

Weyl Calculus and Exactly Solvable Schrödinger Bridges with Quadratic State Cost

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The Problem

Solve linear reaction-diffusion PDE IVP:

$$\frac{\partial}{\partial t} \hat{\varphi} = -\mathcal{L} \hat{\varphi}, \quad \hat{\varphi}(t = t_0, \mathbf{x}) = \hat{\varphi}_0(\mathbf{x}) \text{ (given)}$$

where $\mathcal{L} \equiv -\Delta_{\mathbf{x}} + q(\mathbf{x})$

given bounded continuous reaction rate

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Our contribution: closed-form Green's function for convex quadratic $q(\mathbf{x})$

$$\hat{\varphi}(t, \mathbf{x}) = \int_{\mathcal{X}} \kappa(t_0, \mathbf{x}, t, \mathbf{y}) \hat{\varphi}_0(\mathbf{y}) d\mathbf{y}$$

State-of-the-art

Randomized numerical solution via Feynman-Kac path integral

Estimates $\hat{\varphi}(t, \boldsymbol{x})$ as conditional expectation

Good: meshless / particle-based

Bad: still needs function approximation

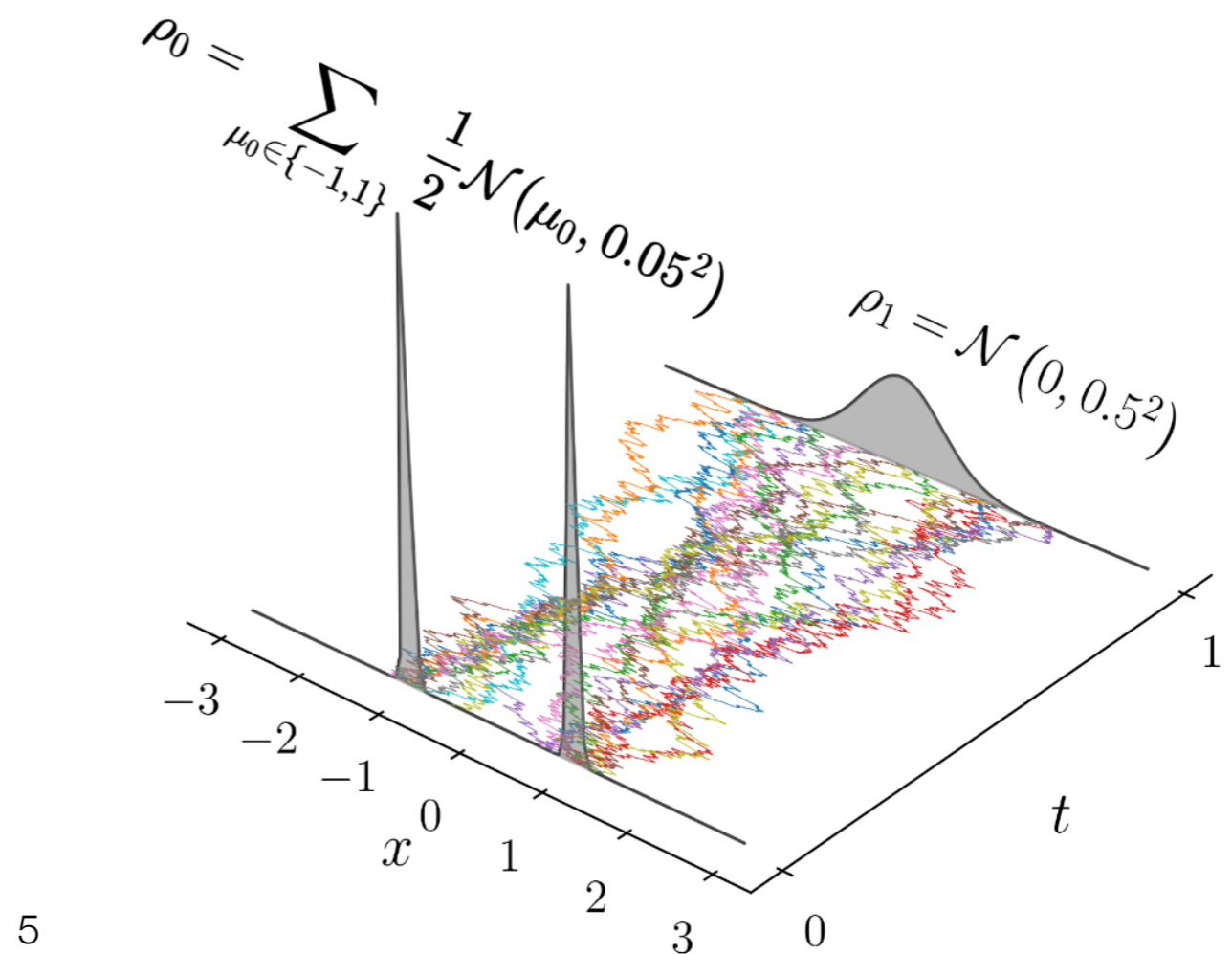
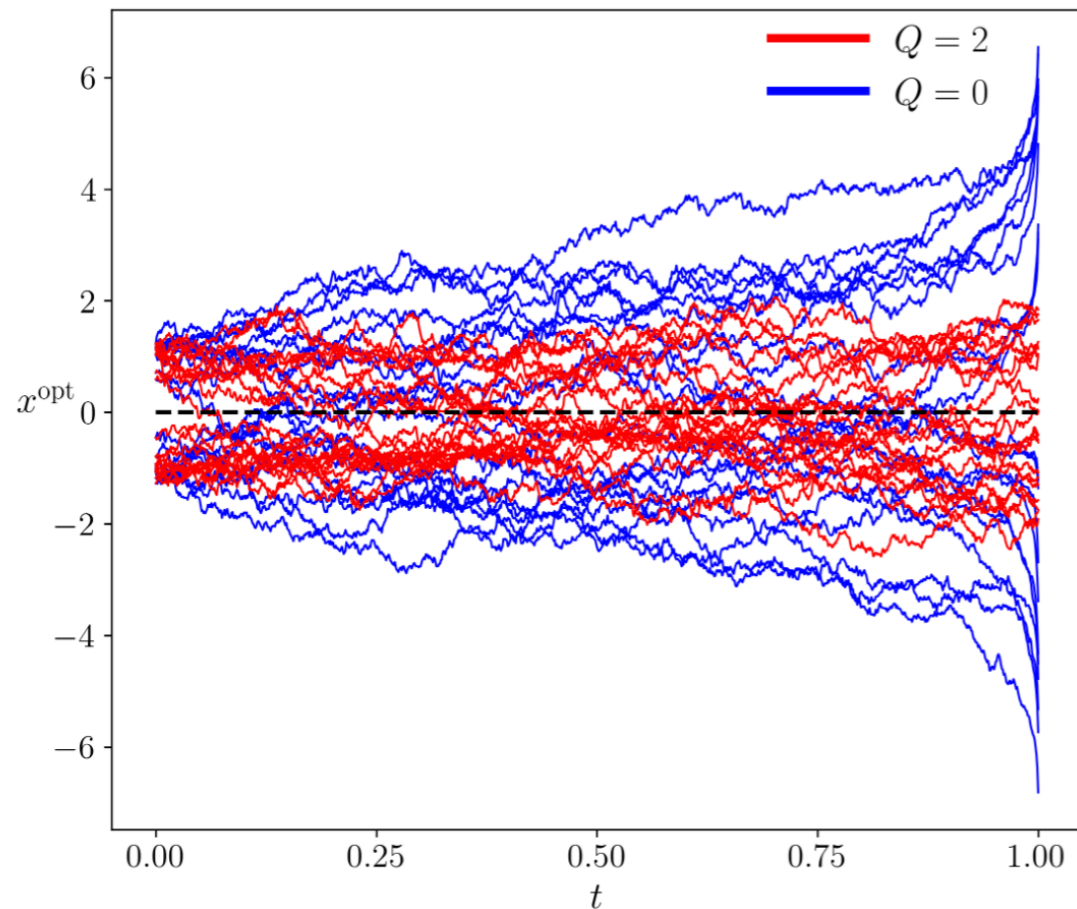
Our Motivation

