

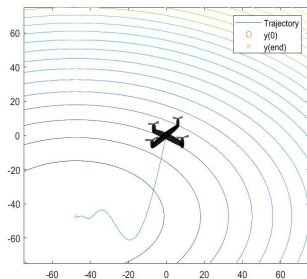
Performance Analysis of Source-Seeking Algorithms with Integral Quadratic Constraints

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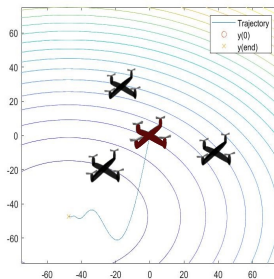
PhD Workshop, 2021
Technical University of Hamburg

28th Feb, 2022

Source-seeking Problem



PhD workshop Sept 2021



PhD workshop Feb 2022

Assumptions (Informal)

- ▶ Field is differentiable and convex
- ▶ Local gradients available at **some** leader agents
- ▶ Connectivity assumptions on interconnection graphs

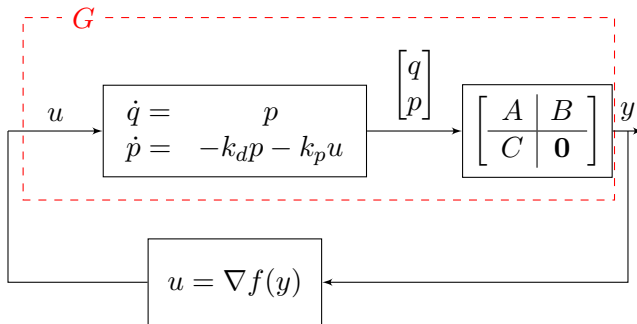
Outline

Review: Single agent case

Extension: Multiple agents

Numerical results

Control Architecture: Single Agent



$$\dot{\eta}(t) = A_G \eta(t) + B_G u(t), \quad \eta(0) = \eta_0$$

$$y(t) = C_G \eta(t)$$

$$u(t) = \nabla f(y(t)),$$

Loop in the deviation variables: Single agent

Equilibrium

$$\begin{aligned}0 &= A_G \eta_* + B_G u_* = A_G \eta_* \\ y &= C_G \eta_* \\ u_* &= \nabla f(y_*) = 0\end{aligned}\tag{1}$$

Loop in the transformed variables

$$\begin{aligned}\dot{\tilde{\eta}}(t) &= A_G \tilde{\eta}(t) + B_G \tilde{u}(t), & \tilde{\eta}(0) &= \eta_0 - \eta_* \\ \tilde{y}(t) &= C_G \tilde{\eta}(t)\end{aligned}\tag{2}$$

and

$$\tilde{u}(t) = \nabla f(\tilde{y}(t) + y_*)\tag{3}$$

Main Results: Single Agent

- ▶ Let y_* minimizes $f \in \mathcal{S}(m, L)$ ¹

¹ $m||y_1 - y_2||^2 \leq (\nabla f(y_1) - \nabla f(y_2))^T (y_1 - y_2) \leq L||y_1 - y_2||^2$

Main Results: Single Agent

► Let y_* minimizes $f \in \mathcal{S}(m, L)$ ¹

►
$$\left[\begin{array}{c|c} \mathcal{A} & \mathcal{B} \\ \hline \mathcal{C} & \mathcal{D} \end{array} \right] = \Pi \begin{bmatrix} G \\ I_d \end{bmatrix}$$

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►
$$\mathbb{P} = \left\{ \begin{bmatrix} \mathbf{0} \\ * \end{bmatrix} \begin{bmatrix} H & -P_3 \\ -P_1^T & \mathbf{0} \end{bmatrix} : \text{LM}(H, P_1, P_3) < 0 \right\}$$

¹ $m||y_1 - y_2||^2 \leq (\nabla f(y_1) - \nabla f(y_2))^T (y_1 - y_2) \leq L||y_1 - y_2||^2$

