proving that the set introduced in Theorem 2.16
 is nonempty.

The calculation can be generalized to the n-dimensional Lotka–Volterra case.

477 4.1.2. Emptyness in the quadratic case

Proposition 4.1. Let (Q, b) be a pair of vector fields on \mathbb{R}^3 with Q being a quadratic homogeneous and b being a constant vector. Then we have:

- 480 (1) $D = \det(b, [b, Q], [[b, Q], b])$: linear form
- $D' = \det(b, [b, Q], [[b, Q], Q])$: cubic form
- $D'' = \det(b, [b, Q], Q)$: cubic form.
- 483 Hence the singular dynamics

$$\frac{\mathrm{d}x}{\mathrm{d}t}(t) = Q(x(t)) - \frac{D'(x(t))}{D(x(t))}b$$

484 is homogeneous and quadratric.

(2) Let be the time reparameterization defined by $Dd\tau = dt$ so that the singular dynamics reads

$$\frac{\mathrm{d}x}{\mathrm{d}\tau}(\tau) = Q(x(\tau))D(x(\tau)) - D'(x(\tau)) b$$

and is cubic. It can be projected on the projectivized space $P(\mathbb{R}^3)$. Moreover the map

$$\hat{\lambda}_s: (Q, b) \mapsto QD - D'b$$

488 *is a semi–covariant.*

Definition 4.2. Let $\dot{x} = H(x)$ be a differential equation on \mathbb{R}^n with H being a cubic homogeneous vector field. A ray is a line ℓ so that the dynamics restricted to ℓ is given by $\dot{x}_1 = \lambda x_1^3$. In the generic case, $\lambda \neq 0$, it is an asymptotic direction but if $\lambda = 0$ it is a set of non isolated equilibria for the dynamics.

⁴⁹³ From which we deduce.

⁴⁹⁴ **Corollary 4.3.** For every pair (Q, b) on \mathbb{R}^3 , every abnormal equilibrium point is not isolated ⁴⁹⁵ so that in this family the set described in Theorem 2.16 is empty.

⁴⁹⁶ Nevertheless, one can use for pair (Q, b) the projectivized singular dynamics on $P(\mathbb{R}^3)$ to ⁴⁹⁷ compute feedback invariants. Indeed every ray projects onto an equilibrium point, where ⁴⁹⁸ we can compute the linearized dynamics. We refer to [3] for the application of this result to ⁴⁹⁹ classify the controlled Euler dynamics in the attitude control problem, where *b* describes ⁵⁰⁰ the position of gas jet on the satellite.

501 4.2. The May–Leonard model

502 4.2.1. Geometric properties

Following Section 3.2.3, we take the May–Leonard model, where the free dynamics is the restriction of (21) to the carrying simplex Π : x + y + z = 1 and the control directions are in the carrying simplex i.e. we consider the system

$$\dot{\mathbf{x}} = X(\mathbf{x}) + u Y(\mathbf{x}), \quad \mathbf{x} = (x, y), \quad u \in [0, 1]$$

506 with

$$X = x(1 - x - 2y)\frac{\partial}{\partial x} - y(1 - 2x - y)\frac{\partial}{\partial y}, \quad Y = \epsilon_1 x \frac{\partial}{\partial x} + \epsilon_2 y \frac{\partial}{\partial y}.$$
(23)

⁵⁰⁷ The singular set and the collinearity locus are given respectively by

$$\mathscr{S}: \quad \epsilon_2(\epsilon_1 + 2\epsilon_2) \, y + \epsilon_1(\epsilon_2 + 2\epsilon_1) \, x = 0 \quad \text{and} \quad \mathscr{C}: \quad (\epsilon_1 + 2\epsilon_2) \, y + (\epsilon_2 + 2\epsilon_1) \, x = \epsilon_1 + \epsilon_2.$$