Schrödinger bridge problem with state cost

$$\inf_{(\mu^u, \mathbf{u})} \int_{\mathbb{R}^n} \int_{t_0}^{t_1} \left\{ \frac{1}{2} |\mathbf{u}|^2 + q(\mathbf{x}_t^u) \right\} dt d\mu^u(\mathbf{x}_t^u)$$

$$\text{subject to } d\mathbf{x}_t^u = \mathbf{u}_t(t, \mathbf{x}_t^u) dt + \sqrt{2} d\mathbf{w}_t, \\ \mathbf{x}_t^u(t = t_0) \sim \mu_0, \quad \mathbf{x}_t^u(t = t_1) \sim \mu_1,$$

$$q(x) = \frac{1}{2} Q x^2$$



Can be reduced to solving

Nonlinearly boundary coupled system of linear PDEs

$$\frac{\partial \hat{\varphi}}{\partial t} = (\Delta_{\mathbf{x}} - q(\mathbf{x}))\hat{\varphi},$$

$$\frac{\partial \varphi}{\partial t} = (-\Delta_{\mathbf{x}} + q(\mathbf{x}))\varphi,$$

$$\hat{\varphi}(t = t_0, \cdot)\varphi(t = t_0, \cdot) = \mu_0, \quad \hat{\varphi}(t = t_0, \cdot)\varphi(t = t_1, \cdot) = \mu_1$$

Equivalently, nonlinear integral equations (Schrödinger system)

$$\mu_0(\mathbf{x}) = \hat{\varphi}_0(\mathbf{x}) \int \kappa(t_0, \mathbf{x}, t_1, \mathbf{y})\varphi_1(\mathbf{y})d\mathbf{y}$$

$$\mu_1(\mathbf{x}) = \hat{\varphi}_1(\mathbf{x}) \int \kappa(t_0, \mathbf{y}, t_1, \mathbf{x})\hat{\varphi}_0(\mathbf{y})d\mathbf{y}$$

Optimal solution: $\mu_{\text{opt}}^u(t, \cdot) = \hat{\varphi}(t, \cdot)\varphi(t, \cdot), \quad \mathbf{u}_{\text{opt}}(t, \cdot) = \nabla_{(\cdot)} \log \varphi(t, \cdot)$

The Case $q(x) = \frac{1}{2}x^\top Qx$, $Q \succeq 0$

Gaussian endpoints $\mu_0 = \mathcal{N}(\mathbf{0}, \Sigma_0)$, $\mu_1 = \mathcal{N}(\mathbf{0}, \Sigma_1)$

↪ coupled Riccati ODEs studied in Chen, Georgiu, Pavon, ACC 2015

Non-Gaussian endpoints

↪ need κ

Closed-form derivation of κ using Hermite polynomials
(elementary + painful + difficult to generalize)

arXiv:2406.00503

SCHRÖDINGER BRIDGE WITH QUADRATIC STATE COST IS
EXACTLY SOLVABLE*

ALEXIS M.H. TETER[†], WENQING WANG[‡], AND ABHISHEK HALDER[§]

More disciplined alternative: Weyl calculus

New Idea: Weyl Calculus to Compute κ

Originally developed for quantum mechanics

Appears less known to solve problems outside quantum mechanics

Many names: Weyl quantization, Wigner-Weyl transform

Proposed computational steps:

PDE \longrightarrow Weyl operator $H(X, D) \longrightarrow$ Weyl symbol $h(x, \xi) \longrightarrow$ Kernel κ

Step 1: PDE \longrightarrow Weyl Operator $H(\mathbf{X}, \mathbf{D})$

$$\text{Let } X_k := x_k, \quad D_k := \frac{1}{i} \frac{\partial}{\partial x_k} \quad \forall k \in [n]$$

$$\mathbf{X} := (X_1 \quad \dots \quad X_n)^\top, \quad \mathbf{D} := (D_1 \quad \dots \quad D_n)^\top$$

Write the semigroup $\exp(-(t - t_0)\mathcal{L})$ in terms of \mathbf{X}, \mathbf{D}

Example (Heat PDE).

$$\mathcal{L} \equiv -\Delta_x, \quad |\mathbf{D}|^2 := \langle \mathbf{D}, \mathbf{D} \rangle = (-i)^2 \langle \nabla_x, \nabla_x \rangle = -\Delta_x$$

