Finite Horizon Density Steering for Multi-input State Feedback Linearizable Systems

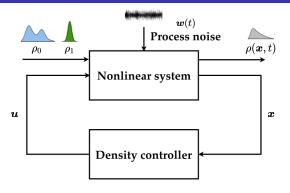
Kenneth Caluya

Joint work with Abhishek Halder

University of California, Santa Cruz

2020 American Control Conference Denver, July 3, 2020

Problem: Finite Horizon Feedback Density Control



$$\inf_{\boldsymbol{u} \in \mathcal{U}} \quad \mathbb{E} \left\{ \int_0^1 \mathcal{L}(\boldsymbol{x}, \boldsymbol{u}, t) \, \mathrm{d}t \right\}$$
 subject to
$$d\boldsymbol{x} = \boldsymbol{f}(\boldsymbol{x}, t) \, \mathrm{d}t + \boldsymbol{G}(\boldsymbol{x}, t) \boldsymbol{u}(\boldsymbol{x}, t) \, \mathrm{d}t + \sqrt{2\epsilon} \boldsymbol{C}(\boldsymbol{x}, t) \, \mathrm{d}\boldsymbol{w}$$
$$\boldsymbol{x}(t=0) \sim \rho_0(\boldsymbol{x}), \quad \boldsymbol{x}(t=1) \sim \rho_1(\boldsymbol{x})$$

Density Control: Schrödinger Bridge Problem

- $\mathcal{L}(\mathbf{x}, \mathbf{u}, t) \equiv \|\mathbf{u}(\mathbf{x}, t)\|_{2}^{2}, \mathbf{f} \equiv \mathbf{0}, \mathbf{G} \equiv \mathbf{I}, \mathbf{C} \equiv \mathbf{I},$
- Compute $ho^{\mathrm{opt}}({m x},t), {m u}^{\mathrm{opt}}({m x},t) =
 abla \psi({m x},t)$ from

$$\begin{split} &\frac{\partial \rho^{\mathrm{opt}}}{\partial t} + \nabla \cdot \left(\rho^{\mathrm{opt}} \nabla \psi \right) = \epsilon \Delta \rho^{\mathrm{opt}} \\ &\frac{\partial \psi}{\partial t} + \frac{1}{2} \| \nabla \psi \|_2^2 = -\epsilon \Delta \psi \\ &\rho^{\mathrm{opt}}(\mathbf{x}, \mathbf{0}) = \rho_0(\mathbf{x}), \quad \rho^{\mathrm{opt}}(\mathbf{x}, \mathbf{1}) = \rho_1(\mathbf{x}) \end{split}$$

Density Control: Schrödinger Bridge Problem

- $\mathcal{L}(\mathbf{x}, \mathbf{u}, t) \equiv \|\mathbf{u}(\mathbf{x}, t)\|_{2}^{2}, \mathbf{f} \equiv \mathbf{0}, \mathbf{G} \equiv \mathbf{I}, \mathbf{C} \equiv \mathbf{I},$
- ullet Compute $ho^{\mathrm{opt}}({m x},t), {m u}^{\mathrm{opt}}({m x},t) =
 abla \psi({m x},t)$ from

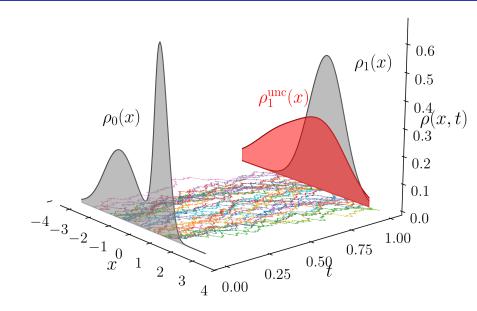
$$\begin{split} &\frac{\partial \rho^{\mathrm{opt}}}{\partial t} + \nabla \cdot \left(\rho^{\mathrm{opt}} \nabla \psi \right) = \epsilon \Delta \rho^{\mathrm{opt}} \\ &\frac{\partial \psi}{\partial t} + \frac{1}{2} \| \nabla \psi \|_2^2 = -\epsilon \Delta \psi \\ &\rho^{\mathrm{opt}}(\mathbf{x}, \mathbf{0}) = \rho_0(\mathbf{x}), \quad \rho^{\mathrm{opt}}(\mathbf{x}, \mathbf{1}) = \rho_1(\mathbf{x}) \end{split}$$