For $\epsilon_1 \neq \epsilon_2$, $\epsilon_1 \neq -2\epsilon_2$ and $\epsilon_2 \neq -2\epsilon_1$, their intersection $\mathscr{S} \cap \mathscr{C}$ is the point

$$x_{se} = \left(\frac{\epsilon_2}{\epsilon_2 + 2\epsilon_1} \frac{\epsilon_1 + \epsilon_2}{\epsilon_2 - \epsilon_1}, \frac{\epsilon_1}{\epsilon_1 + 2\epsilon_2} \frac{\epsilon_1 + \epsilon_2}{\epsilon_1 - \epsilon_2}\right) = \left(\frac{1}{1 + 2\kappa} \frac{1 + \kappa}{1 - \kappa}, \frac{\kappa}{\kappa + 2} \frac{\kappa + 1}{\kappa - 1}\right), \quad \kappa = \frac{\epsilon_1}{\epsilon_2}$$

and the study of the components of x_{se} as functions of κ shows that x_{se} is in the positive orthant if $\kappa \in]-\infty, -2[\cup]-1/2, 0[$ and is outside the triangle $\mathcal{T} = \{\mathbf{x} = (x, y) > 0, x+y < 1\}$ for any value of κ (recall that in the triangle \mathcal{T} the free dynamics consists in periodic orbits).

487

For $x_e = (x_{1e}, x_{2e}) \in \mathcal{C}$, the value of the control $u_e(x_e)$ such that $X(x_e) + u_e Y(x_e) = 0$ is 512

$$u_e(x_e) = \frac{1 - 3x_{1e}}{\epsilon_1 + 2\epsilon_2}$$

and the spectrum of the Jacobian matrix $J(x_e) := \frac{\partial}{\partial \mathbf{x}} (X(\mathbf{x}) + u Y(\mathbf{x}))_{|\mathbf{x}=x_e}$ is

$$\sigma(J(x_e)) = \left\{ 3u(\epsilon_1 + \epsilon_2) \pm \frac{1}{2\sqrt{3}} \sqrt{11u^2(\epsilon_1 + \epsilon_2)^2 + 4(\epsilon_1u + 1)(\epsilon_2u - 1)} \right\}.$$

513 In particular,

• at
$$u = 0$$
, $\sigma(J(x_e)) = \left\{ \pm \frac{i}{\sqrt{3}} \right\}$,
• at $u = 1$, $\sigma(J(x_e)) = \left\{ 3(\epsilon_1 + \epsilon_2) \pm \frac{1}{2\sqrt{3}} \sqrt{11(\epsilon_1 + \epsilon_2)^2 + 4(\epsilon_1 + 1)(\epsilon_2 - 1)} \right\}$,

516

• at $u = u_e(x_{se})$, $\sigma(J(x_e)) = \left\{\frac{\epsilon_1 + \epsilon_2}{\epsilon_1 - \epsilon_2}, -\frac{\epsilon_1 + \epsilon_2}{\epsilon_1 - \epsilon_2}, \frac{3\epsilon_1\epsilon_2}{(2\epsilon_1 + \epsilon_2)(2\epsilon_2 + \epsilon_1)}\right\}$. Note that in this case, the first eigenvalue corresponds to the noncontrollable mode of the system, while the second 517 eigenvalue is not meaningful since it can be replaced by any value via pole shifting. 518

The singular dynamics $\dot{z} = \vec{H}_s(z)$ (restricted to Σ) has eigenvalues at $x = x_{se}$: 519

$$\left\{-\frac{\epsilon_1+\epsilon_2}{\epsilon_1-\epsilon_2},\frac{\epsilon_1+\epsilon_2}{\epsilon_1-\epsilon_2}\right\}$$

and $\lambda = \frac{\epsilon_1 + \epsilon_2}{\epsilon_1 - \epsilon_2}$ is a feedback invariant of the control system. 520

4.2.2. Direct and semi-direct methods 521

We present two numerical schemes to solve the time-minimal control problem for the 522 Lotka-Volterra class of models specifically a direct method and a semi-direct method. 523

The objective is to reach, from an initial position x_0 a terminal manifold of codimen-524 sion one in minimum time, namely, this target is taken as a disk $N(\mathbf{x}) \leq 0$ centered on the 525 collinearity locus \mathscr{C} in which local controllability is guaranteed provided that the forcing 526 feedback is interior. This is an example of controlled stability. 527

