

Dynamics and control for the "Guidance by repulsion" model

Dongnam Ko

DeustoTech, Universidad de Deusto

Joint work with Enrique Zuazua

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- 1 Control of collective dynamics: "guidance-by-repulsion" paradigm
- 2 Stability of the reduced limit model
- 3 Stability by means of Lyapunov functionals
- 4 Optimal control strategies for multiple evaders
- 5 Conclusions

Table of Contents

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Motivation: Shepherd dogs and sheep

The number of individuals is small, yet the interaction dynamics and control strategies is complex

We consider the "guidance by repulsion" model based on the two-agents framework: **the driver tries to drive the evader.**

The drivers want to control the evaders:

- 1 Gathering of the evaders,
- 2 Driving the evaders into a desired area.

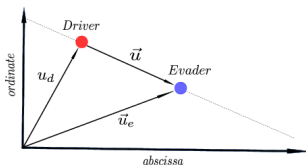
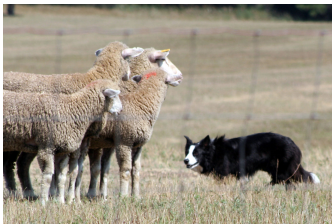


Figure: Picture of Border Collie [from Wikipedia] and the diagram of the model

Motivation: "Guidance by repulsion" model

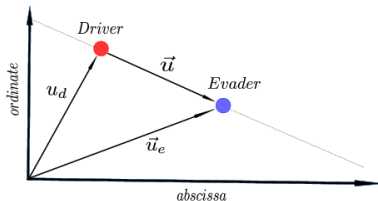
R. Escobedo, A. Ibañez and E. Zuazua, Optimal strategies for driving a mobile agent in a "guidance by repulsion" model, Communications in Nonlinear Science and Numerical Simulation, 39 (2016), 58-72.

[R. Escobedo, A. Ibañez, E. Zuazua, 2016] suggested a **guidance by repulsion** model based on the two-agents framework: *the driver*, which tries to drive the *evader*.

- 1 The driver follows the evader but **cannot be arbitrarily close** to it (because of chemical reactions, animal conflict, etc).
- 2 The **evader moves away** from the driver but doesn't try to escape beyond a not so large distance.
- 3 The driver is faster than the evader.
- 4 At a critical short distance, the driver can display a **circumvention maneuver** around the evader, forcing it to change the direction of its motion.
- 5 By adjusting the circumvention maneuver, **the evader can be driven towards a desired target or along a given trajectory**.

One sheep + one dog + Circumvention control

The control $k(t)$ is chosen in feedback form to align the gate, the sheep and the dog.



In short, the model for $\mathbf{u}_d, \mathbf{u}_e \in \mathbf{R}^2$ can be written with nonlinear interaction kernels $f_d(\cdot)$ and $f_e(\cdot)$:

$$\begin{cases} \dot{\mathbf{u}}_d = \mathbf{v}_d, & \dot{\mathbf{u}}_e = \mathbf{v}_e \\ m_d \dot{\mathbf{v}}_d = -f_d(|\mathbf{u}_d - \mathbf{u}_e|)(\mathbf{u}_d - \mathbf{u}_e) - \nu_d \mathbf{v}_d + \kappa(t)(\mathbf{u}_d - \mathbf{u}_e)^\perp \\ m_e \dot{\mathbf{v}}_e = -f_e(|\mathbf{u}_e - \mathbf{u}_d|)(\mathbf{u}_d - \mathbf{u}_e) - \nu_e \mathbf{v}_e \\ \mathbf{u}_d(0) = \mathbf{u}_d^0, \mathbf{u}_e(0) = \mathbf{u}_e^0, \mathbf{v}_d(0) = 0, \mathbf{v}_e(0) = 0 \end{cases} \quad (1)$$

Studies on the repulsive interactions

In this setting, they considered bang-bang type controls with open-loop and feed-back strategies.

Similar consideration have been addressed with repulsive interactions in control theory:

- Defender-intruder strategy : [Wang, Biegler, 2006],
- Hunting strategy model :
[Muro, Escobedo, Spector, Coppinger, 2011 and 2014],
- Dog-sheep gathering problem :
Well-posedness of optimal control problems [Burger, Pinnau, Roth, Totzeck, Tse, 2016]
and its simulations [Pinnau, Totzeck, 2018].