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Theorem 1

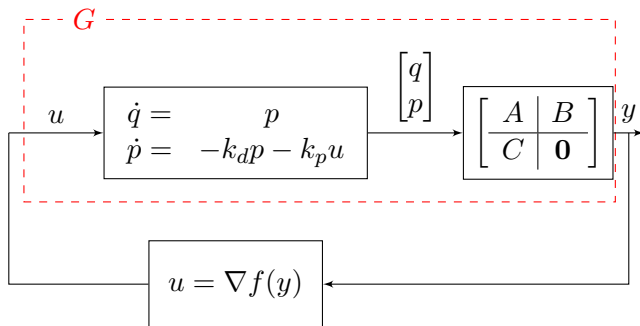
If $\exists \mathcal{X} > 0, P \in \mathbb{P}$ such that

$$\begin{bmatrix} \mathcal{A}^T \mathcal{X} + \mathcal{X} \mathcal{A} + 2\alpha \mathcal{X} & \mathcal{X} \mathcal{B} \\ \mathcal{B}^T \mathcal{X} & \mathbf{0} \end{bmatrix} + \begin{bmatrix} \mathcal{C}^T \\ \mathcal{D}^T \end{bmatrix} (P \otimes I_d) \begin{bmatrix} \mathcal{C} & \mathcal{D} \end{bmatrix} \leq 0,$$

then, $\|y(t) - y_*(t)\| \leq \kappa e^{-\alpha t}$ holds for all $t \geq 0$

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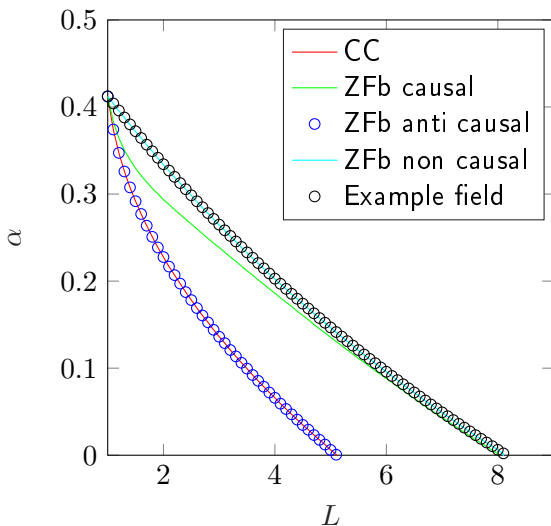
Numerical results for a single quadrotor



Setup

- ▶ Linearized quadrotor model + LQR local controller
- ▶ Questions:
 - ▶ How robust is the given controller for different fields?
 - ▶ How do we select the PD gains for the pre-filter?
 - ▶ How conservative are the estimates on the rates of convergence?

Robustness w.r.t scalar field for a single quadrotor



Control Architecture: Multiple agents

$$\begin{aligned}\dot{\hat{\eta}}(t) &= \hat{A}_G \hat{\eta}(t) + \hat{B}_G \hat{u}(t), & \hat{\eta}(0) &= \hat{\eta}_0, \\ \hat{y}(t) &= \hat{C}_G \hat{\eta}(t).\end{aligned}$$

where, notation $\hat{X} = I_N \otimes X$ and $\hat{x} = [x_1^T, \dots, x_N^T]^T$ is used.

