

Slowly, then all at once: Uncovering the dynamics of a catastrophe

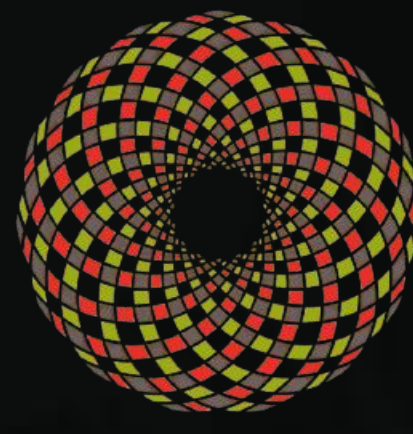
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1. The Question

What does the desertification of the Sahara, 6000 years ago, and the Great Recession of 2008 have in common?

They both unfolded in two phases: first **gradually** and then **suddenly**!

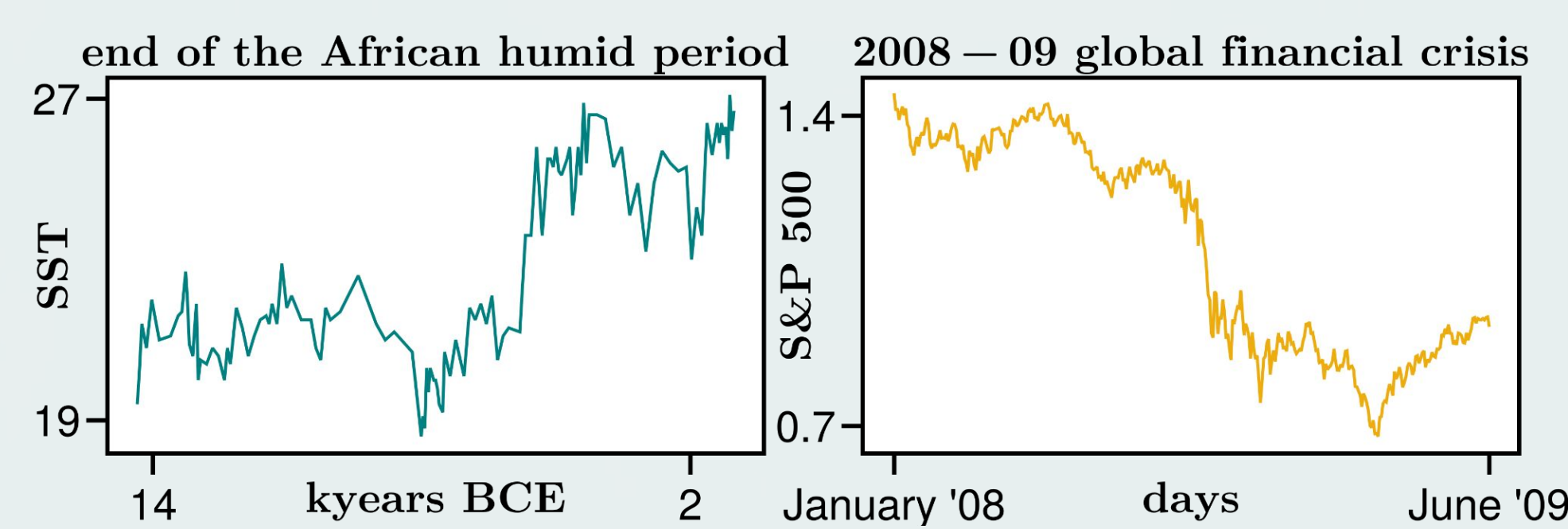


Fig. 1: Examples of real-world catastrophic transitions in dynamical systems.

Is this a generic trait of catastrophic collapses in nature?
Are there early-warning signs that can forecast these events?

We consider a minimal working model that can reproduce tipping phenomena between alternative stable states

$$\begin{aligned} dx &= f(x, \mu) dt + \sigma(x) dW, \\ d\mu &= \varepsilon(t) dt, \end{aligned} \quad (1)$$

- Cubic drift: $f(x, \mu) = -\mu - 2x + 3x^2 - 0.8x^3$
- Additive stochastic diffusion: $\sigma(x) = \sigma > 0$
- Linear ramp of the bifurcation parameter μ : $\varepsilon(t) = \varepsilon > 0$
- Pair of saddle-node bifurcations: $\mu \approx -0.371$ and $\mu \approx 1.621$
- Analogy with Langevin dynamics: overdamped particle in a quartic potential landscape $V(x, \mu)$ s.t. $f(x, \mu) = -\partial_x V$

