

# Recursive Computation of Invariant Distribution of Feller Process: Applications and Numerical experiments

Ruikai CHEN - Zian CHEN - Tiena SORO

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- 2 Application to the discretization scheme of Ito diffusion
- 3 Numerical Experiments
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# General Framework

Feller process

numerical scheme

$$(X_t)_t \longrightarrow (\bar{X}_{\Gamma_n})_n$$



$$\nu \xleftarrow[\text{convergence}]{\text{a.s. weak}} \nu_n^\eta(dx) = \frac{1}{\sum \eta_k} \sum_{k=1}^n \eta_k \delta_{\bar{X}_{\Gamma_n}}(dx)$$

invariant measure

empirical invariant measure

## Notations

- $(X_t)_t$  is a Feller process with Feller semi-group  $(P_t)_t$
- $(\bar{X}_{\Gamma_n})_n$  is a Markov approximation of  $(X_t)_t$
- $V$  Lyapunov function
- $\psi$  test function,  $\phi$  control function
- $A$  is the infinitesimal generator :  $Af = \lim_{t \rightarrow 0} \frac{P_t f - f}{t}$ , where  $f$  is a continuous function
- $\tilde{A}_\gamma$  is the pseudo-generator of  $(\bar{X}_{\Gamma_n})$  :

$$\tilde{A}_{\gamma_{n+1}} f(x) = \frac{1}{\gamma_{n+1}} \mathbb{E}[f(\bar{X}_{\Gamma_{n+1}}) - f(\bar{X}_{\Gamma_n}) \mid \bar{X}_{\Gamma_n} = x]$$

$\mathcal{D}(A)_0$  set of  $f$  where  $\tilde{A}_\gamma f$  is well defined.

# Main Assumptions

- **Mean-Reverting Recursive Control :**

$$\forall x \in E, \sup_{\gamma \in (0, \bar{\gamma}]} \tilde{A}_\gamma \psi \circ V(x) \leq \frac{\psi \circ V(x)}{V(x)} (\beta - \alpha \phi \circ V(x))$$

- **Infinitesimal Generator Approximation :**

$$\forall \gamma \in (0, \bar{\gamma}], \forall f \in \mathcal{D}(A), \forall x \in E, |\tilde{A}_\gamma f(x) - Af(x)| \leq \Lambda_f(x, \gamma)$$

- **Growth Control :**  $\forall f \in F$

$$\mathbb{E} [ |f(\bar{X}_{\Gamma_{n+1}}) - \mathbb{E}[f(\bar{X}_{\Gamma_{n+1}}) | \bar{X}_{\Gamma_n}]|^\rho | \bar{X}_{\Gamma_n} ] \leq C_f \epsilon_I(\gamma_{n+1}) g(\bar{X}_{\Gamma_n})$$

- **Step Weight assumptions**

# Main results

## Theorem (Almost sure tightness, Identification of the limit)

Suppose  $s \geq 1$  and that certain assumptions hold, we have

$$\mathbb{P}\text{-a.s. } \sup_{n \in \mathbb{N}^*} \nu_n^\eta \left( \tilde{V}_{\psi, \phi, s} \right) < +\infty,$$

with  $\tilde{V}_{\psi, \phi, s} = \frac{\phi \circ V(x) \psi \circ V(x)^{1/s}}{V(x)}$ . Therefore,  $(\nu_n^\eta)_{n \in \mathbb{N}^*}$  is a.s. tight, and

$$\mathbb{P}\text{-a.s. } \forall f \in \mathcal{D}(A), \lim_{n \rightarrow +\infty} \nu_n^\eta(Af) = 0.$$

It follows that,  $\mathbb{P}$ -a.s., every weak limiting distribution  $\nu_\infty^\eta$  is a invariant measure. And if the invariant measure  $\nu$  is unique, then for all  $f$  continuous s.t.  $f = o(\tilde{V}_{\psi, \phi, s})$ , we have a.s.  $\lim \nu_n^\eta(f) = \nu(f)$ .

# Ito diffusion

We consider the solution of the  $d$ -dimensional stochastic equation

$$X_t = x + \int_0^t b(X_s) ds + \int_0^t \sigma(X_s) dW_s,$$

where  $b : \mathbb{R}^d \rightarrow \mathbb{R}^d$  and  $\sigma : \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$ .