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Formation control law with gradient-based forcing term

$$\hat{u} = \mathcal{L}_{(d)}(\hat{y} - \hat{r}) + \begin{bmatrix} u_{\psi_1} \\ \vdots \\ u_{\psi_N} \end{bmatrix}. \quad (4)$$

$$u_{\psi_i}(t) = \begin{cases} \nabla \psi(y_i), & \text{if } i \in \mathcal{V}_l, \\ 0, & \text{otherwise.} \end{cases} \quad (5)$$

Control Architecture: Multiple agents

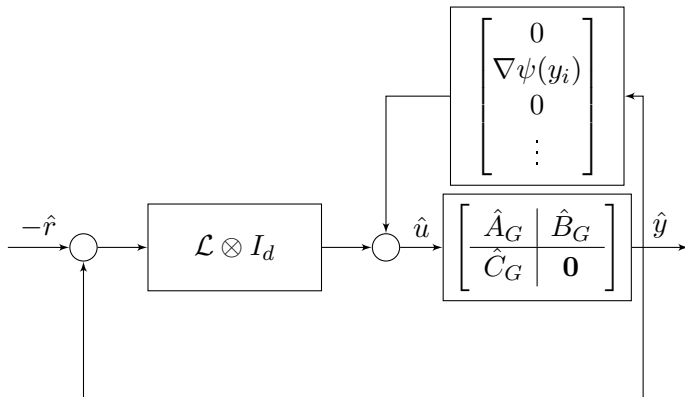


Figure 2: Control architecture

Formulation as a robust control problem

Definition

For a given graph \mathcal{G} of order N (with its corresponding laplacian \mathcal{L}), the leader set \mathcal{V}_l , a scalar field ψ and a given formation reference vector $\hat{r} \in \mathbb{R}^{Nd}$, define a function $f : \mathbb{R}^{Nd} \rightarrow \mathbb{R}$ by

$$f(\hat{y}) = \frac{1}{2}(\hat{y} - \hat{r})^T \mathcal{L}_{(d)}(\hat{y} - \hat{r}) + \sum_{v_i \in \mathcal{V}_l} \psi(y_i). \quad (6)$$

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Overall close-loop system

$$\begin{aligned} \dot{\hat{\eta}}(t) &= \hat{A}_G \hat{\eta}(t) + \hat{B}_G \hat{u}(t), & \hat{\eta}(0) &= \hat{\eta}_0, \\ \hat{y}(t) &= \hat{C}_G \hat{\eta}(t), \\ \hat{u}(t) &= \nabla f(\hat{y}(t)). \end{aligned} \quad (7)$$

Recall Theorem 1

$$\left[\begin{array}{c|c} \mathcal{A} & \mathcal{B} \\ \hline \mathcal{C} & \mathcal{D} \end{array} \right] = (\Pi \otimes I_N) \left[\begin{array}{c} \hat{G} \\ \hline I_{Nd} \end{array} \right] = \left[\begin{array}{c|c} A_\Pi \otimes I_N & B_\Pi \otimes I_N \\ \hline C_\Pi \otimes I_N & D_\Pi \otimes I_N \end{array} \right] \left[\begin{array}{c|c} \hat{A}_G & \hat{B}_G \\ \hline \hat{C}_G & \mathbf{0} \\ \hline \mathbf{0} & I_{Nd} \end{array} \right]$$

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Theorem 1 applied to the overall networked system

If $f \in \mathcal{S}(m, L)$, y_* minimizes f and $\exists \mathcal{X} > 0, P \in \mathbb{P}$ such that

$$\begin{bmatrix} \mathcal{A}^T \mathcal{X} + \mathcal{X} \mathcal{A} + 2\alpha \mathcal{X} & \mathcal{X} \mathcal{B} \\ \mathcal{B}^T \mathcal{X} & \mathbf{0} \end{bmatrix} + \begin{bmatrix} \mathcal{C}^T \\ \mathcal{D}^T \end{bmatrix} (P \otimes I_{Nd}) \begin{bmatrix} \mathcal{C} & \mathcal{D} \end{bmatrix} \leq 0,$$

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Questions

- 1 How to characterize minimizers of f ?
- 2 How to verify if $f \in \mathcal{S}(m, L)$?
- 3 Can we exploit this structure in the LMIs?

