

Taylor expansions for the HJB equation associated with a bilinear control problem

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Motivation

- ▶ dragged Brownian particle following Langevin equation

$$dX_s = Y(s)ds, \quad dY_s = -\beta Y_s ds + F(X, s)ds + \sqrt{2\beta kT/m}dB(s)$$

particle confined by potential V : force $F(X, s) = -\nabla V(X, s)$

$$\beta \gg, \quad t = \frac{s}{\beta}, \quad \nu = \frac{kT}{m},$$

- ▶ Smoluchowski equation

$$dX_t = -\nabla V(X, t) + \sqrt{2\nu}dB_t$$

- ▶ atomic force microscopy, single molecule pulling, optical trapping
- ▶ collimate light from laser into aperture of microscope objective
- ▶ control by optical tweezer $V(X_t, t) = W(X_t) + u(X_t, t)$.

The Fokker-Planck equation

Consider **probability distribution function**

$$\rho(x, t)dx = \mathbb{P}[X_t \in [x, x + dx)]$$

Fokker-Planck equation

$$\begin{aligned}\frac{\partial \rho}{\partial t} &= \nu \Delta \rho + \nabla \cdot (\rho \nabla V) && \text{in } \Omega \times (0, \infty), \\ 0 &= (\nu \nabla \rho + \rho \nabla V) \cdot \vec{n} && \text{on } \Gamma \times (0, \infty), \\ \rho(x, 0) &= \rho_0(x) && \text{in } \Omega,\end{aligned}$$

- ▶ $\Omega \subset \mathbb{R}^n$ bounded open set with boundary $\Gamma = \partial\Omega$,
- ▶ ρ_0 initial probability distribution with $\int_{\Omega} \rho_0(x)dx = 1$.

[E.G. BORZI ET AL., HARTMANN ET AL.

Control of the Fokker-Planck equation

$$\frac{\partial \rho}{\partial t} = \nu \Delta \rho + \nabla \cdot (\rho \nabla V) \quad \text{in } \Omega \times (0, \infty),$$

$$0 = (\nu \nabla \rho + \rho \nabla V) \cdot \vec{n} \quad \text{on } \Gamma \times (0, \infty),$$

$$\rho(x, 0) = \rho_0(x) \quad \text{in } \Omega,$$

- ▶ system converges to **stationary distribution** $\rho_\infty(x)$, (Boltzmann distribution)
- ▶ particles have to cross **energy barrier** between potential wells, \rightsquigarrow may be inadequately slow, $\sim \exp(\Delta W/\nu)$
- ▶ can we **control** the potential $V(x, t) = W(x) + \alpha(x)u(t)$,
- ▶ if we can, what are “good” **choices for** α ?

Assumptions:

- ▶ $W \in W^{2, \max(2, n+\varepsilon)}(\Omega), \varepsilon > 0$
- ▶ $\alpha \in W^{2, \max(2, n+\varepsilon)}(\Omega), \varepsilon > 0$ with $\nabla \alpha \cdot \vec{n} = 0$ on Γ

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Solutions to the Fokker-Planck equation

For $T > 0$ we call ρ a **(variational) solution** on $(0, T)$ if

$$\rho \in W(0, T) = L^2(0, T; H^1(\Omega)) \cap H^1(0, T; (H^1(\Omega))^*)$$

and

$$\langle \rho_t(t), v \rangle + \langle \nu \nabla \rho(t) + \rho(t) \nabla W, \nabla v \rangle + u(t) \langle \rho(t) \nabla \alpha, \nabla v \rangle = 0, \quad \forall v \in H^1(\Omega)$$

$$\rho(0) = \rho_0.$$

Proposition

(i) $u \in L^2(0, T), \rho_0 \in L^2(\Omega) \Rightarrow \exists!$ solution $\rho \in W(0, T)$

(ii) If moreover $\Delta \alpha \in L^\infty(\Omega), \rho_0 \in H^1(\Omega)$

$$\Rightarrow \rho_t \in L^2(0, T; L^2(\Omega)), \rho \in C([0, T]; H^1(\Omega))$$

Proposition

Let $u \in L^2(0, T)$ and $\rho_0 \in L^2(\Omega)$.

(i) For every $t \in [0, T]$ we have $\int_{\Omega} \rho(t) dx = \int_{\Omega} \rho_0 dx$.

(ii) If $\rho_0 \geq 0$ a.e. on Ω , then $\rho(x, t) \geq 0 \forall t > 0$ and a.e. on Ω .

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A bilinear control system

Consider the **bilinear control system**

$$\begin{aligned}\dot{\rho}(t) &= \mathcal{A}\rho(t) + u(t)\mathcal{N}\rho(t), \\ \rho(0) &= \rho_0,\end{aligned}$$

where the operators \mathcal{A} and \mathcal{N} are defined as follows

$$\begin{aligned}\mathcal{A}: \mathcal{D}(\mathcal{A}) \subset L^2(\Omega) &\rightarrow L^2(\Omega), \\ \mathcal{D}(\mathcal{A}) &= \{ \rho \in H^2(\Omega) \mid (\nu \nabla \rho + \rho \nabla W) \cdot \vec{n} = 0 \text{ on } \Gamma \}, \\ \mathcal{A}\rho &= \nu \Delta \rho + \nabla \cdot (\rho \nabla W),\end{aligned}$$

$$\mathcal{N}: H^1(\Omega) \rightarrow L^2(\Omega), \quad \mathcal{N}\rho = \nabla \cdot (\rho \nabla \alpha).$$

$$\sigma(\mathcal{A}) \subset \mathbb{R}_-, \quad \Phi(x) = \log(\nu) + \frac{W(x)}{\nu}$$

$\rho_\infty = e^{-\Phi}$ is an eigenfunction of \mathcal{A} with $\mathcal{A}\rho_\infty = 0$

A shifted problem

Instead of ρ , consider the shifted state $y := \rho - \rho_\infty$

$$\begin{aligned}\dot{y}(t) &= \mathcal{A}y(t) + u(t)\mathcal{N}y(t) + \mathcal{B}u(t), \\ y(0) &= \rho_0 - \rho_\infty,\end{aligned}$$

with $\mathcal{B} = \mathcal{N}\rho_\infty$, thus $\mathcal{B}u(t) = u(t)\mathcal{N}\rho_\infty = u(t)\nabla \cdot (\rho_\infty \nabla \alpha)$.

We decompose our state space

$$L^2(\Omega) =: \mathcal{Y}_{\mathcal{P}} \oplus \mathcal{Y}_{\mathcal{Q}}$$

$$\mathcal{Y}_{\mathcal{P}} = \left\{ v \in L^2(\Omega) : \int_{\Omega} v \, dx = 0 \right\}, \quad \mathcal{Y}_{\mathcal{Q}} = \text{span} \{ \rho_\infty \}$$

$$y = y_{\mathcal{P}} + y_{\mathcal{Q}} = \mathcal{P}y + \mathcal{Q}y.$$

Decoupling the problem

~> applying \mathcal{P} and \mathcal{Q} yields

$$\begin{pmatrix} \dot{y}_{\mathcal{P}} \\ \dot{y}_{\mathcal{Q}} \end{pmatrix} = \begin{pmatrix} \mathcal{P}\mathcal{A} & \mathcal{P}\mathcal{A} \\ \mathcal{Q}\mathcal{A} & \mathcal{Q}\mathcal{A} \end{pmatrix} \begin{pmatrix} y_{\mathcal{P}} \\ y_{\mathcal{Q}} \end{pmatrix} + u \begin{pmatrix} \mathcal{P}\mathcal{N} & \mathcal{P}\mathcal{N} \\ \mathcal{Q}\mathcal{N} & \mathcal{Q}\mathcal{N} \end{pmatrix} \begin{pmatrix} y_{\mathcal{P}} \\ y_{\mathcal{Q}} \end{pmatrix} + u \begin{pmatrix} \mathcal{P}\mathcal{B} \\ \mathcal{Q}\mathcal{B} \end{pmatrix}.$$