There is no use comparing these methods in terms of computational time. The direct 528 method computes open-loop controls, while the semi-direct method computes closed-529 loop controls, which can be used for real-time application since the trajectory is computed 530 step by step. In this sense we use the direct method to have an upper bound of the value 531 function and it is compared to the objective value associated to the semi-direct method 532 presented below. 533

Both methods are implemented in MATHEMATICA programming language using the 534 optimization solver FindMinimum. 535

Direct method. We consider the time-optimal control problem: 536

(OCP)
$$\min_{u(\cdot),T} T \\ \dot{\mathbf{x}}(t) = X(\mathbf{x}(t)) + u(t) Y(\mathbf{x}(t)), \ u(t) \in [0,1], \ a.e. \ t \in [0,T] \\ \mathbf{x}(0) = x_0 \ (given) \\ N(\mathbf{x}(T)) \le 0$$

where $\mathbf{x} = (x, y)$ and X, Y are given by (23). 537

We perform a discretization over the state and the control spaces for (*OCP*) to obtain a nonlinear finite dimensional optimization problem. The optimization variables are the values of the control at each time step and a primal-dual interior point method is used to solve numerically the optimization problem. The optimality conditions – written as a relaxation of the Karush–Kuhn–Tucker conditions – are determined using automatic differentiation.

It is usually a quite robust method with respect to the initialization compared to indirect methods based on the Pontryagin Maximum Principle. However it does not exploit the geometric structure of the optimal control. The BOCOP software¹ provides an implementation of this method based on the IPOPT optimization solver².

Model predictive control method. The semi-direct method is based on a sampled-data control formulation of the problem, which is adapted to medical protocols that can be used to cure the C. difficile infection. The optimal path is constructed iteratively, where at each iteration we solve an optimal control problem on a reduced time horizon, whereas the direct method discretizes the problem on the whole time interval.

In this sense, our method is closely related to model predictive control (MPC) widely used for control theory applications. The current state of the method x_c is initialized to x_0 and is updated iteratively by solving the optimal control problems of the form

 $(OCP') \qquad \begin{array}{l} \min_{u \in \mathbb{R}^h} & N(\mathbf{x}(t_h; u, x_c)) \\ \dot{\mathbf{x}}(t) = X(\mathbf{x}(t)) + u_i \, Y(\mathbf{x}(t)), \ u_i \in [0, 1], \ a.e. \ t \in [t_i, t_{i+1}], \ i = 0, \dots, h-1 \ , \\ \mathbf{x}(0) = x_c \end{array}$

where the integer *h* is the horizon, $0 = t_0 < \cdots < t_h$ are given fixed times and $\mathbf{x}(.; u, x_c)$ is the state response associated to (u_1, \ldots, u_h) and starting at x_c at t = 0. The algorithm terminates when $|N(x_c)|$ is smaller than a given threshold.

To solve (OCP') numerically, we derive a finite dimensional optimization problem by 559 constructing an approximation of the objective function $u = (u_0, \dots, u_{h-1}) \mapsto N(\mathbf{x}(t_h))$ via 560 an approximation of $\mathbf{x}(t_h; u, x_c)$ by discretizing the differential constraint with a midpoint 561 rule (the discretization of the state on $[0, t_h]$ is finer than the partition $0 < t_1 < \cdots < t_h$). The 562 approximation of $N(\mathbf{x}(t_h; u, x_c))$ – together with its derivatives with respect to u_i , i = 1, ..., h563 - can be computed offline using symbolic computations. Then we solve the optimization 564 problem associated to (OCP') using a primal-dual interior point method. Once (OCP')565 is solved for the current value of x_c , we retrieve the values of t_1 and u_1 to update $x_c \leftarrow$ 566 $\mathbf{x}(t_1; u_1, x_c)$ and we iterate considering the resulting new instance of (*OCP'*). 567

568 4.2.3. Numerical results

We apply the previous numerical methods to the specific May–Leonard model and our aim is to control the system in the carrying simplex Π .