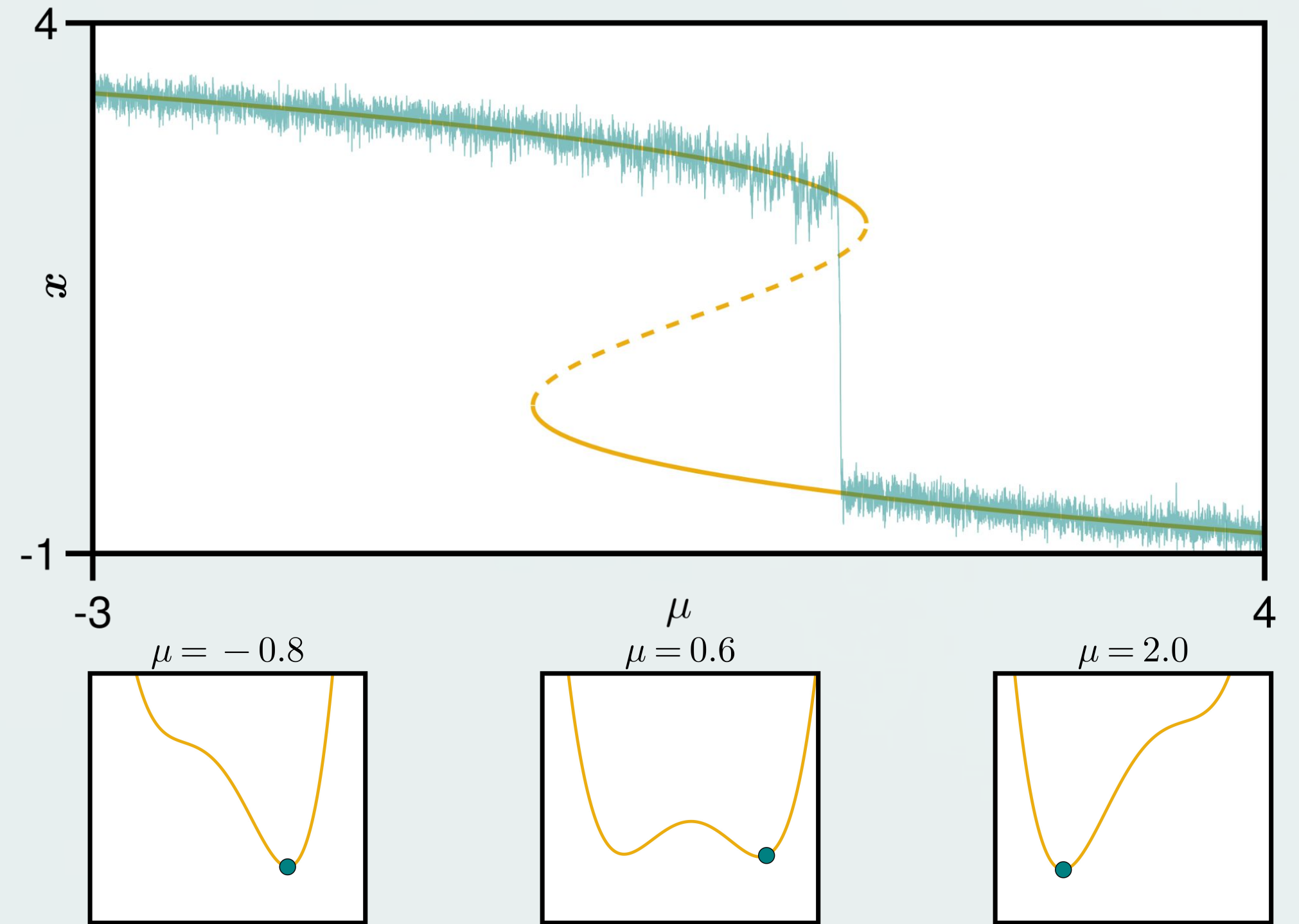


Fig. 2: Timeseries of a solution of (1) and its potential ($\sigma = 0.3$, $\varepsilon = 0.01$).