~> simplifies to

$$\begin{pmatrix} \dot{y}_{\mathcal{P}} \\ \dot{y}_{\mathcal{Q}} \end{pmatrix} = \begin{pmatrix} \mathcal{P}\mathcal{A} & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} y_{\mathcal{P}} \\ y_{\mathcal{Q}} \end{pmatrix} + u \begin{pmatrix} \mathcal{P}\mathcal{N} & \mathcal{P}\mathcal{N} \\ 0 & 0 \end{pmatrix} \begin{pmatrix} y_{\mathcal{P}} \\ y_{\mathcal{Q}} \end{pmatrix} + u \begin{pmatrix} \mathcal{P}\mathcal{B} \\ 0 \end{pmatrix}.$$

~> simplifies to

$$\begin{aligned} \dot{y}_{\mathcal{P}} &= \hat{\mathcal{A}}y_{\mathcal{P}} + u\hat{\mathcal{N}}y_{\mathcal{P}} + \hat{\mathcal{B}}u, & y_{\mathcal{P}}(0) &= \mathcal{P}\rho_0, \\ y_{\mathcal{Q}}(t) &= \mathcal{Q}\rho_0 - \rho_{\infty} = 0, & t &\geq 0. \end{aligned}$$

An LQR problem for the linearized system

For $\delta > 0$ let us focus on the **linearized system**

$$\dot{y}_{\mathcal{P}} = (\hat{\mathcal{A}} + \delta I)y_{\mathcal{P}}(t) + \hat{\mathcal{B}}u, \quad y_{\mathcal{P}}(0) = \mathcal{P}\rho_0,$$

together with the **quadratic cost functional**

$$J(y_{\mathcal{P}}, u) = \frac{1}{2} \int_0^{\infty} \langle y_{\mathcal{P}}(t), \mathcal{M}y_{\mathcal{P}}(t) \rangle_{L^2(\Omega)} dt + \frac{1}{2} \int_0^{\infty} |u(t)|^2 dt,$$

where $\mathcal{M} \in \mathcal{L}(\mathcal{Y}_{\mathcal{P}})$ is a self-adjoint nonnegative operator on $\mathcal{Y}_{\mathcal{P}}$.

Riccati-based feedback law: $u = -\hat{\mathcal{B}}^* \hat{\Pi} y_{\mathcal{P}}$

$$(\hat{\mathcal{A}}^* + \delta I)\hat{\Pi} + \hat{\Pi}(\hat{\mathcal{A}} + \delta I) - \hat{\Pi}\hat{\mathcal{B}}\hat{\mathcal{B}}^*\hat{\Pi} + \mathcal{M} = 0, \quad \hat{\Pi} \in \mathcal{L}(\mathcal{Y}_{\mathcal{P}}).$$

The ∞ -dimensional Hautus test: The pair $(\hat{\mathcal{A}}, \hat{\mathcal{B}})$ is **δ -stabilizable** if

$$\ker(\lambda I - \hat{\mathcal{A}}^*) \cap \ker(\hat{\mathcal{B}}^*) = \{0\} \quad \text{for } \lambda \in \overline{\mathbb{C}}_{-\delta} \cap \sigma(\hat{\mathcal{A}}^*).$$

\leadsto into nonlinear system: compare e.g.: M. Badra, I. Lasiecka, J.-P. Raymond, R. Triggiani, ...

Fokker-Planck equation

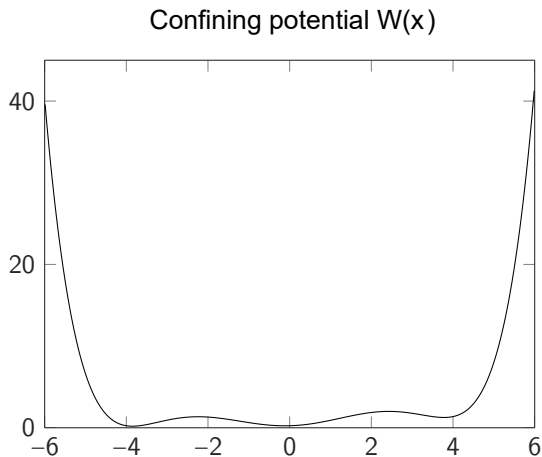


Figure: 1D Fokker-Planck equation, $n = 1024$.

Fokker-Planck equation

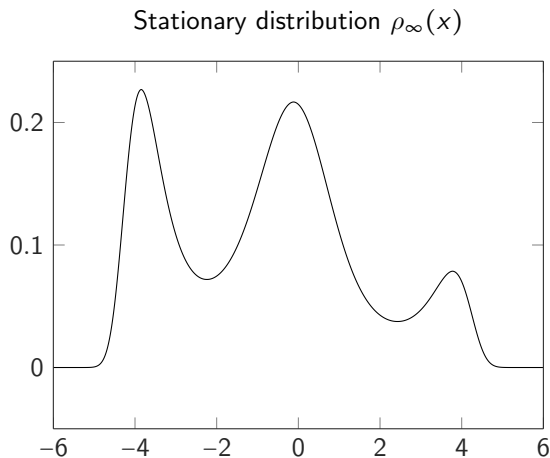


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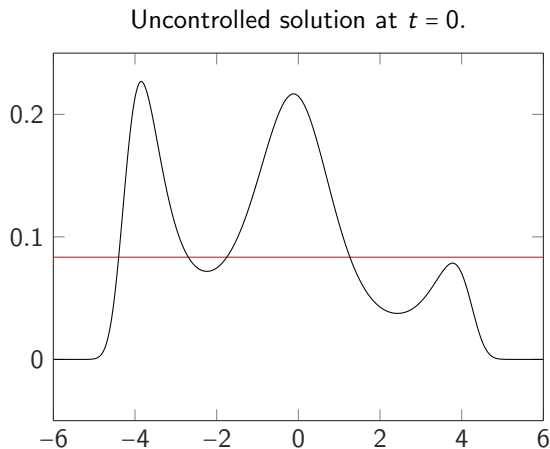


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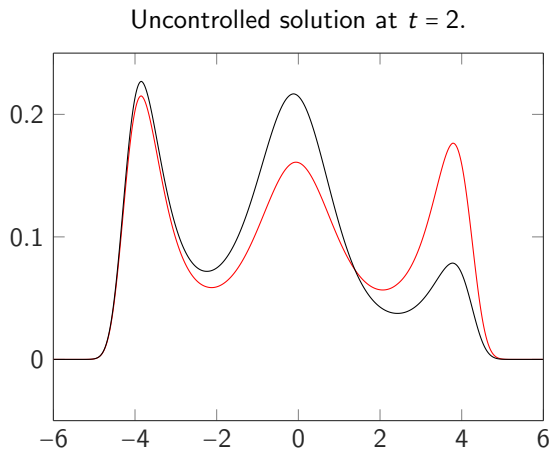


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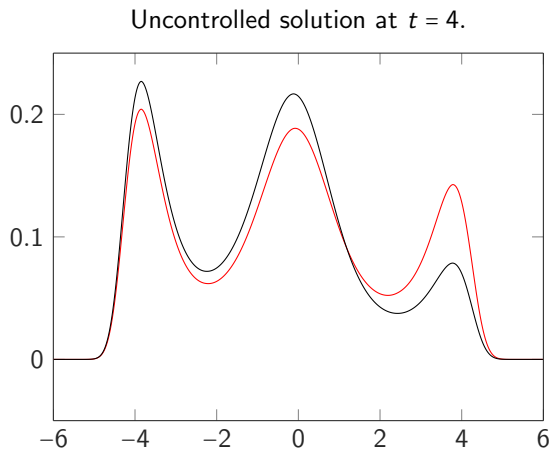


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Fokker-Planck equation

Uncontrolled solution at $t = 6$.

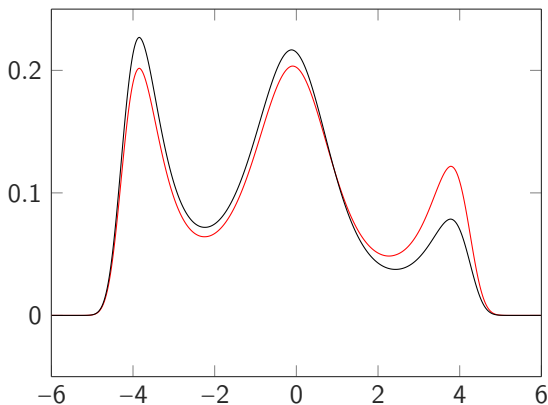


Figure: 1D Fokker-Planck equation, $n = 1024$.

