

Adapted Wasserstein distance between the laws of SDEs

Julio Backhoff-Veraguas (Universität Wien), Sigrid Källblad (KTH Stockholm), Benjamin A. Robinson (Universität Wien)

ben.robinson@univie.ac.at



Goal

For one-dimensional SDEs

$$\begin{aligned} dX_t &= b(X_t)dt + \sigma(X_t)dW_t && \rightsquigarrow && \mu = \text{Law}((X_t)_{t \in [0,1]}) \\ d\bar{X}_t &= \bar{b}(\bar{X}_t)dt + \bar{\sigma}(\bar{X}_t)d\bar{W}_t && \rightsquigarrow && \nu = \text{Law}((\bar{X}_t)_{t \in [0,1]}) \end{aligned}$$

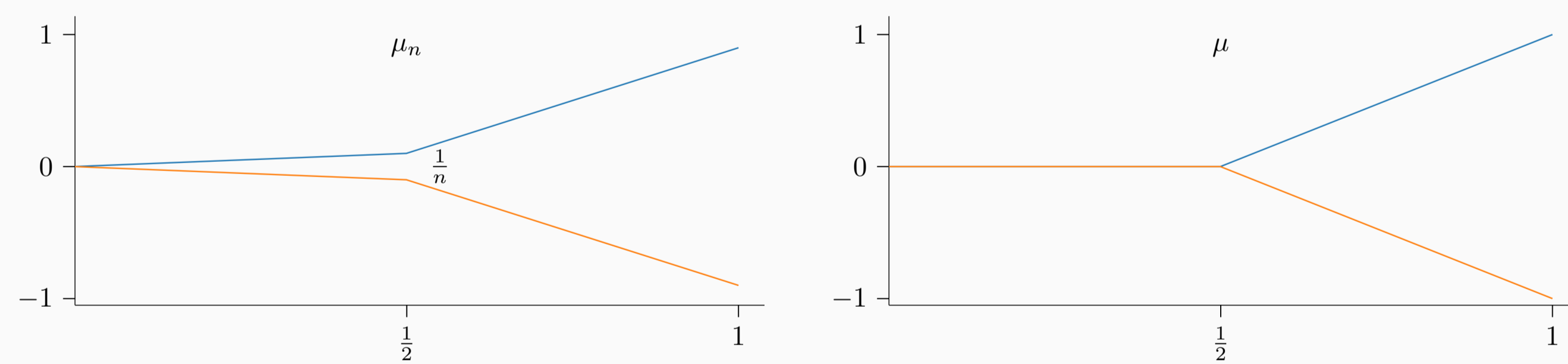
- > Find a **suitable distance** between μ and ν
- > Show **convergence** of discrete to continuous distances

Adapted Wasserstein

Canonical distance between probability measures is the **Wasserstein distance**

$$\mathcal{W}_p^p(\mu, \nu) = \inf_{\pi \in \text{Cpl}(\mu, \nu)} \int_{\Omega} \int_0^1 |\omega_s - \bar{\omega}_s|^p dt d\pi(\omega, \bar{\omega})$$

Example:



- > Optimal coupling is to match the **blue paths** and the **orange paths**, and so $\mathcal{W}_p(\mu_n, \mu) \rightarrow 0$
- > A sequence of **deterministic processes** converging to a **martingale** on $[0, 1/2]$
- > This is **not** a good distance between stochastic processes