In particular, [Pinnau, Totzeck, 2018] uses a transport equation:

$$\begin{cases} \partial_t \mu + \nabla_x \cdot (v \mu) = \nabla_v \cdot (G(t, x, v, \mu) \mu), \\ G(t, x, v, \mu) = \int_{\mathbb{R}^2 \times \mathbb{R}^2} \nabla_x P_1(x - y) d\mu(t, y, v) \\ \quad + \frac{1}{M} \sum_{j=1}^M \nabla_x P_2(x - \mathbf{u}_{dj}(t)) - \nu v, \\ \dot{\mathbf{u}}_{dj} = \kappa_j(t) = (\kappa_j^1(t), \kappa_j^2(t)), \quad j = 1, \dots, M, \\ \mathbf{u}_{dj}(0) = \mathbf{u}_{dj}^0 \in \mathbb{R}^2, \quad \mu(0, \cdot, \cdot) = \mu_0. \end{cases}$$

Our objective is to consider **many drivers and many evaders** in the Guidance-by-repulsion model, however, **based on the interaction between one driver and one evader**.

This is a common strategy on the study of collective behavior model, the emergent dynamics should rely on the model with small number of individuals.

Guidance-by-repulsion model with many individuals

Let $\mathbf{u}_{dj}, \mathbf{u}_{ei} \in \mathbb{R}^2$ are positions of drivers and evaders for $i = 1, \dots, N$ and $j = 1, \dots, M$.

When there are many evaders, we need to suggest a **representative position of evaders** which the drivers follow. We set the **barycenter** of evaders,

$$\mathbf{u}_{ec} := \frac{1}{N} \sum_{k=1}^N \mathbf{u}_{ek},$$

then the dynamics can be described by

$$\left\{ \begin{array}{l} \ddot{\mathbf{u}}_{dj} = -f_d(|\mathbf{u}_{dj} - \mathbf{u}_{ec}|)(\mathbf{u}_{dj} - \mathbf{u}_{ec}) - \nu \dot{\mathbf{u}}_{dj} + \kappa_j(t)(\mathbf{u}_{dj} - \mathbf{u}_{ec})^\perp, \\ \ddot{\mathbf{u}}_{ei} = -\frac{1}{M} \sum_{j=1}^M f_e(|\mathbf{u}_{dj} - \mathbf{u}_{ei}|)(\mathbf{u}_{dj} - \mathbf{u}_{ei}) \\ \quad - \frac{1}{N} \sum_{k=1}^N f_g(|\mathbf{u}_{ek} - \mathbf{u}_{ei}|)(\mathbf{u}_{ek} - \mathbf{u}_{ei}) - \nu \dot{\mathbf{u}}_{ei}, \\ \mathbf{u}_{dj}(0) = \mathbf{u}_{dj}^0, \mathbf{u}_{ei}(0) = \mathbf{u}_{ei}^0, \dot{\mathbf{u}}_{dj}(0) = \mathbf{v}_{dj}^0, \dot{\mathbf{u}}_{ei}(0) = \mathbf{v}_{ei}^0. \end{array} \right.$$

A simulation

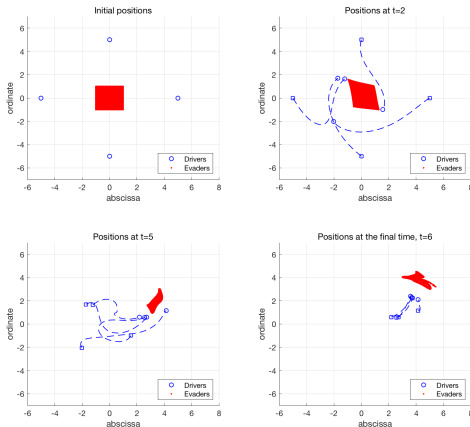


Figure: Trajectories of 4 drivers and 1024 evaders towards the point (4, 4)

Table of Contents

- 1 Control of collective dynamics: "guidance-by-repulsion" paradigm
- 2 Stability of the reduced limit model
- 3 Stability by means of Lyapunov functionals
- 4 Optimal control strategies for multiple evaders
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First order reduced model with one driver and one evader

From now on, we consider **one driver and one evader** model for analytic results.