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Uncontrolled solution at $t = 8$.

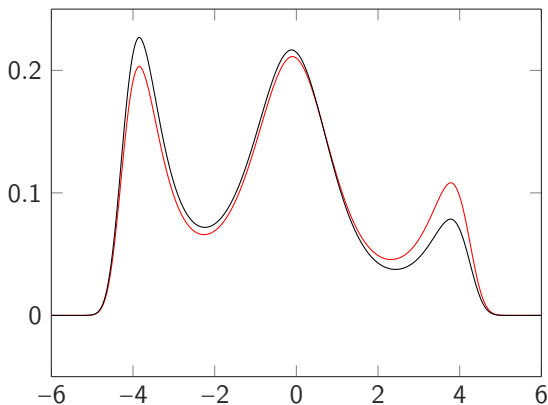


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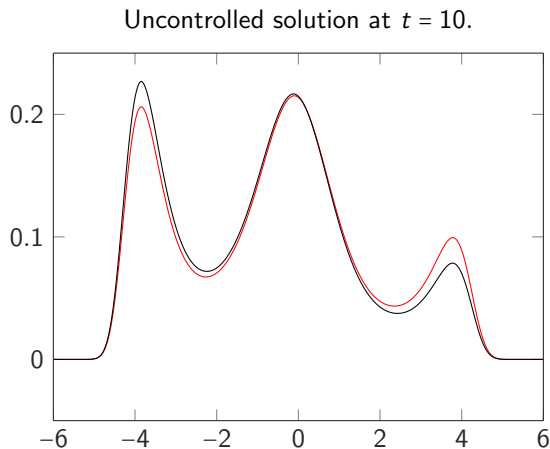


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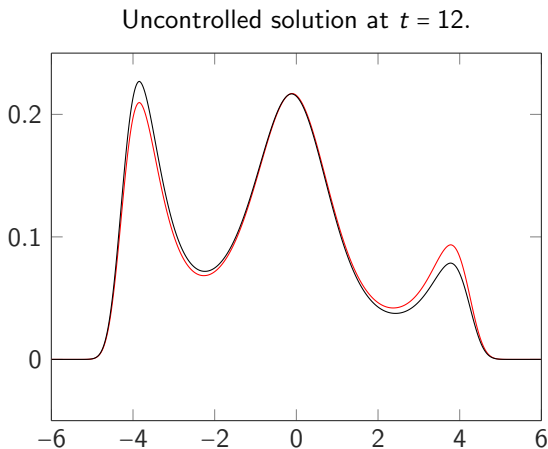


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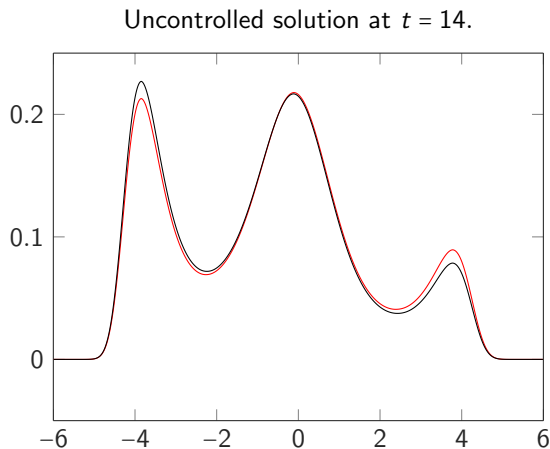


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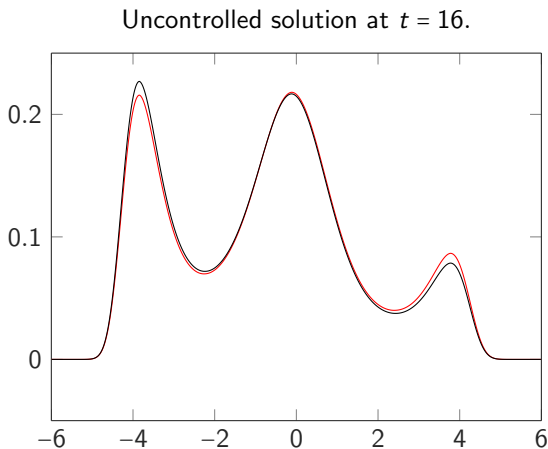


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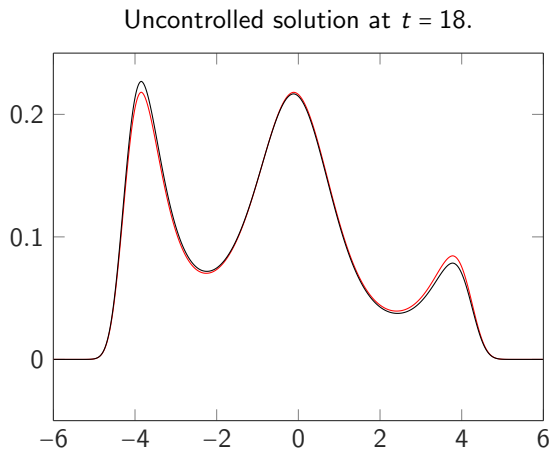


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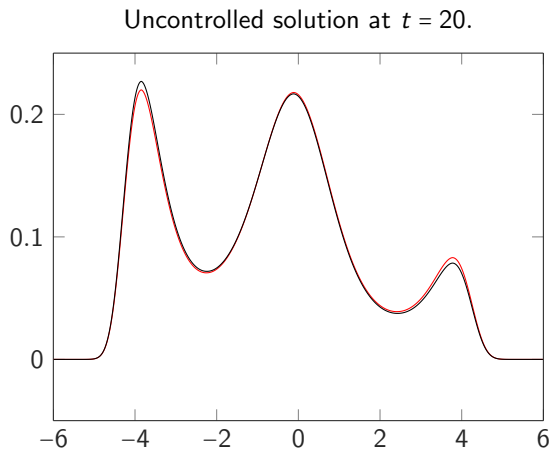


Figure: 1D Fokker-Planck equation, $n = 1024$.

Fokker-Planck equation

Uncontrolled solution at $t = 0$.

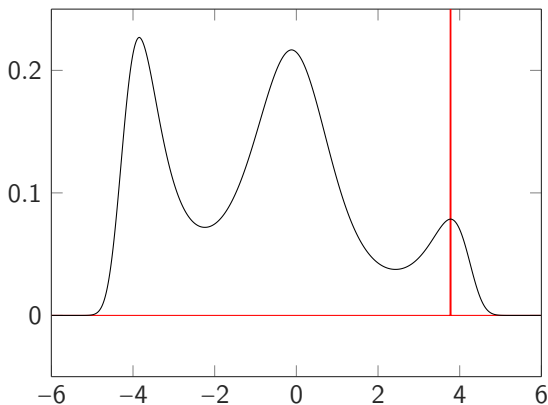


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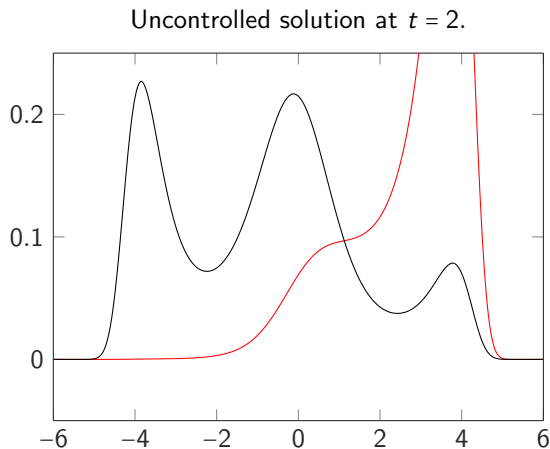