2. The Problem

Tipping points

Mechanisms to escape the basin of attraction:

- A bifurcation is reached (B-tipping)
- The parameter drift is too fast (R-tipping)
- The Brownian noise dominates (N-tipping)

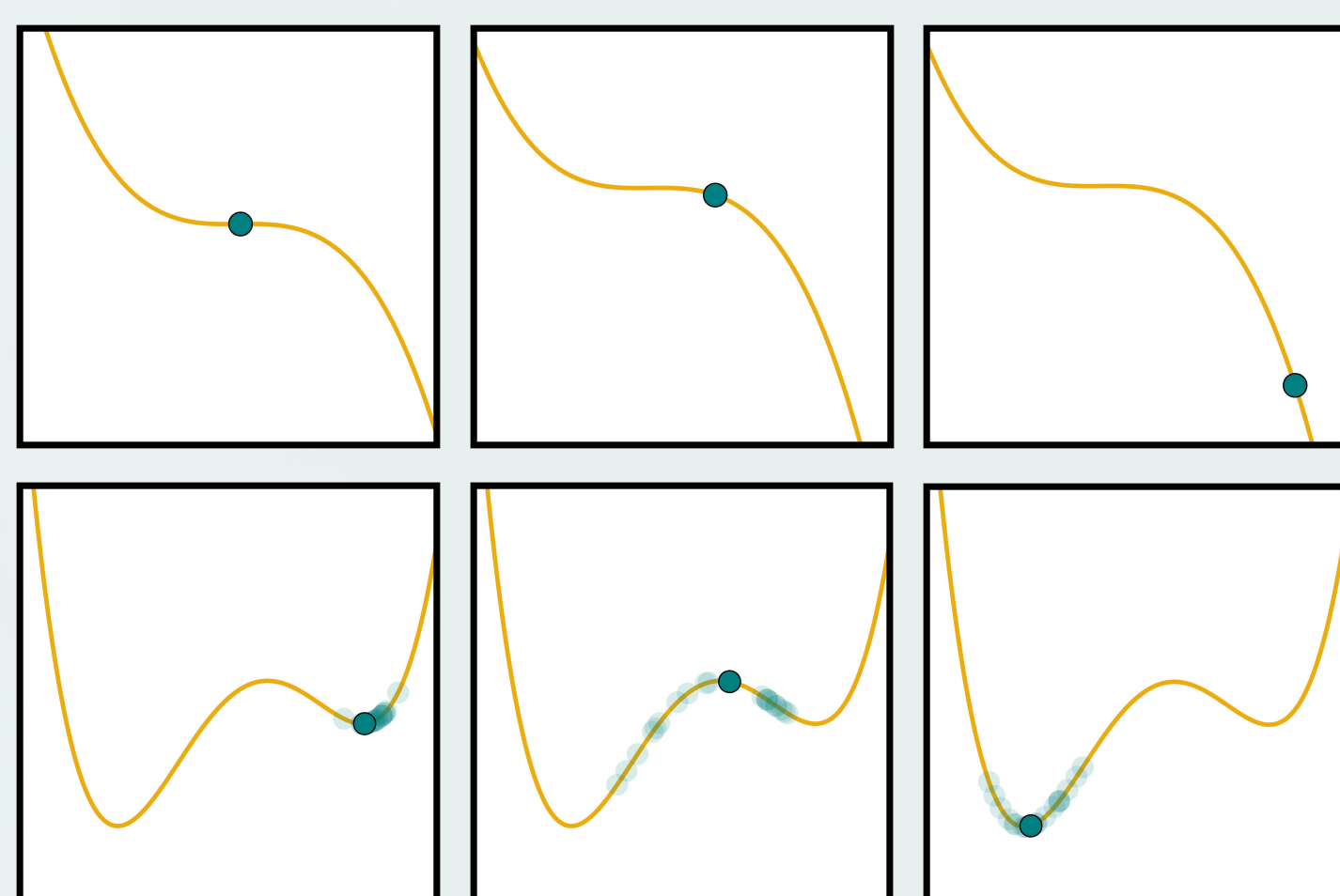


Fig. 3: Tipping mechanism visualised: R-tipping (top) in a deterministic ($\sigma = 0$) cubic [1]; N-tipping (bottom) in a stationary ($\varepsilon = 0$) quartic; B-tipping shown in Fig. 2.

Critical slowing down

As the bifurcation is approached the stable equilibria become increasingly weaker in attracting nearby solutions.

Increases in variance precede B-tipping [2].