For simplicity, we first observe the dynamics of its reduced limit, $m_e, m_d \rightarrow 0$. This singular limit removes the effect of inertia, hence, we get the long-time behavior monotonically.

$$\begin{cases} \dot{\mathbf{u}}_d = \mathbf{v}_d, & \dot{\mathbf{u}}_e = \mathbf{v}_e \\ \nu_d \dot{\mathbf{u}}_d = -f_d(|\mathbf{u}_d - \mathbf{u}_e|)(\mathbf{u}_d - \mathbf{u}_e) + \kappa(t)(\mathbf{u}_d - \mathbf{u}_e)^\perp \\ \nu_e \dot{\mathbf{u}}_e = -f_e(|\mathbf{u}_e - \mathbf{u}_d|)(\mathbf{u}_d - \mathbf{u}_e) \\ \mathbf{u}_d(0) = \mathbf{u}_d^0, \quad \mathbf{u}_e(0) = \mathbf{u}_e^0, \end{cases}$$

where the relative position $\mathbf{u} := \mathbf{u}_d - \mathbf{u}_e$ satisfies a closed equation,

$$\dot{\mathbf{u}} = -f(|\mathbf{u}|)\mathbf{u} + \kappa(t)\mathbf{u}^\perp.$$

From this relation, we can separately treat central velocity $-f(|\mathbf{u}|)\mathbf{u}$ and its perpendicular velocity $\kappa(t)\mathbf{u}^\perp$.

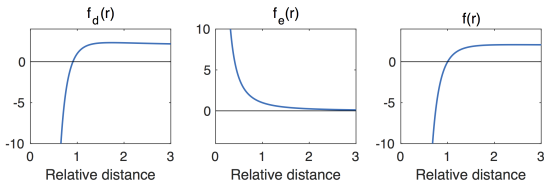
Interaction functions

Since we want the relative position \mathbf{u} to satisfy the regulation between the driver and evader, we assume $f(r) = f_d(r) - f_e(r)$ satisfy

$$f(r) = \begin{cases} \geq 0 & \text{for } r \geq r_c, \\ < 0 & \text{for } 0 < r < r_c \end{cases} \quad \text{with } f'(r_c) > 0,$$

which implies that $|\mathbf{u}|$ tends to r_c in the absence of control $\kappa(t)$.
As an example, we suggest

$$f_d(r) = \frac{2}{r^2} - \frac{3}{r^4} + 2 \quad \text{and} \quad f_e(r) = \frac{1}{r^2},$$



Potential function as a Lyapunov function

For the potential function

$$P(r) := \int_{r_c}^r sf(s)ds,$$

we may describe its gradient property:

$$\dot{\mathbf{u}} = -\nabla P(|\mathbf{u}|) + \kappa(t)\mathbf{u}^\perp,$$

The potential function plays the role of Lyapunov function.

$$\begin{aligned}\dot{P}(|\mathbf{u}|) &= \frac{dP}{d|\mathbf{u}|} \cdot \frac{d|\mathbf{u}|}{dt} = |\mathbf{u}|f(|\mathbf{u}|) \frac{\langle \mathbf{u}, \dot{\mathbf{u}} \rangle}{|\mathbf{u}|} \\ &= f(|\mathbf{u}|) \langle \mathbf{u}, -f(|\mathbf{u}|)\mathbf{u} + \kappa(t)\mathbf{u}^\perp \rangle = -f(|\mathbf{u}|)^2 |\mathbf{u}|^2 \leq 0.\end{aligned}$$

Therefore, if we assume proper conditions on $f(r)$,

$$\int_0^{r_c} rf(r) = -\infty \quad \text{and} \quad \gamma_m := \liminf_{r \rightarrow \infty} f(r) > 0,$$

so that P is smooth, coercive, and blow-up at $r = 0$. Then, from the time derivative,

$$\dot{P}(|\mathbf{u}|) = -f(|\mathbf{u}|)^2 |\mathbf{u}|^2 \leq 0.$$

we obtain dynamical properties.