1. Minimizers of f

Recall definition of f

$$f(\hat{y}) = \frac{1}{2}(\hat{y} - \hat{r})^T \mathcal{L}_{(d)}(\hat{y} - \hat{r}) + \sum_{v_i \in \mathcal{V}_l} \psi(y_i).$$

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Assumptions

- 1 Let $\psi \in \mathcal{S}(m_\psi, L_\psi)$ for some $0 < m_\psi \leq L_\psi$ and let y_* minimize ψ .
- 2 For every node $v_i \in \mathcal{V}$, there is a node $v_j \in \mathcal{V}_l$ such that \mathcal{G} contains an $i - j$ path.

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Separately consider the cases

- 1 Consensus $\hat{r} = \mathbf{0}$, $|\mathcal{V}_l| \geq 1$
- 2 Formation control with single leader $\hat{r} \neq \mathbf{0}$, $|\mathcal{V}_l| = 1$
- 3 Formation control with multiple leaders $\hat{r} \neq \mathbf{0}$, $|\mathcal{V}_l| > 1$

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Lemma (Consensus)

Let Assumptions 1 and 2 hold and let $\hat{r} = 0$. Then, \hat{y} is the minimizer of f iff $\hat{y} = \mathbf{1}_N \otimes y_*$.

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Lemma (Formation with one leader)

Let Assumptions 1 and 2 hold and let $|\mathcal{V}_l| = 1$. Then, \hat{y} is the minimizer of f iff $\hat{y}_i = y_*$ for $i \in \mathcal{V}_l$ and $y_j = y_* + (r_j - r_i)$ for all $j \in \mathcal{V}$.

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Conjecture (Formation with multiple leaders)

Let Assumptions 1 and 2 hold. The unique minimizer y_* of ψ lies in the convex hull of positions of leader agents.

2. Smoothness and convexity of f

Definition of grounded Laplacians

Define grounded Laplacians

$$\begin{aligned}\mathcal{L}_s &= \mathcal{L} + m_\psi E, \\ \mathcal{L}_b &= \mathcal{L} + L_\psi E,\end{aligned}\tag{8}$$

where, E is a diagonal matrix with the i^{th} diagonal entry equal to 1 if $i \in \mathcal{V}_l$ and equal to 0 otherwise.

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Lemma

For constants $0 < m \leq L$, the following two statements are equivalent:

1. $f \in \mathcal{S}(m, L)$ for all $\psi \in \mathcal{S}(m_\psi, L_\psi)$,
2. $mI_N \preceq \mathcal{L}_b \preceq \mathcal{L}_s \preceq LI_N$

3. Decomposition

Define

$$\left[\begin{array}{c|c} \mathcal{A} & \mathcal{B} \\ \hline \mathcal{C} & \mathcal{D} \end{array} \right] = (\Pi \otimes I_N) \left[\begin{array}{c} \hat{G} \\ I_{Nd} \end{array} \right] \text{ and } \left[\begin{array}{c|c} \mathcal{A}_0 & \mathcal{B}_0 \\ \hline \mathcal{C}_0 & \mathcal{D}_0 \end{array} \right] = \Pi \left[\begin{array}{c} G \\ I_d \end{array} \right]$$

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$$\left[\begin{array}{cc} \mathcal{A}^T \mathcal{X} + \mathcal{X} \mathcal{A} + 2\alpha \mathcal{X} & \mathcal{X} \mathcal{B} \\ \mathcal{B}^T \mathcal{X} & \mathbf{0} \end{array} \right] + (*) (P \otimes I_{Nd}) \begin{bmatrix} \mathcal{C} & \mathcal{D} \end{bmatrix} \leq 0, \quad (9)$$

$$\left[\begin{array}{cc} \mathcal{A}_0^T \mathcal{X}_0 + \mathcal{X}_0 \mathcal{A}_0 + 2\alpha \mathcal{X}_0 & \mathcal{X}_0 \mathcal{B}_0 \\ \mathcal{B}_0^T \mathcal{X}_0 & \mathbf{0} \end{array} \right] + (*) (P \otimes I_d) \begin{bmatrix} \mathcal{C}_0 & \mathcal{D}_0 \end{bmatrix} \leq 0, \quad (10)$$

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Theorem

The following statements are equivalent:

1. $\exists \mathcal{X} > 0, P \in \mathbb{P}$ such that (9) is satisfied.
2. $\exists \mathcal{X}_0 > 0, P \in \mathbb{P}$ such that (10) is satisfied.

