Finite Horizon LQG Density Regulator with Wasserstein Terminal Cost

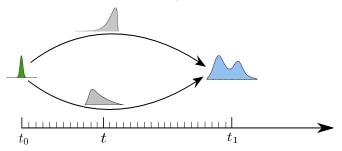
Abhishek Halder

Department of Electrical and Computer Engineering Texas A&M University College Station, TX 77843

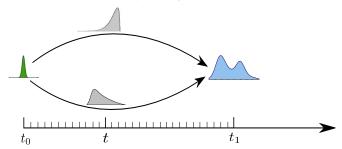
Joint work with Eric D.B. Wendel (Draper Laboratory)



Finite Horizon Density Regulator: Outlook



Finite Horizon Density Regulator: Outlook



AMS/IP Studies in Advanced Mathematics, vol 39, 2007, pp 23-35. Optimal control of the Liouville equation

R. W. Brockett

IEEE TRANSACTIONS ON AUTOMATIC CONTROL, VOL. 61, NO. 5, MAY 2016 1170

Notes on the Control of the Liouville Equation



Roger Brockett

Optimal Steering of a Linear Stochastic System to a Final Probability Distribution, Part I

Yongxin Chen, Student Member, IEEE, Tryphon T. Georgiou, Fellow, IEEE, and Michele Pavon

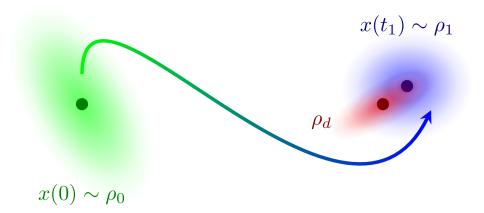
Optimal Steering of a Linear Stochastic System to a Final Probability Distribution, Part II

Yongxin Chen, Student Member, IEEE, Tryphon T. Georgiou, Fellow, IEEE, and Michele Pavon



IEEE TRANSACTIONS ON AUTOMATIC CONTROL. VOL. 61, NO. 5, MAY 2016

Density Regulator with Terminal Cost



LQG State Regulator

$$\min_{u \in \mathcal{U}} \phi(x_1, x_d) + \mathbb{E}_x \left[\int_0^{t_1} (x^\top Q x + u^\top R u) \, \mathrm{d}t \right]$$

$$dx(t) = Ax(t) dt + Bu(t) dt + F dw(t),$$

 $x(0) = x_0$ given, x_d given, t_1 fixed,

Typical terminal cost: MSE

$$\phi(x_1, x_d) = \mathbb{E}_{x_1}\left[(x_1 - x_d)^\top M(x_1 - x_d)\right]$$

LQG Density Regulator

$$\min_{u \in \mathcal{U}} \varphi\left(\rho_1, \rho_d\right) + \mathbb{E}_x\left[\int_0^{t_1} (x^\top Q x + u^\top R u) \, \mathrm{d}t\right]$$

$$dx(t) = Ax(t) dt + Bu(t) dt + F dw(t),$$

$$x(0) \sim
ho_0$$
 given, $x_d \sim
ho_d$ given, t_1 fixed,

Proposed terminal cost: MMSE

$$\varphi(x_1, x_d) = \inf_{y \sim \rho \in \mathcal{P}_2(\rho_1, \rho_d)} \mathbb{E}_y \left[(x_1 - x_d)^\top M(x_1 - x_d) \right],$$

where $y := (x_1, x_d)^\top$

Problem Statement: LQG Density Regulator

$$\begin{split} \varphi(\rho_1, \rho_d) \\ & \underset{u \in \mathcal{U}}{\min} \quad \inf_{y \sim \rho \in \mathcal{P}_2(\rho_1, \rho_d)} \mathbb{E}_y \left[(x_1 - x_d)^\top M(x_1 - x_d) \right] \\ & \quad + \mathbb{E}_x \left[\int_0^{t_1} (x^\top Q x + u^\top R u) \, dt \right] \\ & \quad dx(t) = Ax(t) \, dt + Bu(t) \, dt + F \, dw(t), \\ & \quad x(0) \sim \rho_0 = \mathcal{N} \left(\mu_0, S_0 \right), \quad x_d \sim \rho_d = \mathcal{N} \left(\mu_d, S_d \right), \\ & \quad t_1 \text{ fixed}, \quad \mathcal{U} = \{ u : u(x, t) = K(t)x + v(t) \} \end{split}$$

However, $\varphi \left(\mathcal{N} \left(\mu_1, S_1 \right), \mathcal{N} \left(\mu_d, S_d \right) \right)$ equals $\left(\mu_1 - \mu_d \right)^\top M \left(\mu_1 - \mu_d \right) +$

$$\min_{C \in \mathbb{R}^{n \times n}} \operatorname{tr} \left((S_1 + S_d - 2C)M \right) \text{ s.t. } \begin{bmatrix} S_1 & C \\ C^\top & S_d \end{bmatrix} \succeq 0$$

However, $\varphi \left(\mathcal{N} \left(\mu_1, S_1 \right), \mathcal{N} \left(\mu_d, S_d \right) \right)$ equals $\left(\mu_1 - \mu_d \right)^\top M \left(\mu_1 - \mu_d \right) +$

$$\begin{split} \min_{C \in \mathbb{R}^{n \times n}} \operatorname{tr} \left((S_1 + S_d - 2C)M \right) \, \text{s.t.} \, \begin{bmatrix} S_1 & C \\ C^\top & S_d \end{bmatrix} \succeq 0 \\ \\ & \updownarrow \\ \max_{C \in \mathbb{R}^{n \times n}} \operatorname{tr} (CM) \quad \text{s.t.} \quad CS_d^{-1}C^\top \succeq 0 \\ \\ & \updownarrow \\ \\ C^* = S_1 S_d^{\frac{1}{2}} \left(S_d^{\frac{1}{2}}S_1 S_d^{\frac{1}{2}} \right)^{-\frac{1}{2}} S_d^{\frac{1}{2}} \end{split}$$