• Schrödinger System: $(\rho^{\mathrm{opt}}, \psi) \mapsto (\varphi, \hat{\varphi})$

$$egin{aligned} rac{\partial arphi}{\partial t} &= -\epsilon \Delta arphi \ rac{\partial \hat{arphi}}{\partial t} &= \epsilon \Delta \hat{arphi} \ arphi_0(oldsymbol{x}) \hat{arphi}_0(oldsymbol{x}) &=
ho_0^{ ext{opt}}(oldsymbol{x}), arphi_1(oldsymbol{x}) \hat{arphi}_1(oldsymbol{x}) &=
ho_1^{ ext{opt}}(oldsymbol{x}) \end{aligned}$$

• Recover $ho^{\mathrm{opt}}({m x},t) = arphi({m x},t) \hat{arphi}({m x},t), {m u}^{\mathrm{opt}}({m x},t) = 2\epsilon
abla \log arphi({m x},t)$

Density Control: Schrödinger Bridge Problem



Density Control

 Challenge: How do we solve this problem for nonlinear f?

 Applications: dynamic shaping of swarms, stochastic motion planning

• **Idea:** in many applications, (f, G) is feedback linearizable

Density Control for Feedback Linearizable Systems

$$\begin{split} \inf_{u \in \mathcal{U}} & \qquad \mathbb{E}\left\{\int_0^1 \frac{1}{2}\|u(\mathbf{x},t)\|_2^2 \,\mathrm{d}t\right\} \\ \text{subject to} & \qquad \dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}) + \mathbf{G}(\mathbf{x})\mathbf{u} \\ & \qquad \mathbf{x}(t=0) \sim \rho_0(\mathbf{x}) \quad \mathbf{x}(t=1) \sim \rho_1(\mathbf{x}) \end{split}$$

- $\mathbf{x} \in \mathbb{R}^n$, $\mathbf{u} \in \mathbb{R}^m$ $m \le n$
- f(x), G(x) feedback linearizable \Leftrightarrow there exist $(\delta(x), \Gamma(x), \tau(x))$ s.t.

$$(\nabla \tau (f(x) + G(x)\delta(x)))_{x=\tau^{-1}(z)} = Az$$

 $(\nabla \tau (G(x)\Gamma(x)))_{x=\tau^{-1}(z)} = B$

and (A, B) controllable.

• change of coordinates from $(x, u) \mapsto (z, v)$ where $\dot{z} = Az + Bv$ and recover $u = \delta(x) + \Gamma(x)v$.

Density Control for Feedback Linearizable Systems

Main Idea: Use the diffeomorphism $\tau: \mathcal{X} \mapsto \mathcal{Z}$

$$\sigma_i(\mathbf{z}) := oldsymbol{ au}_\sharp
ho_i = rac{
ho_i(au^{-1}(\mathbf{z}))}{|\mathrm{det}(
abla_\mathbf{z} oldsymbol{ au}_{\mathbf{z} = oldsymbol{ au}^{-1}(\mathbf{z})}|}, \quad i \in \{0, 1\}.$$

Define $\delta_{\tau}:=\delta\circ\tau^{-1}, \Gamma_{\tau}:=\Gamma\circ\tau^{-1}$ to reformulate the problem as:

$$\inf_{m{v} \in \mathcal{V}} \qquad \mathbb{E} \left\{ \int_0^1 \frac{1}{2} \mathcal{J}(\pmb{z}, \pmb{v}, t) \, \mathrm{d}t \right\}$$
 subject to
$$\dot{\pmb{z}} = \pmb{A} \pmb{z} + \pmb{B} \pmb{v}$$

$$\pmb{z}(t=0) \sim \sigma_0(\pmb{z}) \quad \pmb{z}(t=1) \sim \sigma_1(\pmb{z})$$
 where $\mathcal{J}(\pmb{z}, \pmb{v}, t) = \|\pmb{\delta}_{\mathcal{T}}(\pmb{z}) + \pmb{\Gamma}_{\mathcal{T}}(\pmb{z}) \pmb{v}\|_2^2$.