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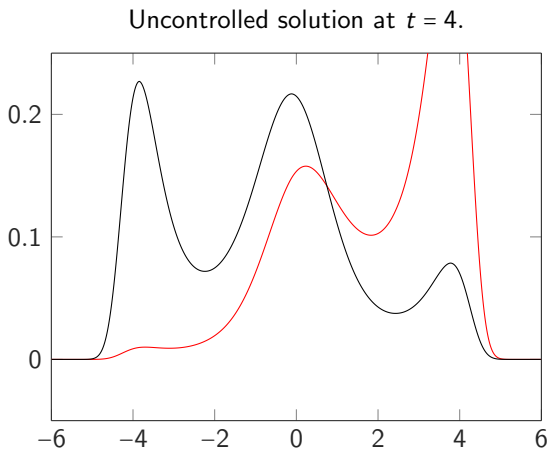


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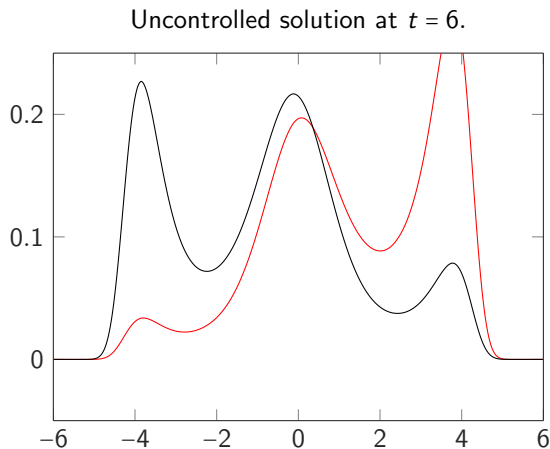


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Fokker-Planck equation

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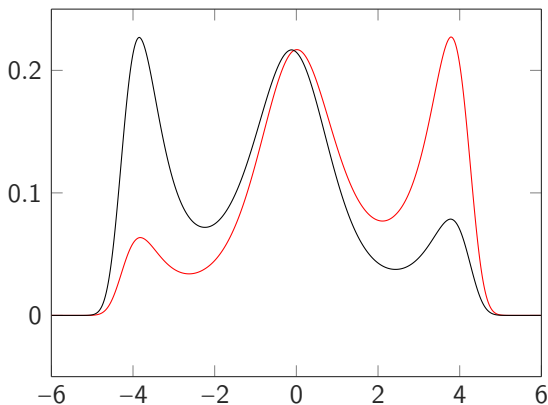


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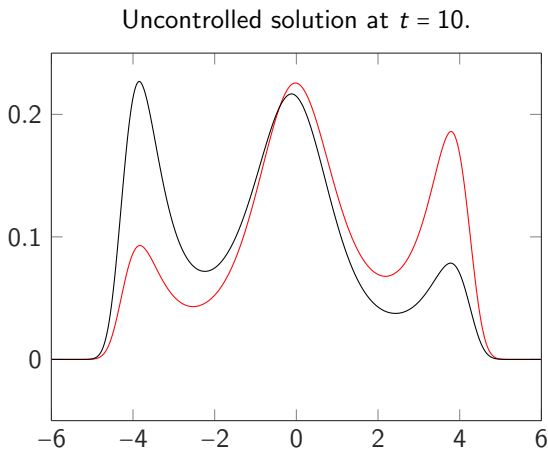


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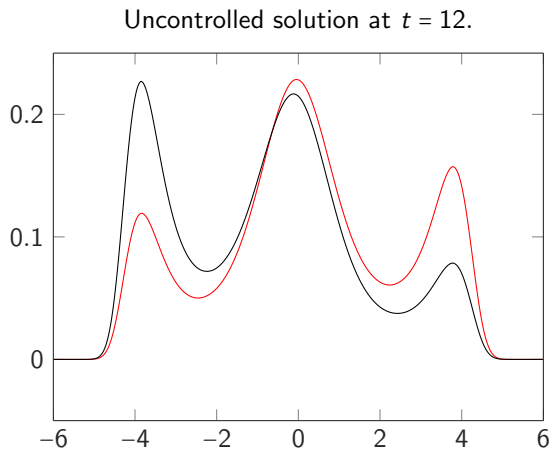


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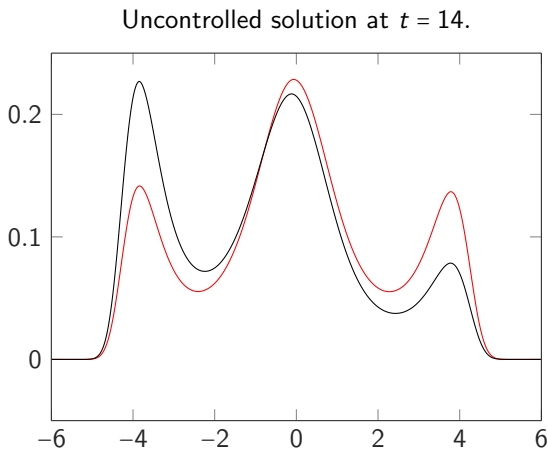


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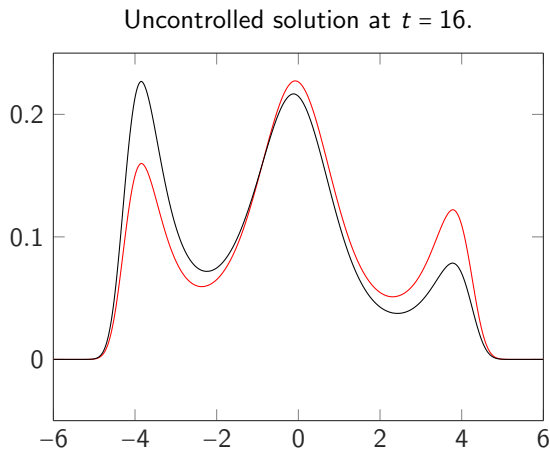


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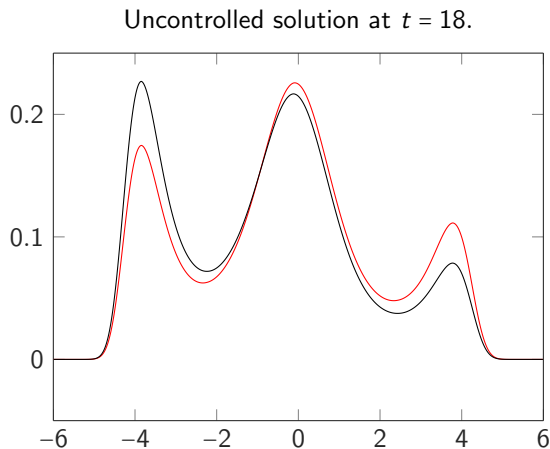


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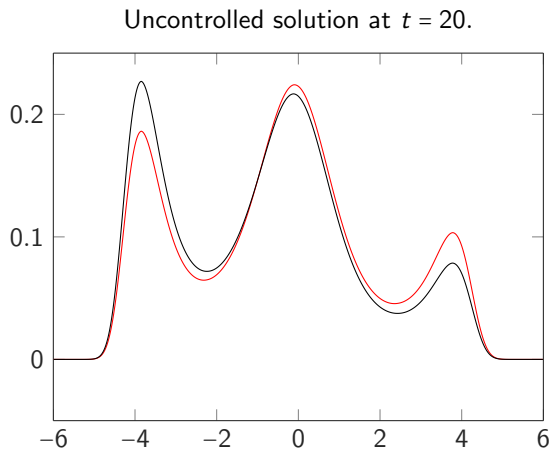


Figure: 1D Fokker-Planck equation, $n = 1024$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 0$.

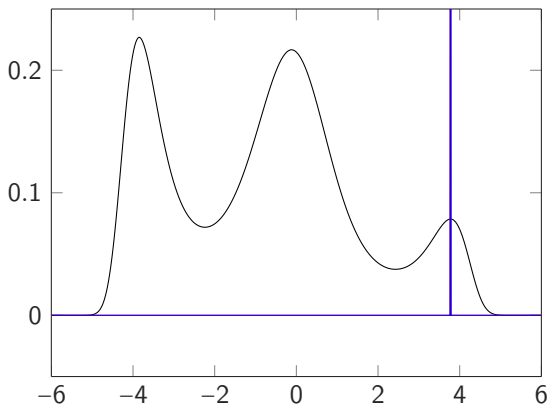


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 2$.