Relative distance of the reduced model

- The relative distance $|\mathbf{u}|$ cannot be 0 from nonzero initial data, and **uniformly bounded along time**.
- \mathbf{u} tends to the steady solution $\bar{\mathbf{u}}(t)$ which satisfies $f(|\bar{\mathbf{u}}|)|\bar{\mathbf{u}}| = 0$, that is,

$$|\mathbf{u}| \rightarrow r_c \quad \text{if} \quad |\mathbf{u}_0| \neq 0.$$

Note that the convergence is exponential since $f'(r_c) \neq 0$ so that $P(r)$ and $\dot{P}(r)$ are both quadratic on $f(r)$ locally.

Steady states and controllability

Finally, we may classify the steady states of \mathbf{u}_d and \mathbf{u}_e .

- If $\kappa(t) \equiv 0$, then the dynamics is in a one-dimensional line including \mathbf{u}_d^0 and \mathbf{u}_e^0 . Eventually, two agents tend to **uniform linear motions**.
- If $\kappa(t) \equiv 1$, then they converge to **circular motions**, where the relative distance is r_c and angular velocities are 1.

From these two states, **we can control** \mathbf{u}_e into a desired position:

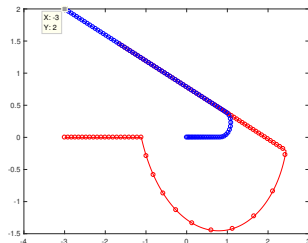
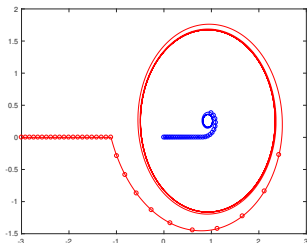


Figure: Rotational states (left) and off-bang-off control using it (right)

Table of Contents

- 1 Control of collective dynamics: "guidance-by-repulsion" paradigm
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- 3 Stability by means of Lyapunov functionals**
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The Guidance-by-repulsion model

Next, we go back to **the second order** Guidance-by-repulsion model.

$$\ddot{\mathbf{u}} + f(|\mathbf{u}|)\mathbf{u} + \nu\dot{\mathbf{u}} = \kappa(t)\mathbf{u}^\perp.$$

For the interaction coefficient $f(r)$, we assume the same condition: for

$$P(r) := \int_{r_c}^r sf(s)ds \geq 0,$$

$P(0) = \infty$ and P grows quadratically $\left(\sim \frac{\gamma_m}{2}|\mathbf{u}|^2\right)$ as $r \rightarrow \infty$.

- The equation now follows the motion of **damped oscillator under a central potential** $P(|\mathbf{u}|)$ with an additional control term.
- The negativity/positivity of f makes the relative distance $\mathbf{u} \sim r_c$.

Two main regimes arise: Pursuit $\kappa(t) = 0$ / Circumvention $\kappa(t) \neq 0$.

Steady states

For each mode, we have the following steady states which characterize the dynamics:

- **Pursuit mode:** $\kappa(t) \equiv 0$:

$$\mathbf{u}(t) = \mathbf{u}_* \in \mathbb{R}^2 \quad \text{and} \quad \mathbf{v}(t) = (0, 0) \quad \text{with} \quad |\mathbf{u}_*| = r_c,$$

where the driver and evader behave uniform **linear motions**,

$$\mathbf{u}_\ell(t) = -\frac{f_d(\mathbf{u}_*)\mathbf{u}_*}{\nu}t + \mathbf{u}_\ell(0), \quad \ell = d, e.$$

- **Circumvention mode,** $\kappa(t) \equiv \kappa$:

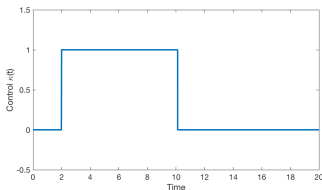
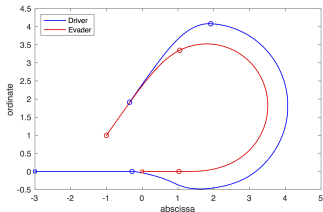
$$\mathbf{u}(t) = r_p \left(\cos\left(\frac{\kappa}{\nu}t\right), \sin\left(\frac{\kappa}{\nu}t\right) \right),$$

where the driver and evader have **rotational motions** on circles centered at the same point,

$$\mathbf{u}_\ell(t) = r_\ell \left(\cos\left(\frac{\kappa}{\nu}t + \phi_\ell\right), \sin\left(\frac{\kappa}{\nu}t + \phi_\ell\right) \right) + \mathbf{u}^*, \quad \mathbf{u}^* \in \mathbb{R}^2, \quad \ell = d, e.$$

Off-Bang-Off control of the evader

Combining these two modes, we can construct an Off-Bang-Off control: choose the direction by rotations in the circumvention mode, and drive the evaders to the target in the pursuit mode.



Theorem [K.-Zuazua (preprint)]

Let $f(r)$ be as before. Then, for a given destination $\mathbf{u}_f \in \mathbb{R}^2$ and $\mathbf{u}_0 \neq (0, 0)$, there exist t_1 , t_2 , t_f and κ such that the control function

$$\kappa(t) = \begin{cases} \kappa & \text{if } t \in [t_1, t_2], \\ 0 & \text{if } t \in [0, t_1) \cup (t_2, t_f], \end{cases} \quad \text{satisfies } \mathbf{u}_e(t_f) = \mathbf{u}_f.$$

Stability to the steady states

In order to analyze the off-bang-off control, we need to show the **asymptotic stability to the steady states** on each constant $\kappa(t)$.

The equation of the relative position \mathbf{u} with constant control $\kappa(t) \equiv \kappa$,

$$\ddot{\mathbf{u}} + f(|\mathbf{u}|)\mathbf{u} + \nu\dot{\mathbf{u}} = \kappa\mathbf{u}^\perp, \quad \mathbf{u} \in \mathbf{R}^2,$$

which is the **damped potential oscillator** with an external source term.

However, the standard energy,

$$E(t) := \frac{1}{2}|\mathbf{v}|^2 + P(|\mathbf{u}|),$$

is **no more non-increasing** from the perpendicular term $\kappa(t)\mathbf{u}^\perp$.

$$\begin{aligned} \dot{E}(t) &= \mathbf{v} \cdot \dot{\mathbf{v}} + f(|\mathbf{u}|)\mathbf{u} \cdot \dot{\mathbf{u}} \\ &= \mathbf{v} \cdot (-f(|\mathbf{u}|)\mathbf{u} - \nu\mathbf{v} + \kappa(t)\mathbf{u}^\perp) + f(|\mathbf{u}|)\mathbf{u} \cdot \mathbf{v} \\ &= -\nu|\mathbf{v}|^2 + \kappa(t)\mathbf{u}^\perp \cdot \mathbf{v}. \end{aligned}$$

To fix it, we use **hypocoercivity theory**¹, and construct a perturbed energy using inner product terms:

$$L_{\pm}(t) = E(t) \pm \frac{\nu}{2} \left(\frac{\nu}{2} |\mathbf{u}|^2 + \mathbf{u} \cdot \mathbf{v} \right).$$

Then, its time derivative is

$$\dot{L}_{\pm}(t) \leq -\frac{\nu}{2} |\mathbf{v}|^2 + \frac{1}{2} (\nu f(|\mathbf{u}|) + \kappa(t)) |\mathbf{u}|^2,$$

which is nonpositive if $|\mathbf{u}|$ is close to 0 or ∞ .

On the other hand, if $\kappa(t)$ is constant, we may define κ dependent functions,

$$L_{\kappa}(t) = E(t) - \frac{\kappa}{\nu} \mathbf{u}^{\perp} \cdot \mathbf{v} \quad \text{and} \quad \dot{L}_{\kappa}(t) = -\nu \left| \mathbf{v} - \frac{\kappa}{\nu} \mathbf{u}^{\perp} \right|^2 \leq 0,$$

which is always nonpositive.