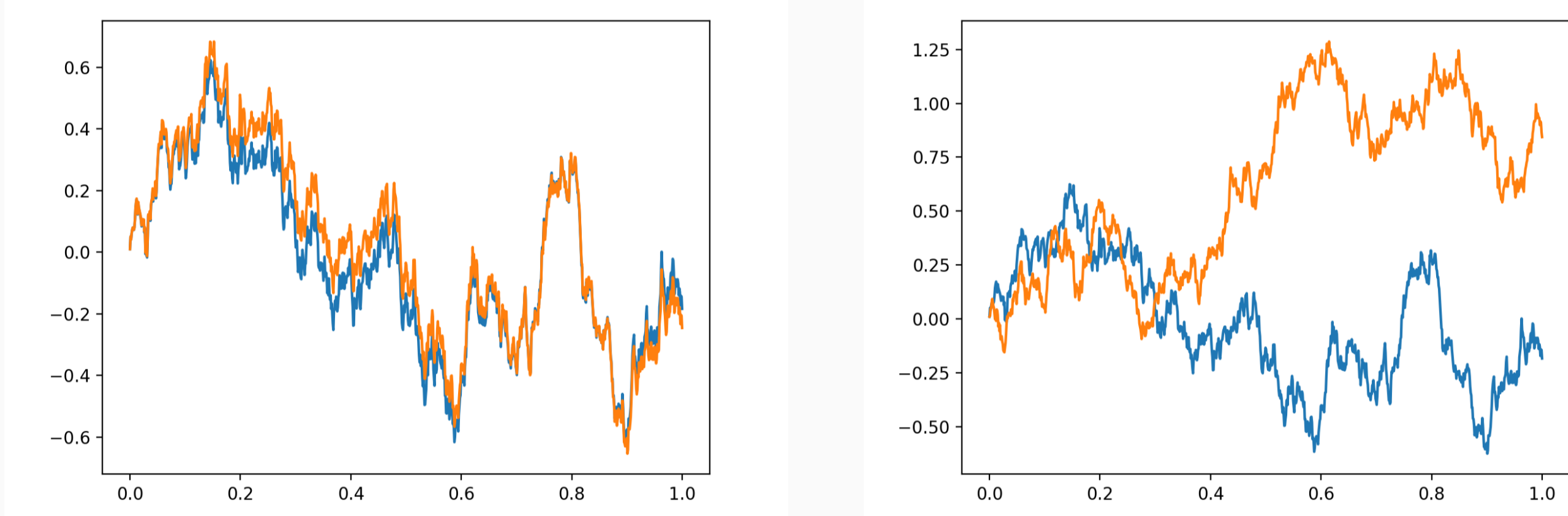
Define the **adapted Wasserstein distance** $\mathcal{AW}_p(\mu, \nu)$ by replacing $\text{Cpl}(\mu, \nu)$ with

$$\text{Cpl}_{bc}(\mu, \nu) = \{\pi \in \text{Cpl}(\mu, \nu) : \pi \text{ is bi-causal}\}$$

- > $\pi = \text{Law}(X, \bar{X})$ is **bi-causal** if given the past of X , the future of \bar{X} is independent of X , and vice-versa
- > The optimal coupling above is no longer admissible
- > $\mathcal{AW}_2(\mu_n, \mu) \approx \frac{1}{2}$ for all n large

Results

Theorem. Optimising over **bi-causal couplings** is equivalent to optimising over **correlations** between W, \bar{W} .



Synchronous coupling: $W = \bar{W}$

Product coupling: W, \bar{W} independent

Theorem. Suppose that the coefficients are **continuous** with **linear growth** and that **pathwise uniqueness** holds.

Then the **synchronous coupling** attains $\mathcal{AW}_p(\mu, \nu)$.