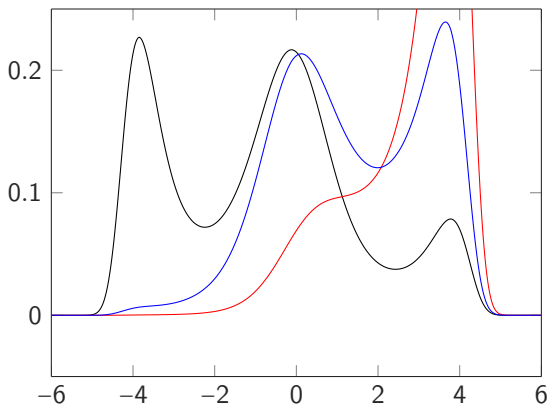


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 4$.

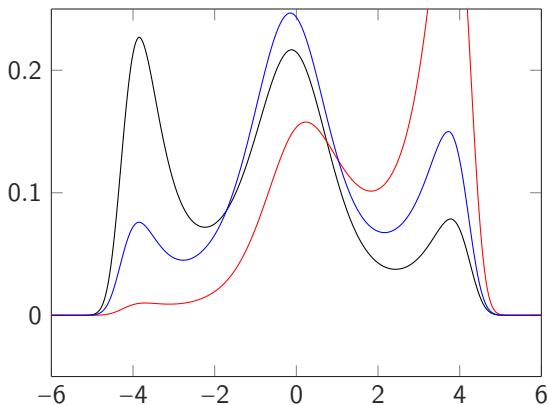


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 6$.

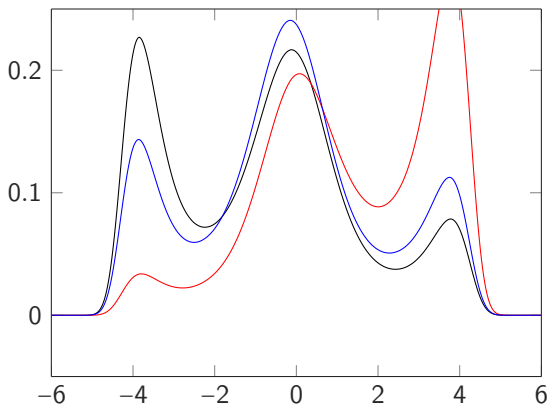


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 8$.

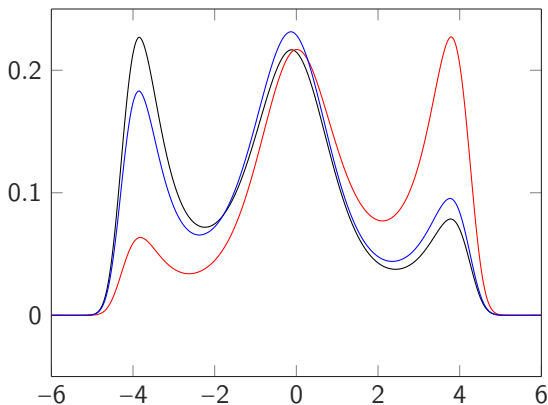


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 10$.

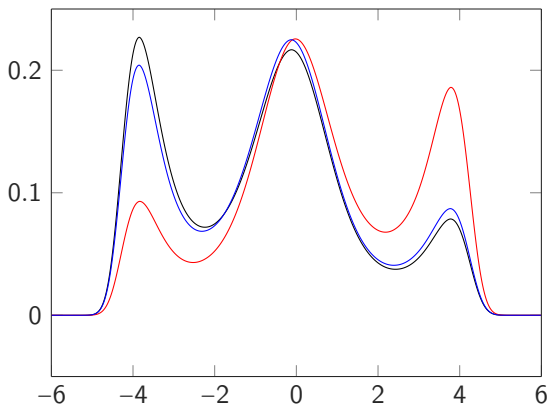


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 12$.

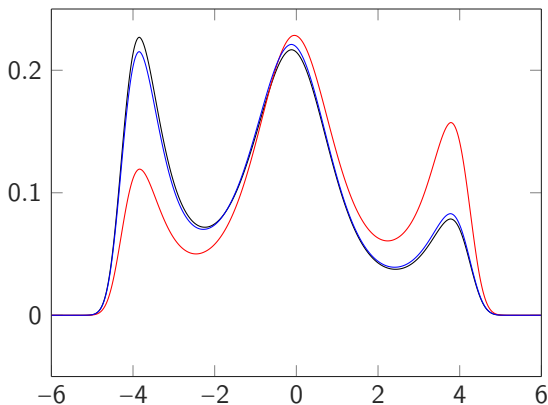


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 14$.

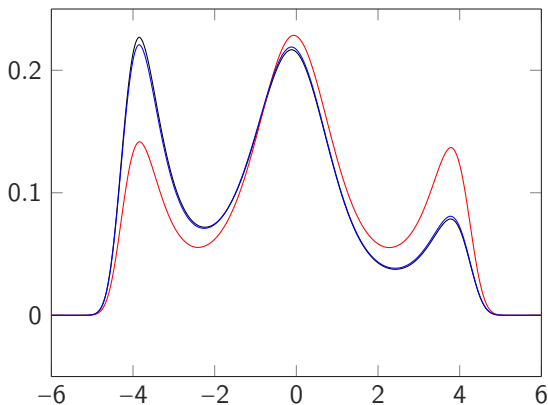


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 16$.

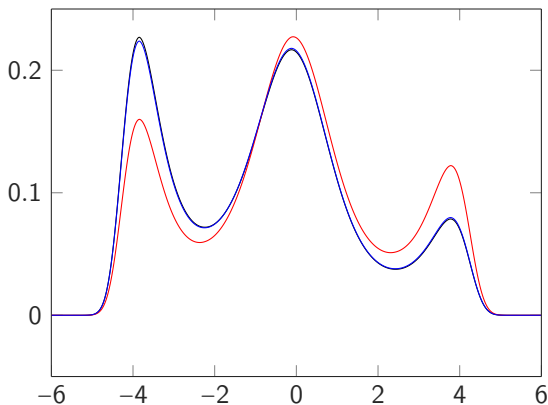


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 18$.

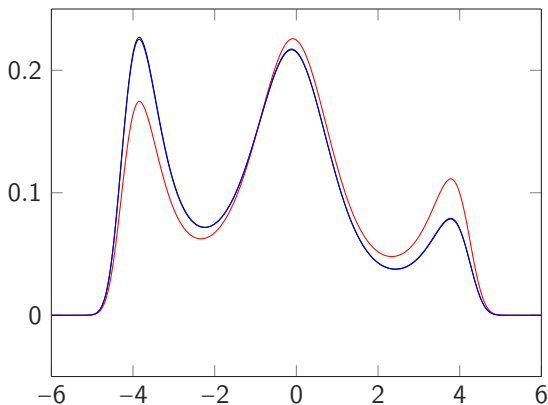


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

Back to the 1D Fokker-Planck equation

(Un)controlled solution at $t = 20$.

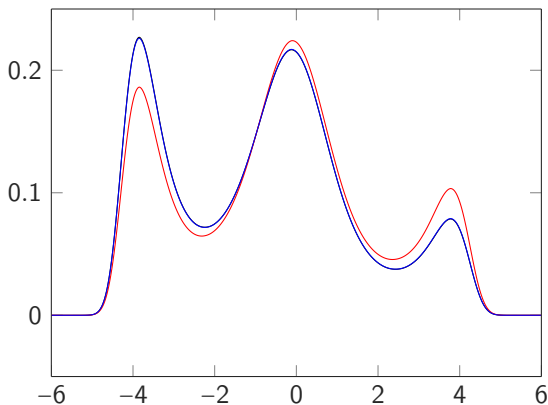


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-5}$.