Optimal Solution in Feedback Linearized Coordinates

Solve for optimal PDF σ^{opt} and optimal control given by

$$\begin{split} & \boldsymbol{v}^{\mathrm{opt}}(\boldsymbol{z},t) = (\boldsymbol{\Gamma}_{\boldsymbol{\tau}}^{\top} \boldsymbol{\Gamma}_{\boldsymbol{\tau}}(\boldsymbol{z}))^{-1} \boldsymbol{B}^{\top} \nabla_{\boldsymbol{z}} \psi - \boldsymbol{\Gamma}_{\boldsymbol{\tau}}^{-1}(\boldsymbol{z}) \delta_{\boldsymbol{\tau}}(\boldsymbol{z}) \\ & \frac{\partial \psi}{\partial t} + \langle \nabla_{\boldsymbol{z}} \psi, \boldsymbol{A} \boldsymbol{z} \rangle - \langle \nabla_{\boldsymbol{z}} \psi, \boldsymbol{B} \boldsymbol{\Gamma}_{\boldsymbol{\tau}}^{-1}(\boldsymbol{z}) \delta_{\boldsymbol{\tau}}(\boldsymbol{z}) \rangle \\ & + \frac{1}{2} \langle \nabla_{\boldsymbol{z}} \psi, \boldsymbol{B} \left(\boldsymbol{\Gamma}_{\boldsymbol{\tau}}^{\top}(\boldsymbol{z}) \boldsymbol{\Gamma}_{\boldsymbol{\tau}}(\boldsymbol{z}) \right)^{-1} \boldsymbol{B}^{\top} \nabla_{\boldsymbol{z}} \psi \rangle = 0 \\ & \frac{\partial \sigma^{\mathrm{opt}}}{\partial t} + \nabla_{\boldsymbol{z}} \cdot \left(\left(\boldsymbol{A} \boldsymbol{z} + \boldsymbol{B} \boldsymbol{v}^{\mathrm{opt}} \right) \sigma^{\mathrm{opt}} \right) = 0 \\ & \sigma^{\mathrm{opt}}(\boldsymbol{z}, t = 0) = \sigma_{0}(\boldsymbol{z}), \quad \sigma^{\mathrm{opt}}(\boldsymbol{z}, t = 1) = \sigma_{1}(\boldsymbol{z}) \end{split}$$

Challenge: How to solve coupled nonlinear system of PDE's?

Schrödinger System for Feedback Linearizable Systems

Idea: Stochastic regularization $\sqrt{2\epsilon} {m B} {m \Gamma}_{m au}^{-1}({m z}) {
m d} {m w}$

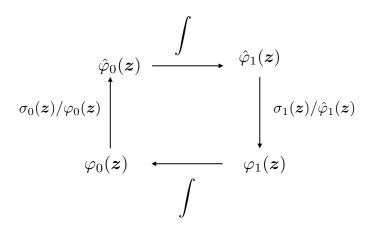
Theorem: Let
$$extbf{ extit{D}}(extbf{ extit{z}}) := extbf{ extit{B}}\Gamma_{ au}^{-1}(extbf{ extit{z}})(extbf{ extit{B}}\Gamma_{ au}^{-1}(extbf{ extit{z}}))^{ op}$$