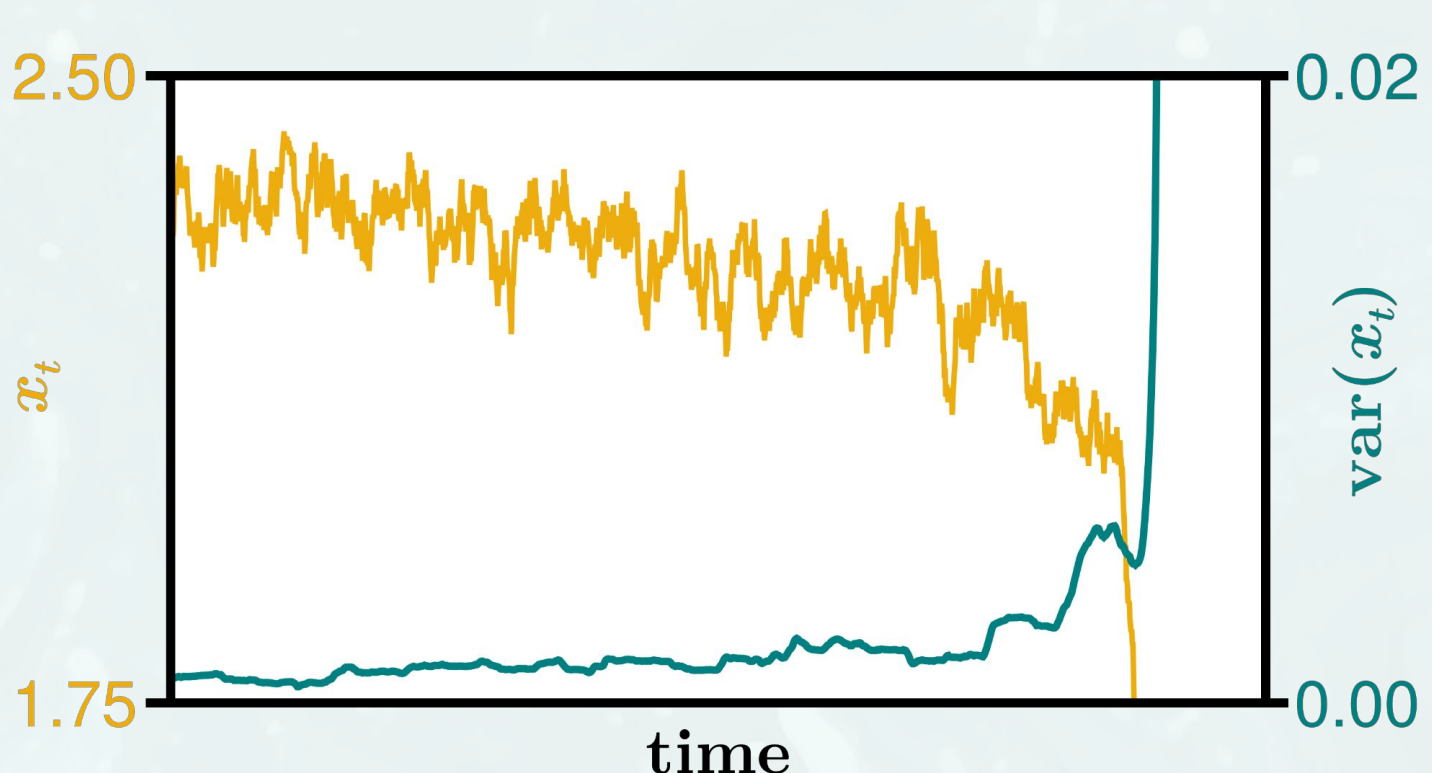


Fig. 4: B-tipping of (1) in a fast-slow ($\varepsilon = 10^{-3}$), stochastic ($\sigma = 5 \cdot 10^{-2}$) regime (yellow) and its variance (teal) computed on a sliding window.

3. The Method

An alternative measure of imminent catastrophe

Far from the bifurcation and at equilibrium (1) is topologically equivalent to an Ornstein-Uhlenbeck process $\dot{x} = -\theta(x - a) + \sigma dW$, where $\theta = \partial_x V(a)$. A solution starting at $x = a$ will thus tip, with some probability $p_{\text{esc}} > 0$, over the local maxima $x = b$ (boundary of the basin of attraction) of the scalar potential.