2-D and movies removed

Comparison of control laws $u(t)$

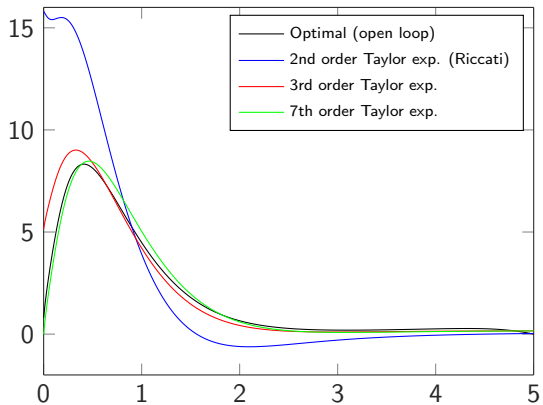


Figure: 1D Fokker-Planck equation, $n = 1024$.

Bilinear quadratic optimization

Consider a **bilinear control system**

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Nx(t)u(t) + Bu(t), & x(0) &= x_0, \\ y(t) &= Cx(t),\end{aligned}$$

- ▶ $A, N \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}$,
- ▶ **control** $u: [0, \infty) \rightarrow \mathbb{R}$ and
- ▶ **output** $y: [0, \infty) \rightarrow \mathbb{R}^p$ of the system,
- ▶ (A, B) stabilizable.

For this system, we introduce the **minimal value functional**

$$\mathcal{V}(x_0) = \inf_{u \in L^2(0, \infty)} \frac{1}{2} \int_0^\infty \|y(t)\|^2 dt + \frac{\alpha}{2} \int_0^\infty u(t)^2 dt.$$

Dynamic programming

By the **dynamic programming principle**, for any x_0 and $\tau > 0$:

$$\mathcal{V}(x_0) = \inf_{u \in L^2(0, \tau)} \int_0^\tau \ell(y(u, x_0; t), u(t)) dt + \mathcal{V}(x(u, x_0; \tau)),$$

henceforth $\ell(y, u) = \frac{1}{2} \|y\|^2 + \frac{\alpha}{2} u^2$.

Under smoothness assumptions on \mathcal{V} , we obtain

$$\min_{u \in \mathbb{R}} \left[(Ax + (Nx + B)u)^T \nabla \mathcal{V}(x) + \frac{1}{2} \|Cx\|^2 + \frac{\alpha}{2} u^2 \right] = 0, \quad \mathcal{V}(0) = 0.$$

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The Hamilton-Jacobi-Bellman equation

Consider again

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Minimization yields **Hamilton-Jacobi-Bellman** (HJB) equation

$$x^T A^T \nabla \mathcal{V}(x) + \frac{1}{2} \|Cx\|^2 - \frac{1}{2\alpha} ((Nx + B)^T \nabla \mathcal{V}(x))^2 = 0, \quad \mathcal{V}(0) = 0.$$

Optimal feedback law via solving HJB equation

$$u_{\text{opt}}(x) = -\frac{1}{\alpha} (Nx + B)^T \nabla \mathcal{V}(x).$$

Problem: The HJB equation is a nonlinear n -dimensional PDE...

The Hamilton-Jacobi-Bellman equation

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Problem: The HJB equation is a nonlinear n -dimensional PDE...

HJB equation: the linear case

For the linear case, we obtain

$$x^T A^T \nabla \mathcal{V}(x) + \frac{1}{2} \|Cx\|^2 - \frac{1}{2\alpha} (B^T \nabla \mathcal{V}(x))^2 = 0, \quad \mathcal{V}(0) = 0.$$

The **ansatz** $\mathcal{V}(x) = \frac{1}{2} x^T \Pi x$, with $\Pi = \Pi^T \in \mathbb{R}^{n \times n}$ yields

$$x^T A^T \Pi x + \frac{1}{2} \|Cx\|^2 - \frac{1}{2\alpha} (B^T \Pi x)^2 = 0.$$

The more familiar expression is

$$\frac{1}{2} x^T \left(A^T \Pi + \Pi A + C^T C - \frac{1}{\alpha} \Pi B B^T \Pi \right) x = 0.$$

\Rightarrow **algebraic Riccati equation**

Taylor expansions – basic idea

Assume that \mathcal{V} can be **expanded around 0** as follows

$$\mathcal{V}(x) = \underbrace{\mathcal{V}(0)}_{\in \mathbb{R}} + \underbrace{D\mathcal{V}(0)}_{\in \mathbb{R}^n}(x) + \frac{1}{2!} \underbrace{D^2\mathcal{V}(0)}_{\in \mathbb{R}^{n \times n}}(x, x) + \frac{1}{3!} \underbrace{D^3\mathcal{V}(0)}_{\in \mathbb{R}^{n \times n \times n}}(x, x, x) + \dots$$

Feedback law can be determined via

$$u = -\frac{1}{\alpha} \sum_{k=2}^{\infty} \frac{1}{(k-1)!} D^k \mathcal{V}(0)(Nx + B, x, \dots, x)$$

Finite-dimensional case: [LUKES, CEBUHAR/COSTANZA, KRENER]

Infinite-dimensional case: [THEVENET/BUCHOT/RAYMOND]

Question: Precise structure of $D^k \mathcal{V}(0)$?

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Taylor expansions – Differentiating HJB

Let us come back to

$$x^T A^T \nabla \mathcal{V}(x) + \frac{1}{2} \|Cx\|^2 - \frac{1}{2\alpha} ((Nx + B)^T \nabla \mathcal{V}(x))^2 = 0,$$

Taylor expansions – Differentiating HJB

Let us come back to

$$x^T A^T \nabla \mathcal{V}(x) + \frac{1}{2} \|Cx\|^2 - \frac{1}{2\alpha} ((Nx + B)^T \nabla \mathcal{V}(x))^2 = 0,$$

⇒ **one differentiation** in direction $z_1 \in \mathbb{R}^n$ yields

$$D^2 \mathcal{V}(x)(Ax, z_1) + D\mathcal{V}(x)Az_1 + \langle Cx, Cz_1 \rangle \\ - \frac{1}{\alpha} (D^2 \mathcal{V}(x)(Nx + B, z_1) + D\mathcal{V}(x)Nz_1)(D\mathcal{V}(x)(Nx + B)) = 0.$$

Taylor expansions – Differentiating HJB

Let us come back to

$$x^T A^T \nabla \mathcal{V}(x) + \frac{1}{2} \|Cx\|^2 - \frac{1}{2\alpha} ((Nx + B)^T \nabla \mathcal{V}(x))^2 = 0,$$

⇒ **two differentiations** in directions $z_1, z_2 \in \mathbb{R}^n$ yield

$$\begin{aligned} & D^3 \mathcal{V}(x)(Ax, z_1, z_2) + D^2 \mathcal{V}(x)(Az_2, z_1) + D^2 \mathcal{V}(x)(Az_1, z_2) + \langle Cz_1, Cz_2 \rangle \\ & - \frac{1}{\alpha} \left(D^2 \mathcal{V}(x)(Nx + B, z_1) + D\mathcal{V}(x)Nz_1 \right) \cdot \\ & \quad \left(D^2 \mathcal{V}(x)(Nx + B, z_2) + D\mathcal{V}(x)Nz_2 \right) \\ & - \frac{1}{\alpha} \left(D^3 \mathcal{V}(x)(Nx + B, z_1, z_2) + D^2 \mathcal{V}(x)(Nz_2, z_1) + D^2 \mathcal{V}(x)(Nz_1, z_2) \right) \\ & \quad \left(D\mathcal{V}(x)(Nx + B) \right) = 0. \end{aligned}$$

Taylor expansions – Differentiating HJB

Let us come back to

$$x^T A^T \nabla \mathcal{V}(x) + \frac{1}{2} \|Cx\|^2 - \frac{1}{2\alpha} ((Nx + B)^T \nabla \mathcal{V}(x))^2 = 0,$$

⇒ **two differentiations** in directions $z_1, z_2 \in \mathbb{R}^n$ yield

$$\begin{aligned} & \cancel{D^3 \mathcal{V}(0)(A0, z_1, z_2)} + D^2 \mathcal{V}(0)(Az_2, z_1) + D^2 \mathcal{V}(0)(Az_1, z_2) + \langle Cz_1, Cz_2 \rangle \\ & - \frac{1}{\alpha} \left(D^2 \mathcal{V}(0)(\cancel{N0} + B, z_1) + \cancel{D \mathcal{V}(0) N z_1} \right) \cdot \\ & \quad \left(D^2 \mathcal{V}(0)(\cancel{N0} + B, z_2) + \cancel{D \mathcal{V}(0) N z_2} \right) \\ & - \frac{1}{\alpha} \left(D^3 \mathcal{V}(0)(\cancel{N0} + B, z_1, z_2) + D^2 \mathcal{V}(0)(Nz_2, z_1) + D^2 \mathcal{V}(0)(Nz_1, z_2) \right) \\ & \quad \left(\cancel{D \mathcal{V}(0)(N0 + B)} \right) = 0. \end{aligned}$$

This is the Riccati equation...