$$\varphi(\mathbf{z},t) := \exp\left(\frac{\psi(\mathbf{z},t)}{2\epsilon}\right), \quad \hat{\varphi}(\mathbf{z},t) := \sigma^{\mathrm{opt}}(\mathbf{z},t) \exp\left(-\frac{\psi(\mathbf{z},t)}{2\epsilon}\right),$$

then we have the boundary coupled linear system

$$\begin{split} &\frac{\partial \varphi}{\partial t} + \langle \nabla_{\mathbf{z}} \varphi, \mathbf{A} \mathbf{z} - \mathbf{B} \Gamma_{\boldsymbol{\tau}}^{-1} \boldsymbol{\delta}_{\boldsymbol{\tau}}(\mathbf{z}) \rangle + \epsilon \langle \mathbf{D}, \operatorname{Hess}(\varphi) \rangle = 0, \\ &\frac{\partial \hat{\varphi}}{\partial t} + \nabla_{\mathbf{z}} \cdot \left(\left(\mathbf{A} \mathbf{z} - \mathbf{B} \Gamma_{\boldsymbol{\tau}}^{-1} \boldsymbol{\delta}_{\boldsymbol{\tau}}(\mathbf{z}) \right) \hat{\varphi} \right) \\ &- \epsilon \mathbf{1} \left(\mathbf{D}(\mathbf{z}) \odot \operatorname{Hess}(\hat{\varphi}) \right) \mathbf{1} = 0, \\ &\varphi_0(\mathbf{z}) \hat{\varphi}_0(\mathbf{z}) = \sigma_0(\mathbf{z}), \quad \varphi_1(\mathbf{z}) \hat{\varphi}_1(\mathbf{z}) = \sigma_1(\mathbf{z}) \end{split}$$

Algorithm: Fixed Point Recursion on $(\varphi_1, \hat{\varphi}_0)$



This recursion is contractive in the Hilbert Metric.

- Y. Chen, T. T. Georgiou, and M. Pavon, "Entropic and Displacement Interpolation: A Computational Approach Using the Hilbert metric, "SIAM Journal on Applied Mathematics"

Optimal Solution in Original Coordinates

- Fixed point recursion gives $\varphi_1(z), \hat{\varphi}_0(z) \rightarrow \varphi(z,t), \hat{\varphi}(z,t)$
- Obtain
 - optimal controlled PDF:

$$\sigma^{\mathrm{opt}}(\mathbf{z},t) = \varphi(\mathbf{z},t)\hat{\varphi}(\mathbf{z},t)$$

• optimal control:

$$oldsymbol{v}^{ ext{opt}}(oldsymbol{z},t) = (oldsymbol{\Gamma}_{oldsymbol{ au}}^ op oldsymbol{\Gamma}_{oldsymbol{ au}}(oldsymbol{z}))^{-1} oldsymbol{B}^ op 2\epsilon
abla_{oldsymbol{z}} \log arphi - oldsymbol{\Gamma}_{oldsymbol{ au}}^{-1}(oldsymbol{z}) oldsymbol{\delta}_{oldsymbol{ au}}(oldsymbol{z})$$

ullet Recover $ho^{\mathrm{opt}}({m x},t)$ and ${m u}^{\mathrm{opt}}({m x},t)$ as

$$ho^{ ext{opt}}(\mathbf{x},t) = \sigma^{ ext{opt}}(\mathbf{ au}(\mathbf{x}),t) | ext{det}(
abla_{\mathbf{x}}\mathbf{ au})|$$
 $\mathbf{u}^{ ext{opt}}(\mathbf{x},t) = \mathbf{\delta}(\mathbf{x}) + \mathbf{\Gamma}(\mathbf{x})\mathbf{v}^{ ext{opt}}(\mathbf{ au}^{-1}(\mathbf{x}),t)$



Summary

- Density Control subject to Multi input state feedback linearizable dynamics
- Reduced the system of coupled nonlinear PDEs to a system of boundary-coupled linear PDE's
- Algorithm via Fixed Point Recursion on initial-terminal condition pair

Thank You!

Support: NSF 1923278