From the computations in Section 4.2.1, we can choose ϵ_1, ϵ_2 to fix the positions of x_{se} and x_e (corresponding to $u_e = 1$) and so that the singular dynamics goes toward \mathscr{C} and is hyperbolic. The target is taken as the circle *N* centered on \mathscr{C} and is reachable with an admissible control $u \in]0, 1[$.

We fix the initial point to $x_0 = (0.1, 0.8)$. The direct method converges in about 200 iterations to a bang–bang–singular–bang control (see Fig. 1) and the corresponding trajectory reaches *N* in less than 7.7 unit of time.

¹www.bocop.org
²www.coin-or.github.io/Ipopt

The semi-direct method is tuned with an horizon of $t_h = 2$ unit of time with h = 3 i.e. we compute three controls u_1, u_2, u_3 over this horizon. The resulting control seems to have the same structure as the control of the direct method. The trajectory reaches N in about 12 unit of time (see Fig. 1). The depicted singular behavior depends on the size of the horizon t_h and the number of controls on that horizon. A smaller horizon t_h would typically keep the trajectory away from the singular arc.

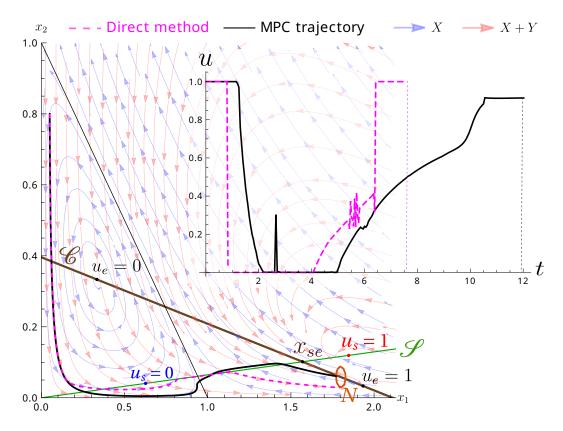


Figure 1. Geometric picture corresponding to the May–Leonard model presented in Section 4.2.3. (*left*) The dashed trajectory starting from $x_0 = (0.1, 0.8)$ and obtained with a direct method is bang–bang–singular–bang. It reaches the terminal circle *N* in less than 7.7 unit of time. The continuous MPC trajectory, obtained with an horizon of $t_h = 2$ and with three controls (h = 3), seems to reproduce the singular behavior. It reaches *N* in about 12 unit of time. (*right*) Time evolution of the control for the direct and MPC methods.

584 5. Conclusion

In this article we describe feedback invariants to classify single-input control affine systems in relation with the time-minimal control problem. They can be explicitly calculated using algebraic computations in the jets spaces of the dynamics at a given point corresponding to an abnormal stationary geodesic. They correspond at such points to the spectrum of the projectivized Hamiltonian geodesic dynamics.

This result completes the computation of the feedback invariants related to the concept of conjugate points for both normal and abnormal geodesics. This gives a neat common geometric frame since those are related to the spectrum of the "projectivized" secondorder intrinsic derivative, in relation with algebraic computations in the jets space of geodesics. ⁵⁹⁵ Our result is briefly applied to the controlled Lotka–Volterra, where abnormal stationary ⁵⁹⁶ geodesics are shifted equilibria of the free dynamics, which can be determined using lin-⁵⁹⁷ ear calculations only. This gives an algebraic frame in relation with population dynamics ⁵⁹⁸ control. Another applications are for quadratic systems in relation for instance with the ⁵⁹⁹ attitude control problem of a rigid spacecraft.

A final numerical application shows the relation of the study to the time-minimal control problem, combining Pontryagin Maximum Principle with direct and semi-direct numerical frame. Additionally it provides an example of controlled stability method popular in biological models. More generally our analysis leads to structurally stable results based on Lie algebraic computations in the jets space of the geodesics dynamics. It opens the road to fine applications translating for instance results for the free dynamics of the May-Leonard model [1] to controlled stability or time optimal results, in the controlled case.

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