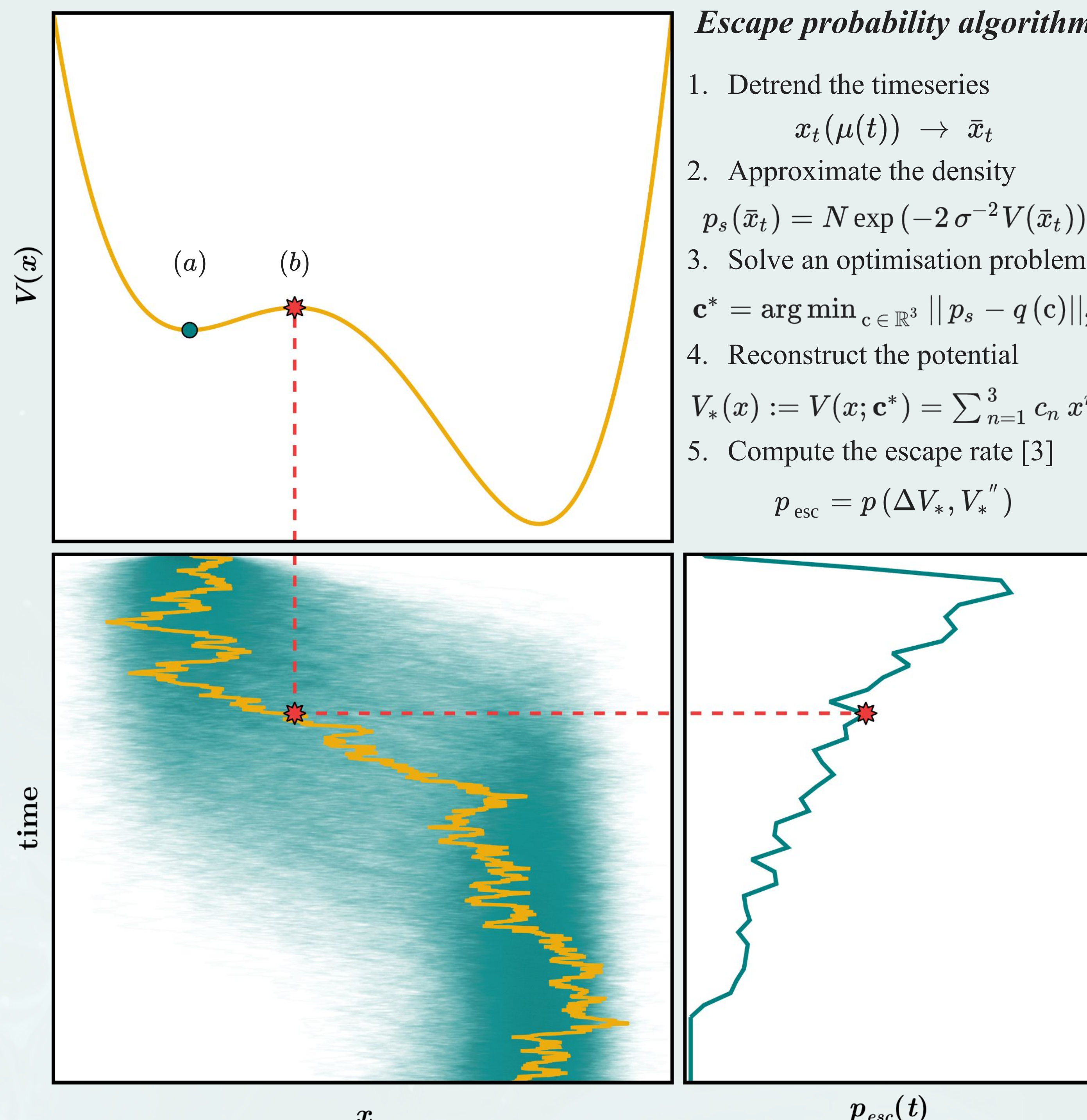


Fig. 5: Stationary ($\varepsilon = 0$) escape time distribution (bottom right) of an ensemble of $N=3000$ overdamped particles (bottom left) out of the basin of attraction of the potential (top) at fixed parameter value $\mu=0$. The histogram $p_{\text{esc}}(t)$ approximates an inverse Gaussian distribution at noise level $\sigma = 1.0$.