¹[C. Villani, 2009] and [K. Beauchard, E. Zuazua, 2011]

Therefore, we have the following dynamical properties.

Boundedness of relative distance

Suppose that the control $\kappa(t)$ is bounded: $\limsup_{t \rightarrow \infty} |\kappa(t)| < \nu\sqrt{\gamma_m}$.

Then, the relative position $\mathbf{u}(t)$ **does not hit $(0,0)$ or blow-up** in a finite time. Moreover, if $\kappa(t)$ is constant, then $\mathbf{u}(t)$ **is uniformly bounded**.

Global stability of steady states

The positions $\mathbf{u}_d(t)$ and $\mathbf{u}_e(t)$ converge to the steady states asymptotically if $\kappa(t) \equiv \kappa$ and $\kappa < \nu\sqrt{\gamma_m}$:

- If $\kappa = 0$, then $\mathbf{u}_d(t)$ and $\mathbf{u}_e(t)$ tend to **linear motions**.
- If $0 < |\kappa| < \nu\sqrt{\gamma_m}$, then $\mathbf{u}_d(t)$ and $\mathbf{u}_e(t)$ tend to **rotational motions**.

By combining these asymptotic steady states, we may prove the **controllability of the evader's position** to any desired point.

Since we can apply the Off-Bang-Off controls to any initial data, we may use it to **pass multiple target points**:

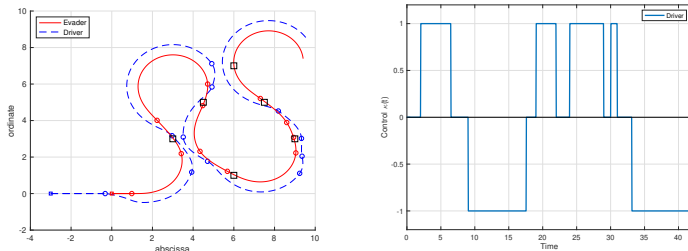


Figure: A trajectory of the evader which passes near points (3, 3), (4.5, 5), (6, 1), (9, 3), (7.5, 5) and (6, 7) denoted by black boxes.

This can be done by **turning on and off $\kappa(t)$** using two control modes, where the dynamics converges to the corresponding steady state ('rotational motion' and 'linear motion') in a short time.

The effect of the number of evaders

If the evaders are gathered initially, the dynamics are similar to the one evader case, as we have **one fat evader**.

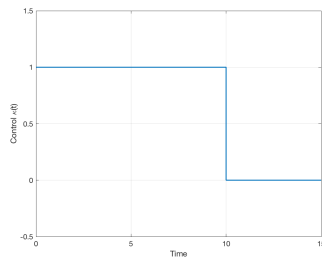
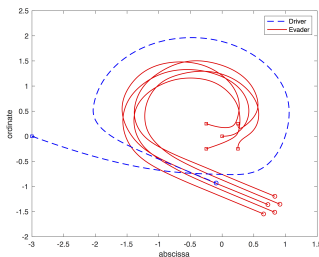


Figure: Trajectories of five evaders with a bang-off control $\kappa(t)$.

$$f_d(r) = \frac{2}{r^2} - \frac{3}{r^4} + 2, \quad f_e(r) = \frac{1}{r^2}, \quad \text{and} \quad f_g(r) = 10 \left(\frac{(0.2)^2}{r^2} - \frac{(0.2)^4}{r^4} \right).$$

Table of Contents

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- 2 Stability of the reduced limit model
- 3 Stability by means of Lyapunov functionals
- 4 Optimal control strategies for multiple evaders
- 5 Conclusions

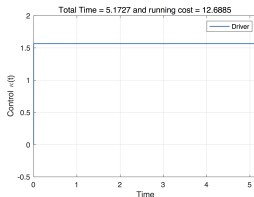
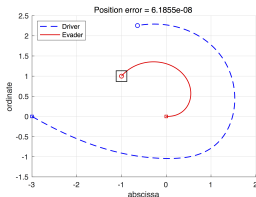
Optimal control strategies

While Off-Bang-Off controls can drive the evaders properly, it is natural to find an optimal control strategy which minimizes a given cost.