Final Analysis result

Definition

$$\begin{aligned}\Delta_{m,L} &= \{(\mathcal{G}, \mathcal{V}_l, \psi) : \psi \in \mathcal{S}(m_\psi, L_\psi), f \in \mathcal{S}(m, L)\} \\ &= \{(\mathcal{G}, \mathcal{V}_l, \psi) : \psi \in \mathcal{S}(m_\psi, L_\psi), mI_N \preceq \mathcal{L}_b \preceq \mathcal{L}_s \preceq LI_N\} \\ &\hspace{15em} (11)\end{aligned}$$

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Theorem

Let $(\mathcal{G}, \mathcal{V}_l, \psi) \in \Delta_{m,L}$ for some $0 < m \leq L$. If $\exists \mathcal{X}_0 > 0, P \in \mathbb{P}$ such that

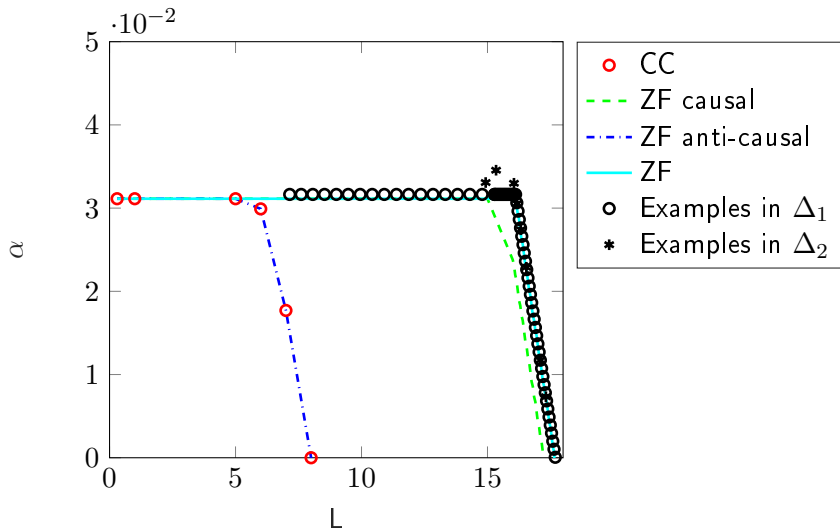
$$\begin{bmatrix} \mathcal{A}_0^T \mathcal{X}_0 + \mathcal{X}_0 \mathcal{A}_0 + 2\alpha \mathcal{X}_0 & \mathcal{X}_0 \mathcal{B}_0 \\ \mathcal{B}_0^T \mathcal{X}_0 & \mathbf{0} \end{bmatrix} + (*) (P \otimes I_d) \begin{bmatrix} \mathcal{C}_0 & \mathcal{D}_0 \end{bmatrix} \leq 0,\tag{12}$$

is satisfied, then, $\|y(t) - y_*(t)\| \leq \kappa e^{-\alpha t}$ holds for all $t \geq 0$.