Escape probability algorithm

1. Detrend the timeseries $x_t(\mu(t)) \rightarrow \bar{x}_t$
2. Approximate the density $p_s(\bar{x}_t) = N \exp(-2\sigma^{-2}V(\bar{x}_t))$
3. Solve an optimisation problem $\mathbf{c}^* = \arg \min_{\mathbf{c} \in \mathbb{R}^3} \|p_s - q(\mathbf{c})\|_2$
4. Reconstruct the potential $V_*(x) := V(x; \mathbf{c}^*) = \sum_{n=1}^3 c_n x^n$
5. Compute the escape rate [3] $p_{\text{esc}} = p(\Delta V_*, V_*'')$

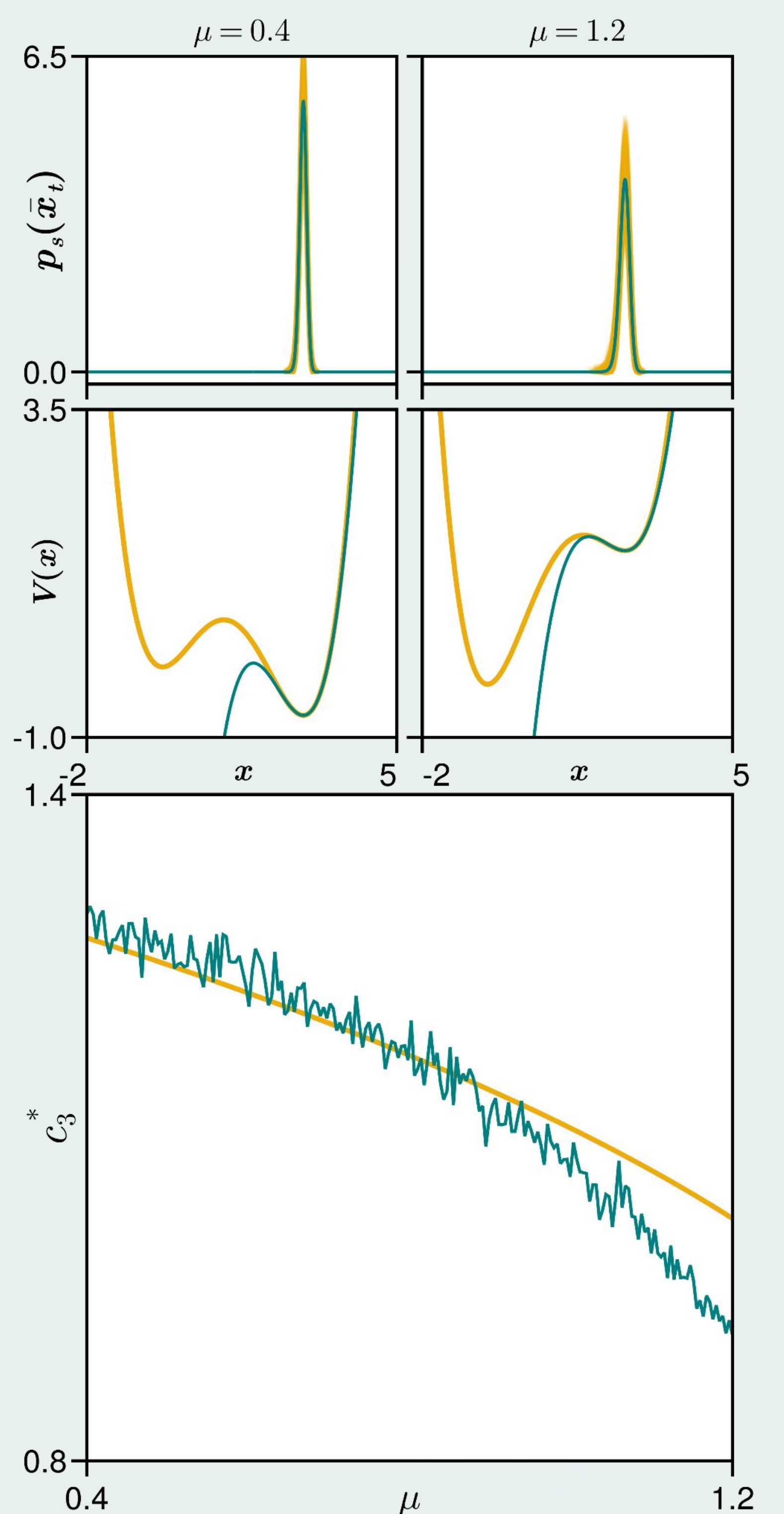


Fig. 6: Steps of the escape probability algorithm illustrated: parameter sweep simulation ($\varepsilon = 0$, $\sigma = 0.2$) of an ensemble of $N=3000$ particles evolving according to (1). (Top) Ensemble histograms of the stationary density (yellow) and the solution of the (non-linear) optimisation problem (teal). (Middle) True potential function $V(x)$ (yellow) and its local reconstruction $V_*(x)$ (teal) from the solution of the optimisation problem. (Bottom) Cubic term of the Taylor expansion of the potential around $x = a$ (yellow) and its approximation (teal).