This gives

$$\varphi\left(\mathcal{N}\left(\mu_{1}, S_{1}\right), \mathcal{N}\left(\mu_{d}, S_{d}\right)\right) = \left(\mu_{1} - \mu_{d}\right)^{\top} M\left(\mu_{1} - \mu_{d}\right)$$
$$+ \operatorname{tr}\left(MS_{1} + MS_{d} - 2\left[\left(\sqrt{S_{d}}MS_{1}\sqrt{S_{d}}\right)\left(\sqrt{S_{d}}S_{1}\sqrt{S_{d}}\right)^{-\frac{1}{2}}\right]\right)$$

Applying maximum principle:

$$K^o(t) = R^{-1}B^{ op}P(t),$$

 $v^o(t) = R^{-1}B^{ op}(z(t) - P(t)\mu(t)),$

 ∞ dim. TPBVP $\rightsquigarrow 2\left(n + \frac{n(n+1)}{2}\right)$ dim. TPBVP $\begin{pmatrix} \dot{\mu}(t) \\ \dot{z}(t) \end{pmatrix} = \begin{pmatrix} A & BR^{-1}B^{+} \\ O & -A^{\top} \end{pmatrix} \begin{pmatrix} \mu(t) \\ z(t) \end{pmatrix},$ $\dot{S}(t) = (A + BK^{o})S(t) + S(t)(A + BK^{o})^{\top} + FF^{\top},$ $\dot{P}(t) = -A^{\top}P(t) - P(t)A - P(t)BR^{-1}B^{\top}P(t) + Q,$ **Boundary conditions:**

 $\mu(0) = \mu_0, \quad z(t_1) = M(\mu_d - \mu_1),$

$$S(0) = S_0, \quad P(t_1) = \left(S_d \# S_1^{-1} - I_n\right) M$$

Matrix Geometric Mean

The minimal geodesic $\gamma^* : [0,1] \mapsto \mathbf{S}_n^+$ connecting $\gamma(0) = S_d$ and $\gamma(1) = S_1^{-1}$, associated with the Riemannian metric $g_A(S_d, S_1^{-1}) = \operatorname{tr} (A^{-1}S_d A^{-1}S_1^{-1})$, is $\gamma^*(t) = S_d \#_t S_1^{-1} = S_d^{\frac{1}{2}} \left(S_d^{-\frac{1}{2}} S_1^{-1} S_d^{-\frac{1}{2}} \right)^t S_d^{\frac{1}{2}}$ $= S_1^{-1} \#_{1-t} S_d = S_1^{-\frac{1}{2}} \left(S_1^{\frac{1}{2}} S_d S_1^{\frac{1}{2}} \right)^{1-t} S_1^{-\frac{1}{2}}$

Geometric Mean: $\gamma^*\left(\frac{1}{2}\right) = S_d \#_{\frac{1}{2}} S_1^{-1} = S_1^{-1} \#_{\frac{1}{2}} S_d$

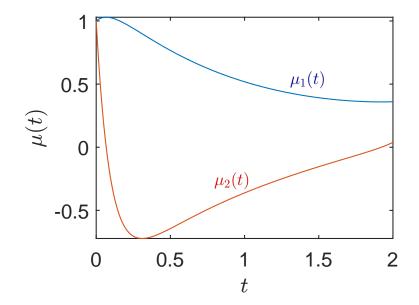
Example

$$\begin{pmatrix} dx_1 \\ dx_2 \end{pmatrix} = \begin{bmatrix} 0 & 1 \\ 2 & -3 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} dt + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u dt + \begin{bmatrix} 0.01 \\ 0.01 \end{bmatrix} dw$$

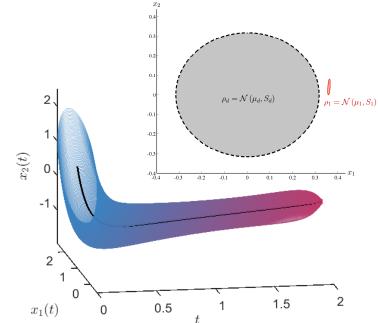
$$ho_0 = \mathcal{N}\left((1,1)^ op, I_2
ight), \hspace{1em}
ho_d = \mathcal{N}\left((0,0)^ op, 0.1\, I_2
ight),$$

 $Q = 100 I_2, \quad R = 1, \quad M = I_2, \quad t_1 = 2$

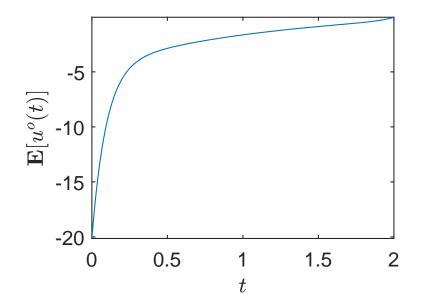
Controlled State Mean



Controlled State Covariance



Expected Optimal Control





Density regulator problem with terminal cost

LQGDR with affine state feedback

Recovers LQG/LQR as special case

Many possible extensions:

- (conjecture) affine state feedback is optimal
- output feedback
- geometry of different terminal costs

Thank you