For the cost function, we suggest two optimal control problems: On one hand, we want to **minimize the running cost** of controls:

$$J(\kappa(\cdot)) = \frac{1}{N} \sum_{i=1}^N |\mathbf{u}_{ei}(t_f) - \mathbf{u}_f|^2 + \frac{0.001}{M} \sum_{k=1}^M \int_0^{t_f} |\kappa_k(t)|^2 dt.$$

The simulations are done by gradient descent methods with flexible final time t_f , where the initial guess is given by constant control functions. For example, $\kappa(t) \equiv 1.5662$ and $t_f = 5.1727$ to make $\mathbf{u}(t_f) = (-1, 1)$:



One driver and one evader

We can observe that the optimal strategy is not an Off-Bang-Off control, but it shares the main idea: 'rotate and then drive'.

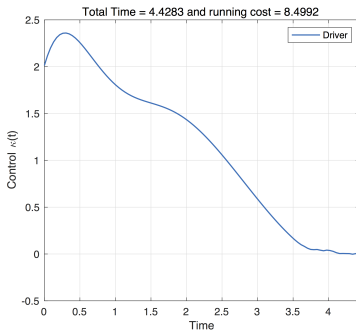
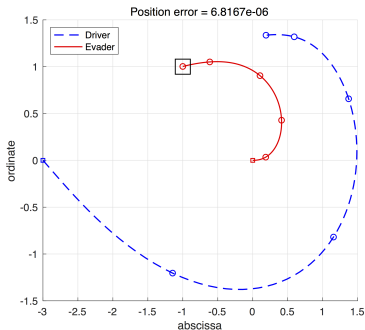


Figure: Diagrams for the optimal control leading to $\mathbf{u}_e(t_f) \simeq (-1, 1)$ when initially $\mathbf{u}_e^0 = (0, 0)$, $\mathbf{u}_d^0 = (-3, 0)$ and zero velocities.

Two drivers and one evader

This 'rotate and then drive' strategy also works with two drivers. In a similar initial data from the previous simulation, we can observe that **two drivers act like one driver**.

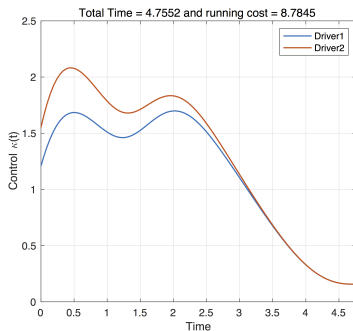
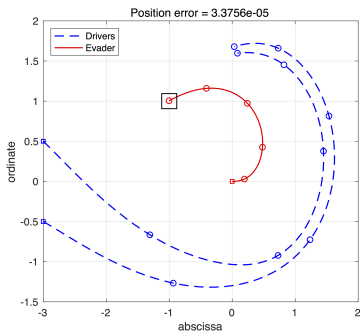


Figure: Diagrams for the control leading to $\mathbf{u}_{e1}(t_f) \simeq (-1, 1)$ when initially $\mathbf{u}_e^0 = (0, 0)$, $\mathbf{u}_{d1}^0 = (-3, 0.5)$, $\mathbf{u}_{d2}^0 = (-3, -0.5)$ and zero velocities.

Two drivers and one evader: Minimizing the driving time

It is not changed much even if we want to minimize the driving time,

$$J(\kappa(\cdot)) = \frac{1}{N} \sum_{i=1}^N |\mathbf{u}_{ei}(t_f) - \mathbf{u}_f|^2 + \frac{0.001}{M} \sum_{k=1}^M \int_0^{t_f} |\kappa_k(t)|^2 dt + 0.1 t_f.$$

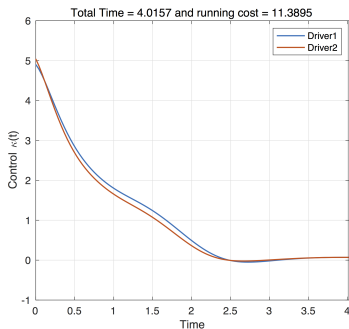
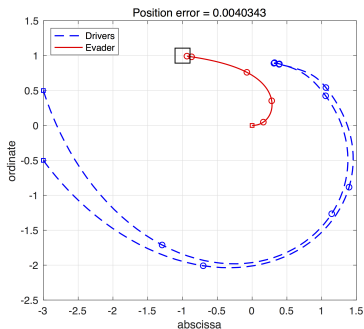