4. The Results

- The error improves as we approach the bifurcation, i.e. $\|V - V_*\|_2 \rightarrow 0$ as $\mu \rightarrow \mu_c$
- We find a probabilistic early-warning signal for (1) in stationary ($\varepsilon = 0$) and low-noise regimes ($\sigma \sim 10^{-1}$)
- **What's next?** Generalisation of the method to systems in quasi-stationary ($0 < \varepsilon \ll 1$) regimes [4]

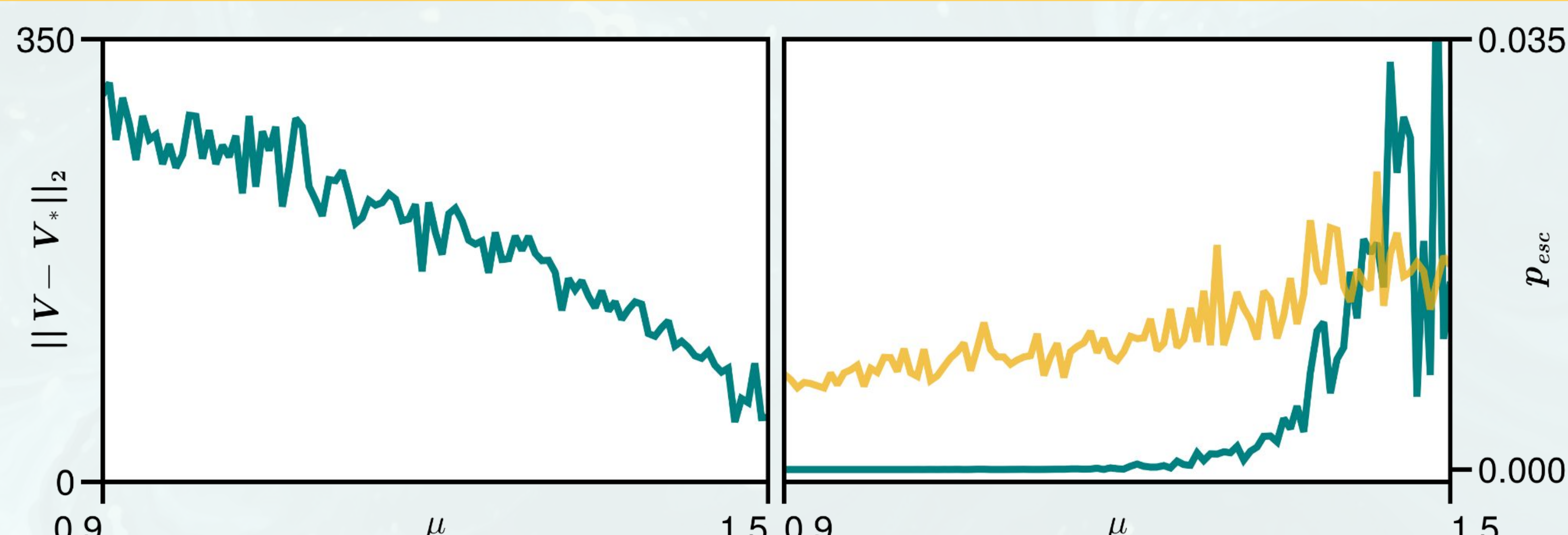


Fig. 7: (Left) Approximation error of the reconstruction of the potential landscape $V(x)$ for the ensemble simulation performed in Fig. 6. (Right) Probability-to-tip early-warning signal (teal), computed out of the reconstructed potential $V_*(x)$, compared to the stationary variance (yellow) for an indicative timeseries in the ensemble.

5. The Literature

- [1] Ashwin, P. et al., Philos. Trans. R. Soc. A, Vol. 370 (2012)
- [2] Kuehn, C., Physica D, Vol. 240 (2011)
- [3] Kramers, H., Physica, Vol. 7 (1940)
- [4] Donovan, G., Physica A, Vol. 646 (2024)

6. Bella ciao! 🙌 😊

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