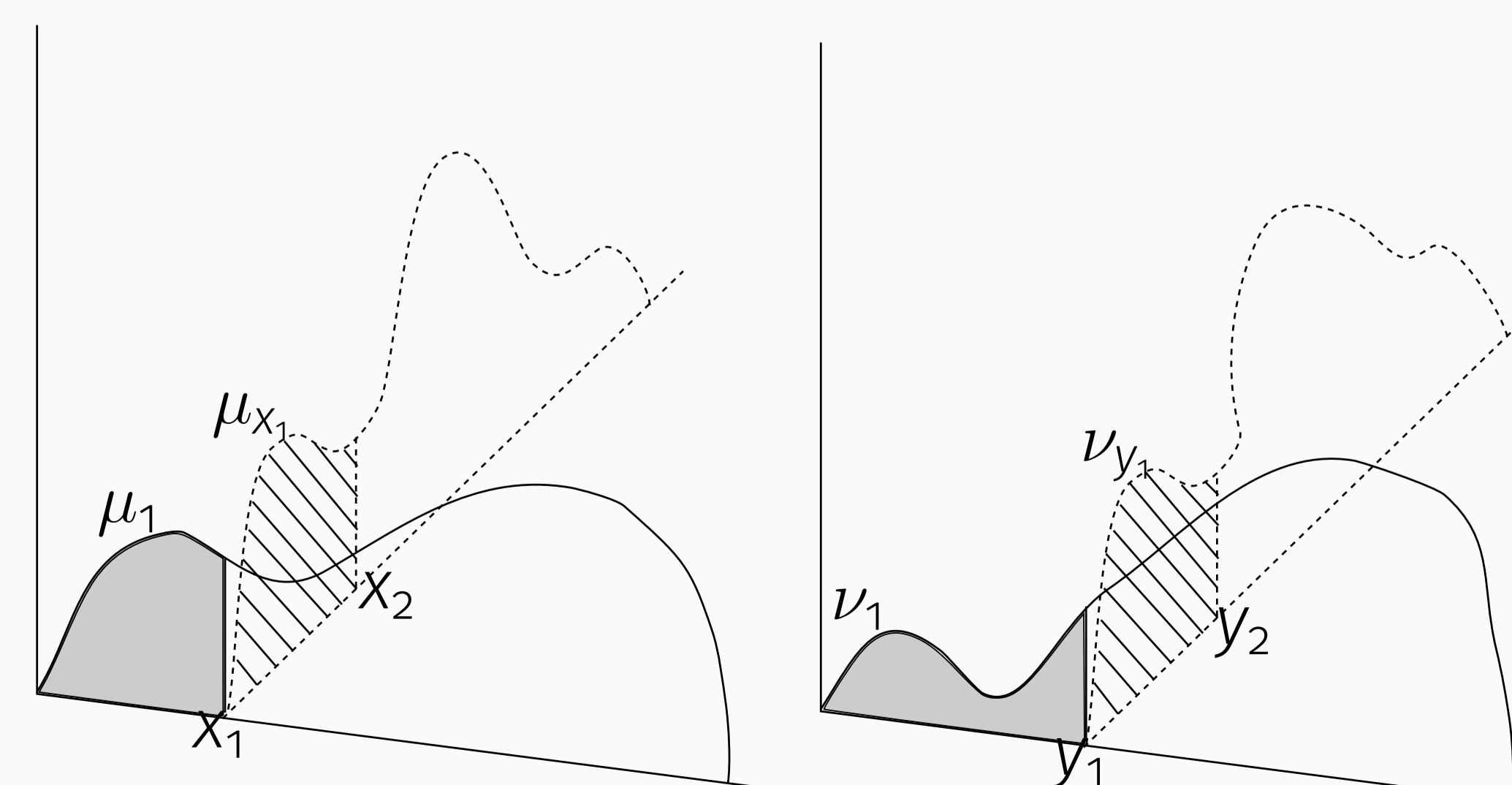
Knothe–Rosenblatt

> For μ, ν on \mathbb{R}^n , $U_1, \dots, U_n \stackrel{iid}{\sim} \mathcal{U}[0, 1]$, $X_1 = F_{\mu_1}^{-1}(U_1)$, $Y_1 = F_{\nu_1}^{-1}(U_1)$,

$$X_k = F_{\mu_{X_1, \dots, X_{k-1}}}^{-1}(U_k), \quad Y_k = F_{\nu_{Y_1, \dots, Y_{k-1}}}^{-1}(U_k)$$

> The **Knothe–Rosenblatt rearrangement** between μ, ν is

$$\pi_{\mu, \nu}^{KR} = \text{Law}(X_1, \dots, X_n, Y_1, \dots, Y_n)$$



Knothe–Rosenblatt rearrangement for $n = 2$

Discretisation

Monotone Euler–Maruyama scheme: $n \in \mathbb{N}, h = 1/n$

$$\begin{aligned} X_0^h &= X_0, \\ X_t^h &= X_{kh}^h + b(X_{kh}^h)(t - kh) + \sigma(X_{kh}^h)(W_t^h - W_{kh}^h), \quad t \in (kh, (k+1)h] \end{aligned}$$

- > The process W^h is a martingale with **Brownian increments stopped** when hitting some barrier

Theorem. Convergence in L^p holds for all $p \geq 1$ and, for $h \ll 1$,

$$\mathbb{E} \left[\sup_{t \in [0,1]} |X_t^h - X_t|^2 \right] \leq Ch.$$

Theorem (Rüschendorf '85). For μ_n, ν_n on \mathbb{R}^n **stochastically co-monotone**, the **Knothe–Rosenblatt** rearrangement attains $\mathcal{AW}_p(\mu_n, \nu_n)$.

- > If b, σ are Lipschitz and $h \ll 1$, then $\mu^h = \text{Law}((X_{kh}^h)_{k=1, \dots, n})$ is **stochastically increasing**
- > Therefore

$$\mathcal{AW}_p(\mu^h, \nu^h) \rightarrow \mathcal{AW}_p(\mu, \nu)$$

- > The **Knothe–Rosenblatt** rearrangement attains the *LHS* for each h and the **synchronous** coupling attains the *RHS*
- > By a **stability** argument, we arrive at the main theorem

Reference

[1] J. Backhoff-Veraguas, S. Källblad, and B. A. Robinson. Adapted Wasserstein distance between the laws of SDEs. *arXiv:2209.03243 [math]*, Sep. 2022.



Supported by Austrian Science Fund (FWF) projects Y782-N25 and P35519.