Figure: Diagrams for the control leading to $\mathbf{u}_{e1}(t_f) \simeq (-1, 1)$ when initially $\mathbf{u}_e^0 = (0, 0)$, $\mathbf{u}_{d1}^0 = (-3, 0.5)$, $\mathbf{u}_{d2}^0 = (-3, -0.5)$ and zero velocities.

The trajectories can be significantly different **if initial positions are not well-ordered**, in terms of the initial velocity of the evader. However, for any case, **the drivers want the evader to get the right direction in a short time**.

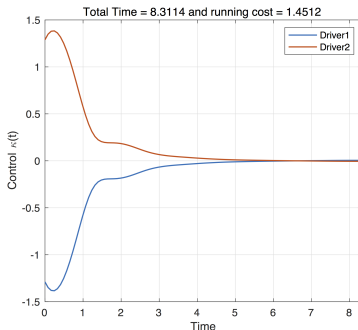
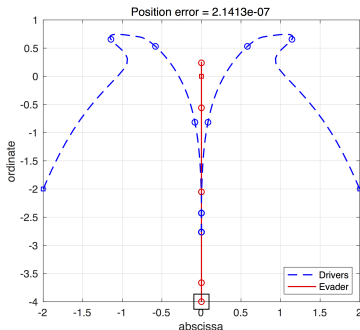


Figure: Diagrams for the control leading to $\mathbf{u}_{e1}(t_f) \simeq (0, -4)$ when initially $\mathbf{u}_e^0 = (0, 0)$, $\mathbf{u}_{d1}^0 = (-2, -2)$, $\mathbf{u}_{d2}^0 = (-2, 2)$ and zero velocities.

In the same way, the minimal time optimal strategy contains strong control functions and wants to decrease the relative position.

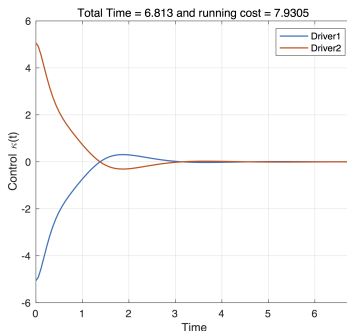
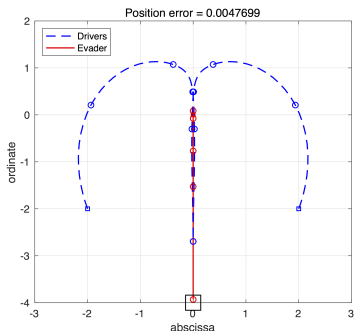


Figure: Diagrams for the control leading to $\mathbf{u}_{e1}(t_f) \simeq (0, -4)$ when initially $\mathbf{u}_e^0 = (0, 0)$, $\mathbf{u}_{d1}^0 = (-2, -2)$, $\mathbf{u}_{d2}^0 = (-2, 2)$ and zero velocities.

Table of Contents

- 1 Control of collective dynamics: "guidance-by-repulsion" paradigm
- 2 Stability of the reduced limit model
- 3 Stability by means of Lyapunov functionals
- 4 Optimal control strategies for multiple evaders
- 5 Conclusions**

Conclusions

■ Results

- The guidance-by-repulsion is difficult since it is a **bi-linear control** problem which deals with **partial controllability**. (Null-controllability is trivially false)
- In summary, **one-driver and one-evader model** with special assumptions (same frictions, constant control, good potential) leads to the **controllability** of the evader's position.

■ Future objectives

- It is natural to extend this model to **measure-valued** distributions (transport equation) or **stochastic** particles (Fokker-Planck equation) by adding diffusive effects.
- Numerical calculation costs too much in gradient methods since it has **a lot of possible controls which may have locally optimal costs**.

THANK YOU FOR YOUR INVITATION AND ATTENTION!