Example 1: Conservatism

$$\Delta_1 = \{(\mathcal{G}, \mathcal{V}_l, \psi) : \mathcal{G} = \mathcal{G}_{\text{star}}^5, \mathcal{V}_l = \{1\}, \psi \in \mathcal{S}(2.02, L_\psi)\} \subset \Delta_{0.3, L}$$

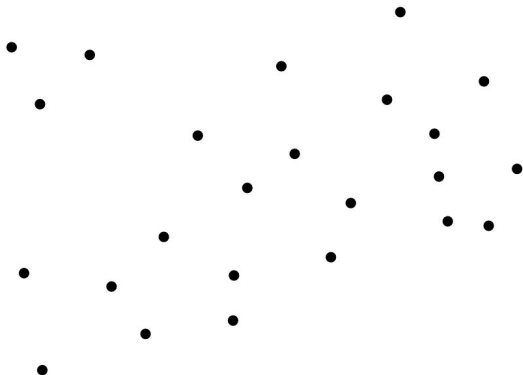
$$\Delta_2 = \{(\mathcal{G}^{25}, \mathcal{V}_l, \psi) : \psi = 1.85||y - y_*||^2\} \subset \Delta_{m, L}$$



Example 2: Spectrum of Laplacian unknown

Assume:

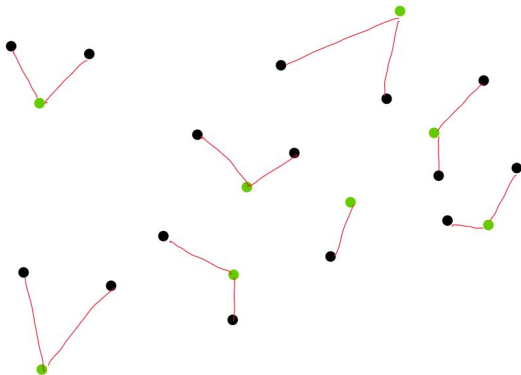
- i At least one third of total number of agents are leaders
- ii Maximum degree of all agents is 2
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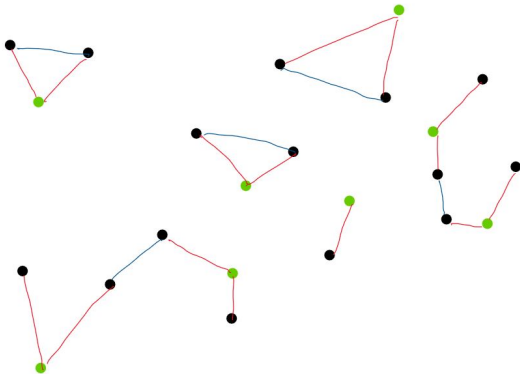
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Example 2: Spectrum of Laplacian unknown

Assume:

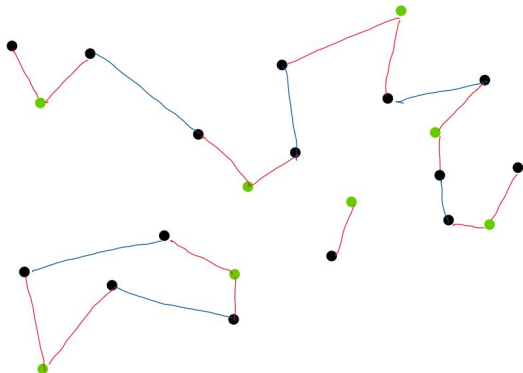
- i At least one third of total number of agents are leaders
- ii Maximum degree of all agents is 2
- iii Every agent has an edge with at least one leader



Example 2: Spectrum of Laplacian unknown

Assume:

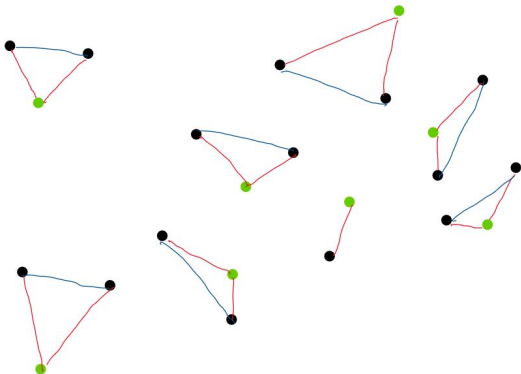
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Example 2: Spectrum of Laplacian unknown