Therefore, $H_{\text{heat}}(\mathbf{X}, \mathbf{D}) = \exp\left(- (t - t_0) |\mathbf{D}|^2\right)$

Step 1: PDE \longrightarrow Weyl Operator $H(\mathbf{X}, \mathbf{D})$

Reaction-diffusion PDE:

$$\frac{\partial \hat{\varphi}}{\partial t} = \left(\Delta_{\mathbf{z}} - \frac{1}{2} \mathbf{z}^\top \mathbf{Q} \mathbf{z} \right) \hat{\varphi}, \quad \hat{\varphi}(t_0, \cdot) = \hat{\varphi}_0, \quad \text{w.l.o.g. } \mathbf{Q} \succ \mathbf{0}$$



$$\frac{1}{2} \mathbf{Q} = \mathbf{V}^\top \mathbf{\Lambda} \mathbf{V}$$

$$\mathbf{z} \mapsto \mathbf{x} := \mathbf{V} \mathbf{z}, \quad \hat{v}(t, \mathbf{x}) := \hat{\varphi}(t, \mathbf{z} = \mathbf{V}^\top \mathbf{x})$$

$$\frac{\partial \hat{v}}{\partial t} = \Delta_{\mathbf{x}} \hat{v} - (\mathbf{x}^\top \mathbf{\Lambda} \mathbf{x}) \hat{v} = \sum_{k=1}^n \left(\frac{\partial^2}{\partial x_k^2} - \lambda_k x_k^2 \right) \hat{v}$$

Weyl operator is a composition:

$$Q_{\mathbf{\Lambda}}(\mathbf{X}, \mathbf{D}) := |\mathbf{D}|^2 + \sum_{k=1}^n \lambda_k X_k^2$$

$$H_{\mathbf{\Lambda}}(\mathbf{X}, \mathbf{D}) = \exp(-(t - t_0) Q_{\mathbf{\Lambda}}(\mathbf{X}, \mathbf{D}))$$

Step 2: Weyl Operator $H(X, D) \longrightarrow$ Weyl Symbol $h(x, \xi)$

Derive a PDE IVP for the Weyl symbol

$$\frac{\partial}{\partial t} h_\Lambda(\mathbf{x}, \boldsymbol{\xi}) = - \sum_{j=0}^2 \frac{1}{j!} \{q_\Lambda, h_\Lambda\}_j(\mathbf{x}, \boldsymbol{\xi}), \quad h_\Lambda|_{t=t_0} = 1$$

where $q_\Lambda(\mathbf{x}, \boldsymbol{\xi}) = |\boldsymbol{\xi}|^2 + \sum_{k=1}^n \lambda_k x_k^2$ and the j th order Poisson bracket

$$\{f, g\}_j(\mathbf{x}, \boldsymbol{\xi}) := \left(\frac{1}{2i} \right)^j \left(\sum_{k=1}^n \left(\frac{\partial^2}{\partial y_k \partial \xi_k} - \frac{\partial^2}{\partial x_k \partial \nu_k} \right) \right)^j f(\mathbf{x}, \boldsymbol{\xi}) g(\mathbf{y}, \boldsymbol{\eta}) \Big|_{\mathbf{y}=\mathbf{x}, \boldsymbol{\eta}=\boldsymbol{\xi}} \quad \forall j = 0, 1, 2, \dots$$

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Derive a PDE IVP for the Weyl symbol

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Solution of this PDE IVP:

$$h_{\Lambda}(\mathbf{x}, \boldsymbol{\xi}) = \left(\prod_{k=1}^n \frac{1}{\cosh(\sqrt{\lambda_k}(t - t_0))} \right) \exp \left(- \sum_{k=1}^n \frac{\lambda_k x_k^2 + \xi_k^2}{\sqrt{\lambda_k}} \tanh \left(\sqrt{\lambda_k}(t - t_0) \right) \right)$$

