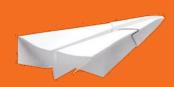


# Parareal computation of SDEs with time-scale separation



Frédéric Legoll<sup>1</sup>, Tony Lelièvre<sup>1</sup>, Keith Myerscough<sup>2</sup> and Giovanni Samaey<sup>2</sup> <sup>1</sup>Ecole des Ponts ParisTech, <sup>2</sup>KU Leuven

#### In a nutshell

Goal Simulate slow-fast SDEs over long time, quickly

Model Slow-fast system of SDEs, and a macroscopic model taken from the "fast" limit

Method Parallel-in-time algorithm that iteratively improves the macroscopic result

Result Reduction in wall clock time

Bonus Lower variance than full microscopic model

# Microscopic model

Slow-fast system of coupled SDEs

$$dX = (-X_t^3 + X_t + Y_t^2) dt + \sqrt{\frac{2}{\beta}} dB_t^{(x)}$$

$$dY = \frac{1}{\epsilon} (X_t - Y_t) dt + \sqrt{\frac{2}{\beta \epsilon}} dB_t^{(y)}.$$

- Modeled as an ensemble  $\mathcal{X}_t$  of particles with positions  $(X_t^p, Y_t^p)$  and weight  $W^p$
- lacktriangle Time integrator: a Lie-Trotter splitting, updating  $X_t$ first, then  $Y_t$
- Validation: deterministic solution given by the Fokker-Planck equation, akin to the macroscopic model

## Macroscopic model

Only slow variable, assume the fast  $Y_t$  is equilibrated and use only the expected value of the term  $Y_t^2$ 

$$dZ = (-Z_t^3 + Z_t + Z_t^2 + \frac{1}{\beta}) dt + \sqrt{\frac{2}{\beta}} dB^{(x)},$$

or in potential form

$$\mathrm{d}Z = -\partial_z V_{\mathsf{eff}}(Z_t) \,\mathrm{d}t + \sqrt{\frac{2}{\beta}} \,\mathrm{d}B^{(x)}.$$

■ The associated Fokker-Planck equation reads

$$\partial_t \rho(z) = \partial_z \left( \rho(z) \partial_z V_{\text{eff}} \right) + \frac{1}{\beta} \partial_{zz} \rho(z).$$

■ The macroscopic state is represented by integral quantities over a regular grid

### Coupling

**Restriction** ( $\mathcal{R}$ , from micro to macro) sum the weights of all particles in each bin

**Matching** ( $\mathcal{M}$ , from macro to micro) reweight particles from a known microstate  $(\bar{X}^p, \bar{Y}^p, \bar{W}^p)$ 

**Resampling** ( $\mathcal{M}^*$ , optional) retrieve an ensemble with all particles equal in weight

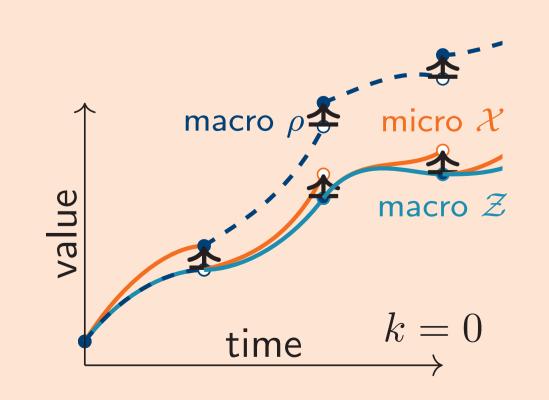
The parareal algorithm

**Iteratively improves** the macroscopic propagator by computing the discrepancies between the macroscopic and the microscopic models in parallel

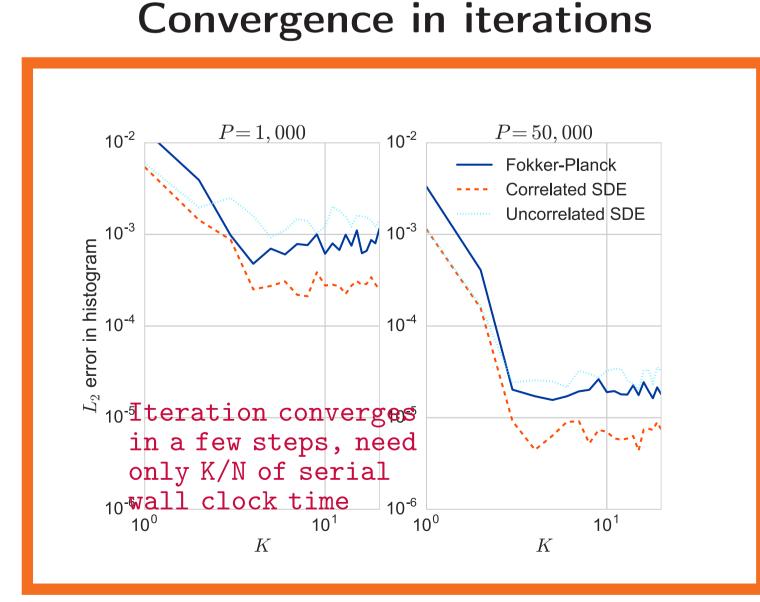
Parallel use of the microscopic propagator gives a reduction in wall clock time if there are fewer iterations needed than time steps

Variance of the stochastic microscopic propagator dominates the error (see figures  $\rightarrow$ )

Reduction in variance by using a particle propagator with correlated noise for the macroscopic model in computing the discrepancies between the models



Parameters used below:  $\beta = 5, \epsilon = 0.1, K = N = 20, \Delta t = 0.025, \delta t = 1.0 \times 10^{-4}$ 



### Convergence in number of particles