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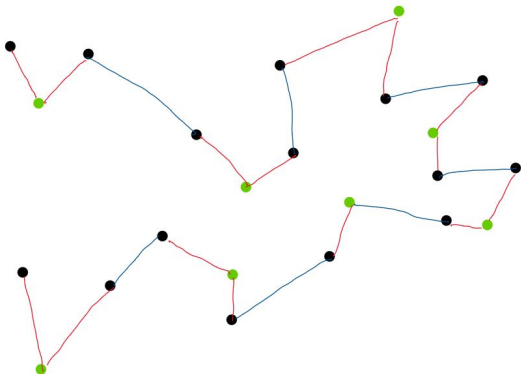
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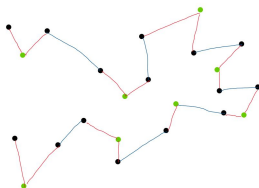
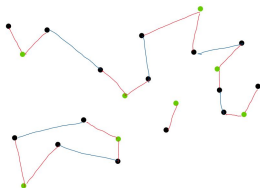
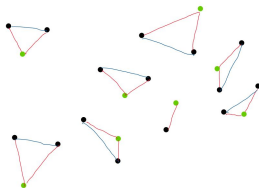
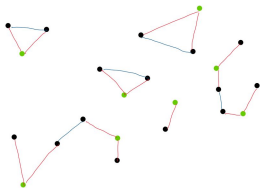
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Example 2: Spectrum of Laplacian unknown

Setup

- ▶ Linearized quadrotor model + LQR tracking local controller

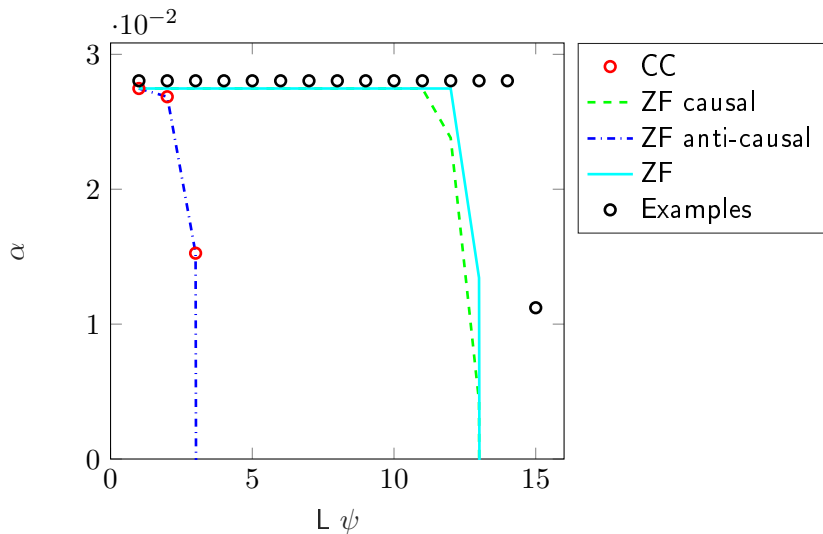
1 Conditions on the graph

- i At least one third of total number of agents are leaders, i.e., $|\mathcal{V}_l| \geq |\mathcal{V}|/3$.
- ii Maximum degree of all agents is 3
- iii Every agent has an edge with at least one leader, i.e., For any $i \in \mathcal{V}$, there is a $j \in \mathcal{V}_l$ such that $(i, j) \in \mathcal{E}$

2 Let $\psi \in \mathcal{S}(1, L_\psi)$

- ▶ Let $\bar{\Delta} = \{(\mathcal{G}, \mathcal{V}_l, \psi) : (1), (2)\}$
- ▶ Can show: $\bar{\Delta} \subset \Delta_{m,L}$ with $m = 0.3$ and $L = L_\psi + 2 * 3$

Example 2: Spectrum of Laplacian unknown



Thank you