Step 3: Weyl Symbol $h(\mathbf{x}, \boldsymbol{\xi}) \longrightarrow$ Kernel κ

In general,

$$\kappa(t_0, \mathbf{x}, t, \mathbf{y}) = \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} h\left(\frac{\mathbf{x} + \mathbf{y}}{2}, \boldsymbol{\xi}\right) e^{i\langle \mathbf{x} - \mathbf{y}, \boldsymbol{\xi} \rangle} d\boldsymbol{\xi}$$

|
non-unitary inverse Fourier transform of $\boldsymbol{\xi} \mapsto h((\mathbf{x} + \mathbf{y})/2, \boldsymbol{\xi})$

Applying this to our Weyl symbol h_Λ gives

$$\begin{aligned} & \kappa_\Lambda(t_0, \mathbf{x}, t, \mathbf{y}) \\ &= \left(\prod_{k=1}^n \frac{\lambda_k^{1/4}}{\sqrt{2\pi \sinh(2\sqrt{\lambda_k}(t - t_0))}} \right) \\ & \times \exp\left(- \sum_{k=1}^n \frac{\sqrt{\lambda_k}}{2} (x_k^2 + y_k^2) \frac{\cosh(2\sqrt{\lambda_k}(t - t_0))}{\sinh(2\sqrt{\lambda_k}(t - t_0))} \right) \\ & \times \exp\left(\sum_{k=1}^n \sqrt{\lambda_k} x_k y_k \left(\frac{1}{\sinh(2\sqrt{\lambda_k}(t - t_0))} \right) \right) \end{aligned}$$

Summary for the Derived κ_Λ

Recovers the formula in **SCHRÖDINGER BRIDGE WITH QUADRATIC STATE COST IS EXACTLY SOLVABLE***

ALEXIS M.H. TETER[†], WENQING WANG[‡], AND ABHISHEK HALDER[§]

Limit $\frac{1}{2}\mathbf{Q} = \Lambda \downarrow \mathbf{0}$ recovers the heat kernel

Special case $\frac{1}{2}\mathbf{Q} = \Lambda = \mathbf{I}_n$ recovers the Mehler kernel in quantum mechanics

Summary for the Derived κ_Λ (contd.)

Permits generalization: $q(\mathbf{z}) = \frac{1}{2} \mathbf{z}^\top \mathbf{Q} \mathbf{z} + \mathbf{r}^\top \mathbf{z} + s, \quad \mathbf{Q} \succ \mathbf{0}$

$$\begin{aligned} & \kappa_{(\Lambda, \mathbf{r}, s)}(t_0, \mathbf{x}, t, \mathbf{y}) \\ &= \left(\prod_{k=1}^n \frac{\lambda_k^{1/4} \exp(-c_k(t-t_0))}{\sqrt{2\pi \sinh(2\sqrt{\lambda_k}(t-t_0))}} \right) \\ & \times \exp \left(\sum_{k=1}^n -\frac{\sqrt{\lambda_k}}{2} (x_k^2 + y_k^2) \frac{\cosh(2\sqrt{\lambda_k}(t-t_0))}{\sinh(2\sqrt{\lambda_k}(t-t_0))} + \frac{\sqrt{\lambda_k} x_k y_k}{\sinh(2\sqrt{\lambda_k}(t-t_0))} - \frac{\left(\frac{1}{2} \mathbf{r}^\top \mathbf{v}_k (x_k + y_k) + \frac{1}{4\lambda_k} (\mathbf{r}^\top \mathbf{v}_k)^2 \right) \tanh(\sqrt{\lambda_k}(t-t_0))}{\sqrt{\lambda_k}} \right) \end{aligned}$$

$$c_k := \frac{1}{4\lambda_k} (\mathbf{r}^\top \mathbf{v}_k)^2 - \frac{s}{n}, \quad \mathbf{v}_k \text{ is } k\text{th column of } \mathbf{V}^\top$$

Helps in solving the Schrödinger bridge problem with quadratic state cost:

$$\hat{\varphi}(t, \mathbf{z}) = \hat{\nu}(t, \mathbf{x}) \Big|_{\mathbf{x}=\mathbf{V}\mathbf{z}} = \int_{\mathbb{R}^n} \kappa_{(\Lambda, \mathbf{r}, s)}(t_0, \mathbf{V}\mathbf{z}, t, \mathbf{y}) \varphi_0(\mathbf{V}^\top \mathbf{y}) d\mathbf{y}$$

Thank You

Details:



Support:



2112755, 2111688

