## Optimal topological SB

- Schrödinger system characterizes the optimality
- Disintegration of measures: gives us static TSBP (OT formulation)

$$\min D_{KL}(\mathbb{P}_{01} || \mathbb{Q}_{\mathcal{T}_{01}}) \ s.t. \mathbb{P}_0 = \rho_0, \mathbb{P}_1 = \rho_1$$

- . An E-OT with transport cost:  $\|y_1 \Psi_1 y_0 \xi_1\|_{K_{11}^{-1}}^2$
- Lagrange multipliers: gives us a topological Schrödinger system — iterative proportional fitting (cont. Sinkhorn all.) for half-bridge prob
- Stochastic control: tells us how optimal TSB follows a forward-backward SDE system

$$dX_t = \begin{bmatrix} f_t + g_t Z_t \end{bmatrix} dt + g_t dW_t, \quad X_0 \sim \rho_0, \quad Z_t \equiv g_t \nabla \log \varphi_t(X_t)$$

$$dX_t = \begin{bmatrix} f_t - g_t \hat{Z}_t \end{bmatrix} dt + g_t dW_t, \quad X_1 \sim \rho_1, \quad \hat{Z}_t \equiv g_t \nabla \log \hat{\varphi}_t(X_t)$$

Enables learning!!!

• Nonlinear Feynman-Kac formula: gives us the dynamics of  $\log \varphi_t$  and  $\log \hat{\varphi}_t$ 

## Topological SBP

- A topological domain  $\mathcal{T}$ , e.g., a graph or a simplicial complex
- Topological stochastic process:  $X := (X_t)_{t \in [0,1]} : [0,1] \times \mathcal{X} \to \mathbb{R}^n$ 
  - $\mathcal{X}$ : the node/edge space of  $\mathcal{T}$ , with n the dimension
  - X follows some unknown dynamics with distr. law:  $X \sim \mathbb{P} \to X_t \sim \mathbb{P}_t$  (time-marginal)
- Given the initial and final (empirical) signal distr.  $X_0 \sim \rho_0$  and  $X_1 \sim \rho_1$

## Topological Schrödinger Bridge Problem

$$\min D_{KL}(\mathbb{P}||\mathbb{Q}_{\mathcal{T}}) \ s.t. \mathbb{P}_0 = \rho_0, \mathbb{P}_1 = \rho_1$$