Taylor expansions – Differentiating HJB

Let us come back to

$$x^T A^T \nabla \mathcal{V}(x) + \frac{1}{2} \|Cx\|^2 - \frac{1}{2\alpha} ((Nx + B)^T \nabla \mathcal{V}(x))^2 = 0,$$

⇒ **three differentiations** in directions $z_1, z_2, z_3 \in \mathbb{R}^n$ yield

$$\begin{aligned} & D^3 \mathcal{V}(0)(Az_3, z_1, z_2) + D^3 \mathcal{V}(0)(Az_2, z_1, z_3) + D^3 \mathcal{V}(0)(Az_1, z_2, z_3) \\ & - \frac{1}{\alpha} \left(D^3 \mathcal{V}(0)(B, z_1, z_3) + D^2 \mathcal{V}(0)(Nz_3, z_1) + D^2 \mathcal{V}(0)(Nz_1, z_3) \right) \left(D^2 \mathcal{V}(0)(B, z_2) \right) \\ & - \frac{1}{\alpha} \left(D^3 \mathcal{V}(0)(B, z_2, z_3) + D^2 \mathcal{V}(0)(Nz_3, z_2) + D^2 \mathcal{V}(0)(Nz_2, z_3) \right) \left(D^2 \mathcal{V}(0)(B, z_1) \right) \\ & - \frac{1}{\alpha} \left(D^3 \mathcal{V}(0)(B, z_1, z_2) + D^2 \mathcal{V}(0)(Nz_2, z_1) + D^2 \mathcal{V}(0)(Nz_1, z_2) \right) \left(D^2 \mathcal{V}(0)(B, z_3) \right) \\ & = 0. \end{aligned}$$

Taylor expansions – Differentiating HJB

Let us come back to

$$x^T A^T \nabla \mathcal{V}(x) + \frac{1}{2} \|Cx\|^2 - \frac{1}{2\alpha} ((Nx + B)^T \nabla \mathcal{V}(x))^2 = 0,$$

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$$\begin{aligned} & D^3 \mathcal{V}(0)(Az_3, z_1, z_2) + D^3 \mathcal{V}(0)(Az_2, z_1, z_3) + D^3 \mathcal{V}(0)(Az_1, z_2, z_3) \\ & - \frac{1}{\alpha} \left(D^3 \mathcal{V}(0)(B, z_1, z_3) + D^2 \mathcal{V}(0)(Nz_3, z_1) + D^2 \mathcal{V}(0)(Nz_1, z_3) \right) \left(D^2 \mathcal{V}(0)(B, z_2) \right) \\ & - \frac{1}{\alpha} \left(D^3 \mathcal{V}(0)(B, z_2, z_3) + D^2 \mathcal{V}(0)(Nz_3, z_2) + D^2 \mathcal{V}(0)(Nz_2, z_3) \right) \left(D^2 \mathcal{V}(0)(B, z_1) \right) \\ & - \frac{1}{\alpha} \left(D^3 \mathcal{V}(0)(B, z_1, z_2) + D^2 \mathcal{V}(0)(Nz_2, z_1) + D^2 \mathcal{V}(0)(Nz_1, z_2) \right) \left(D^2 \mathcal{V}(0)(B, z_3) \right) \\ & = 0. \end{aligned}$$

☹ Looks complicated ☺ Linear in $D^3 \mathcal{V}(0)$

The general structure

For $i, j \in \mathbb{N}$, consider the following set of **permutations**:

$$S_{i,j} = \{\sigma \in S_{i+j} \mid \sigma(1) < \dots < \sigma(i) \text{ and } \sigma(i+1) < \dots < \sigma(i+j)\},$$

where S_{i+j} is the set of permutations of $\{1, \dots, i+j\}$.

Example

$$\begin{aligned} S_{2,2} &= \{\sigma \in S_4 \mid \sigma(1) < \sigma(2) \text{ and } \sigma(3) < \sigma(4)\} \\ &= \{(1, 2, 3, 4), (1, 3, 2, 4), (1, 4, 2, 3), \\ &\quad (2, 3, 1, 4), (2, 4, 1, 3), (3, 4, 1, 2)\} \end{aligned}$$

For given **multilinear form** \mathcal{T} (of order $i+j$), we define

$$\text{Sym}_{i,j}(\mathcal{T})(z_1, \dots, z_{i+j}) := \binom{i+j}{i}^{-1} \left[\sum_{\sigma \in S_{i,j}} \mathcal{T}(z_{\sigma(1)}, \dots, z_{\sigma(i+j)}) \right].$$

The general structure

Define $A_{\Pi} = A - \frac{1}{\alpha} BB^* \Pi$. For $k \geq 3$ and $z_1, \dots, z_k \in \mathbb{R}^n$ consider

$$\sum_{i=1}^k D^k \mathcal{V}(0)(z_1, \dots, z_{i-1}, A_{\Pi} z_i, z_{i+1}, \dots, z_k) = \frac{1}{2\alpha} \mathcal{R}_k(z_1, \dots, z_k) \quad (*),$$

where $\mathcal{R}_k(z_1, \dots, z_k)$ is given by:

$$\begin{aligned} \mathcal{R}_k(z_1, \dots, z_k) = & 2k(k-1) \text{Sym}_{1, k-1} (C_1(z_1) \mathcal{G}_{k-1}(z_2, \dots, z_k)) \\ & + \sum_{i=2}^{k-2} \binom{k}{i} \text{Sym}_{i, k-i} \left((C_i(z_1, \dots, z_i) + i \mathcal{G}_i(z_1, \dots, z_i)) \right. \\ & \left. \times (C_{k-i}(z_{i+1}, \dots, z_k) + (k-i) \mathcal{G}_{k-i}(z_{i+1}, \dots, z_k)) \right), \end{aligned}$$

where:

$$C_i(z_1, \dots, z_i) = D^{i+1} \mathcal{V}(0)(B, z_1, \dots, z_i)$$

$$\mathcal{G}_i(z_1, \dots, z_i) = \frac{1}{i} \left[\sum_{j=1}^i D^j \mathcal{V}(0)(z_1, \dots, z_{j-1}, Nz_j, z_{j+1}, \dots, z_i) \right].$$

Tensor calculus

Main numerical task:

$$\text{Solve } \underbrace{\left(\sum_{i=1}^k I^{k-i} \otimes A_{\Pi}^T \otimes I^{i-1} \right)}_{\mathbf{A}} T_k = \underbrace{R_k(T_2, \dots, T_{k-1})}_{? \text{ low rank ?}}.$$

Since \mathbf{A} is **stable**: $\mathbf{A}^{-1} = - \int_0^{\infty} e^{t\mathbf{A}} dt = - \int_0^{\infty} \bigotimes_{i=1}^k e^{tA_{\Pi}^T} dt.$

Approximate by **quadrature formula**

[GRASEDYCK, HACKBUSCH, STENGER]

$$\mathbf{A}^{-1} \approx - \sum_{j=-r}^r w_j \bigotimes_{i=1}^k e^{t_j A_{\Pi}^T},$$

with suitable quadrature weights w_j and points t_j .

The infinite dimensional setup

Let us focus on the abstract **bilinear control system**

$$\dot{x}(t) = \mathcal{A}x(t) + \mathcal{N}x(t)u(t) + \mathcal{B}u(t), \quad x(0) = x_0 \in X,$$

- ▶ $V \subset X \subset V^*$ Gelfand triple of Hilbert spaces
- ▶ $a: V \times V \rightarrow \mathbb{R}$ a bounded V - X bilinear form, i.e., $\exists \nu > 0$ and $\lambda \in \mathbb{R}$

$$a(v, v) \geq \nu \|v\|_V^2 - \lambda \|v\|_X^2 \quad \forall v \in V$$

- ▶ $\mathcal{N} \in \mathcal{L}(V, X) \cap \mathcal{L}(\mathcal{D}(\mathcal{A}), V)$, $\mathcal{N}^* \in \mathcal{L}(V, X)$, $\mathcal{B} \in X$
- ▶ For $\beta > 0$ large enough define $\mathcal{A}_0 := -\mathcal{A} + \beta I$

$$\Rightarrow [\mathcal{D}(\mathcal{A}_0), X]_{\frac{1}{2}} = [\mathcal{D}(\mathcal{A}_0^*), X]_{\frac{1}{2}} = V$$

- ▶ $(\mathcal{A}, \mathcal{B})$ stabilizable

A multilinear operator equation

Well-posedness of $\mathcal{T}_k \equiv D^k \mathcal{V}(0)$

For $k \geq 3$, and $z_1, \dots, z_k \in X$ define the **multilinear form**

$$\begin{aligned}\mathcal{T}_k: X \times \dots \times X &\rightarrow \mathbb{R}, \\ \mathcal{T}_k(z_1, \dots, z_k) &= -\frac{1}{2\alpha} \int_0^\infty \mathcal{R}_k(e^{A\eta t} z_1, \dots, e^{A\eta t} z_k) dt.\end{aligned}$$

Then \mathcal{T}_k is the **unique solution** of (*). Moreover, it holds that

$$|\mathcal{T}_k(z_1, \dots, z_k)| \leq C \prod_{i=1}^k \|z_i\|_X.$$

Definition:

$$\mathcal{V}_p: Y \rightarrow \mathbb{R}, \quad \mathcal{V}_p(y) = \sum_{k=2}^p \frac{1}{k!} \mathcal{T}_k(y, \dots, y).$$

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A suboptimal feedback law

Consider now the **polynomial feedback** law

$$u_p(x) = - \sum_{k=2}^p \frac{1}{(k-1)!} \mathcal{T}_k(\mathcal{N}x + \mathcal{B}, x, \dots, x)$$

and the corresponding **(nonlinear) closed-loop system**

$$(CL) \quad \dot{x} = \mathcal{A}x - (\mathcal{N}x + \mathcal{B})u_p(x), \quad x(0) = x_0.$$

Local well-posedness

There exist constants $C_1, C_2 > 0$ such that: if $\|x_0\|_X \leq C_1$, then

- ▶ (CL) admits a unique solution
 $x \in W(0, \infty) = \{\varphi \in L^2(0, \infty; X) \mid \varphi_t \in L^2(0, \infty; V^*)\}$
- ▶ this solution satisfies $\|x\|_{W(0, \infty)} \leq C_2$
- ▶ and $\lim_{t \rightarrow \infty} \|x(t)\|_X = 0$

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$$(CL) \quad \dot{x} = \mathcal{A}x - (\mathcal{N}x + \mathcal{B})u_p(x), \quad x(0) = x_0.$$

Local suboptimality

There exists a constant $C_3 > 0$, C_4 such that: if $\|x_0\|_X \leq C_3$,

- ▶ $\int_0^\infty \ell(x(u_p, x_0; t), u_p(t)) dt \leq \mathcal{V}(x_0) + C_4(\|x_0\|_X^{p+1})$
- ▶ $|\mathcal{V}(x_0) - \mathcal{V}_p(x_0)| \leq C\|y_0\|_X^{p+1}$
- ▶ $\|\bar{x}(\bar{u}, x_0) - x(u_p, x_0)\|_{W(0,\infty)} \leq C_4\|x_0\|_X^{\frac{p+1}{2}}$
- ▶ $\|\bar{u} - u_p\|_{W(0,\infty)} \leq C_4\|x_0\|_X^{\frac{p+1}{2}}$

Comparison of control laws $u(t)$

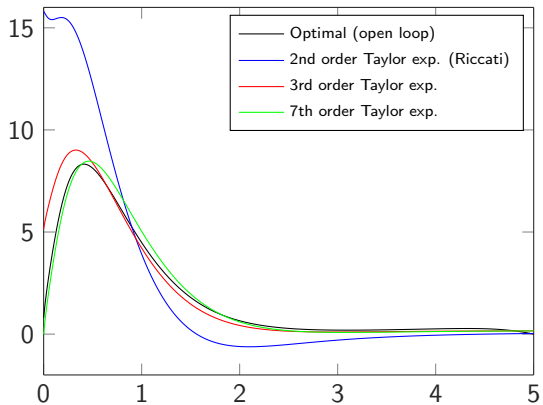


Figure: 1D Fokker-Planck equation, $n = 1024$.

Comparison of control laws $u(t)$

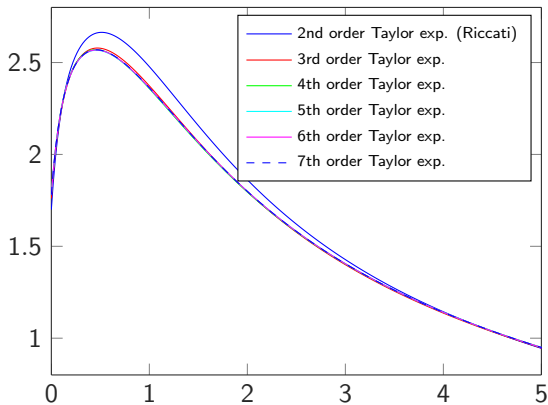


Figure: Fokker-Planck, $n = 1024$, $r = 10$, $\beta = 10^{-4}$.

Miscellanea

- ▶ Applicable to Fokker Planck
- ▶ Feasible for general infinite dimensional control systems
- ▶ Combined with balanced truncation

- ▶ When is higher order useful ?
- ▶ Efficient tensor numerics

Finite-dimensional properties

Assume spatial discretization yields $A, N, \rho^d, e = \frac{1}{d} (1, \dots, 1)^T$.

- ▶ $\int_{\Omega} \rho(t) \, dx = \int_{\Omega} \rho_0 \, dx \rightsquigarrow e^T \rho^d(t) = e^T \rho_0^d = 1$
- ▶ $\rho_0 \geq 0 \Rightarrow \rho(x, t) \geq 0 \, \forall t \rightsquigarrow A$ is a Metzler matrix
- ▶ $A_s = DAD^{-1}$ with $D = \text{diag}(e^{\frac{\phi^d}{2}})$ is a symmetric matrix
- ▶ $A^T e = N^T e = 0 = A \rho_{\infty}^d$
- ▶ $\mu = \max(\varepsilon, \varepsilon + \frac{1}{2} \lambda_{\max}(\hat{A} + \hat{A}^T))$

A two dimensional double well potential

As a numerical example, let us consider

$$\begin{aligned}\frac{\partial \rho}{\partial t} &= \nu \Delta \rho + \nabla \cdot (\rho \nabla W) + u \nabla \cdot (\rho \nabla \alpha) && \text{in } \Omega \times (0, \infty), \\ 0 &= (\nu \nabla \rho + \rho \nabla W) \cdot \vec{n} && \text{on } \Gamma \times (0, \infty), \\ \rho(x, 0) &= \rho_0(x) && \text{in } \Omega.\end{aligned}$$

- ▶ $\Omega = (-1.5, 1.5) \times (-1.1)$
- ▶ $W(x) = 3(x_1^2 - 1)^2 + 6x_2^2$
- ▶ **finite differences** with $n_x \cdot n_y = 96 \cdot 64 = 6144$ grid points
- ▶ **upwind scheme** for convective terms

Conclusion

- ▶ Fokker-Planck equation yields a bilinear control system.
- ▶ Decoupled the system by spectral projections.
- ▶ Riccati-based feedback law \Rightarrow local stabilization.
- ▶ Lyapunov-based feedback law \Rightarrow global stabilization .
- ▶ Numerically efficient approaches?
- ▶ Optimal feedback law for bilinear system?

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Thank you for your attention!