Assume a Lipschitz condition holds for  $b$  and  $\sigma$  :

$$\exists L, \forall x, y \in \mathbb{R}^d, \|b(x) - b(y)\| + \|\sigma(x) - \sigma(y)\|_F \leq L\|x - y\|,$$

which ensures the existence and uniqueness of the solution  $(X_t)_{t \geq 0}$ ,  $(X_t)_{t \geq 0}$  is a Feller process with Feller semi-group

$$P_t f : x \mapsto \mathbb{E}^x[f(X_t)].$$

**Notation** :  $\|\sigma\|_F := \text{Tr}(\sigma\sigma^*)^{1/2}$  Frobenius norm.

# Discretization scheme of Ito diffusion

When the exact solution  $(X_t)_{t \geq 0}$  is known, we can build its empirical invariant measures with a weight sequence  $\eta := (\eta_n)_{n \in \mathbb{N}^*}$  and a times grid  $\Gamma_n = \sum_{k=1}^n \gamma_k$  :

$$\nu_n^\eta(dx) = \frac{1}{\sum_{k=1}^n \eta_k} \sum_{k=1}^n \eta_k \delta_{X_{\Gamma_{k-1}}}(dx).$$

We will show that, under certain conditions, every weak limiting distribution  $\nu$  of  $(\nu_n^\eta)_{n \in \mathbb{N}^*}$  is an invariant distribution of  $(X_t)$ .



## Some crucial hypotheses

Let  $V : \mathbb{R}^d \rightarrow [v_*, \infty)$ , ( $v_* > 0$ ) be a Lyapunov function,  $\lim_{x \rightarrow \infty} V(x) = +\infty$ , and is essentially quadratic in the sense

$$\|\nabla V\|^2 \leq C_V V, \quad \|D^2 V\|_\infty < +\infty.$$

Let  $\phi : [v_*, \infty) \rightarrow \mathbb{R}^+$  be a continuous function. Let  $p > 0$  and define  $\psi_p(y) = y^p$ .

- Mean-reverting assumptions

$$\forall x, \underbrace{\langle \nabla V(x), b(x) \rangle + \frac{1}{2} C(V, p) \text{Tr}(\sigma \sigma^*(x))}_{\approx AV(x)} \leq \beta - \alpha \phi \circ V(x)$$

where  $AV(x) := \lim_{t \rightarrow 0} \frac{P_t V(x) - V(x)}{t} = \lim_{t \rightarrow 0} \frac{1}{t} (\mathbb{E}^x[V(X_t)] - V(x))$ .

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- **Coefficient control**

$$\begin{aligned} \forall x, \quad \|\nabla V(x)\|(\|b(x)\|^2 + \|\sigma(x)\|_F^2)^{\frac{1}{2}} + \|b(x)\|^2 + \|\sigma(x)\|_F^2 \\ \leq C\phi \circ V(x). \end{aligned}$$

- **Assumptions on time steps and weights**

# Key properties of Ito diffusion

## Proposition (Burkholder inequality)

For all  $p \geq 2$ , there exists  $C_p > 0$  s.t.

$$\mathbb{E} \left[ \sup_{s \leq T} \left\| \int_0^s \sigma(X_u) dW_u \right\|^p \right] \leq C_p \mathbb{E} \left[ \left( \int_0^T \|\sigma(X_u)\|_F^2 du \right)^{\frac{p}{2}} \right]$$

## Lemma

Suppose that  $p \geq 1$ , we have

$$\begin{aligned} \mathbb{E} \left[ \left( \int_{\Gamma_n}^{\Gamma_{n+1}} \|b(X_s)\|^2 ds \right)^p + \left( \int_{\Gamma_n}^{\Gamma_{n+1}} \|\sigma(X_s)\|_F^2 ds \right)^p \mid X_{\Gamma_n} \right] \\ \leq K \gamma_{n+1}^p (\|b(X_{\Gamma_n})\|^{2p} + \|\sigma(X_{\Gamma_n})\|_F^{2p}). \end{aligned}$$

## Key properties of Ito diffusion

## Corollary

Suppose that  $p > 0$ , then  $\exists n_0, M_p$  s.t. for all  $n \geq n_0$ ,

$$\mathbb{E}[\|X_{\Gamma_{n+1}} - X_{\Gamma_n}\|^{2p} \mid X_{\Gamma_n}] \leq M_p \gamma_{n+1}^p (\|b(X_{\Gamma_n})\|^2 + \|\sigma(X_{\Gamma_n})\|_F^2)^p.$$

A powerful tool for the control of quantities in the form

$$\mathbb{E}[f(X_{\Gamma_{n+1}}) - f(X_{\Gamma_n}) \mid X_{\Gamma_n}] \text{ and } \mathbb{E}\left[\int_{\Gamma_n}^{\Gamma_{n+1}} f(X_s) ds \mid X_{\Gamma_n}\right] - \gamma_{n+1} f(X_{\Gamma_n}).$$

**Example (Recursive Control) :**

$$\mathbb{E}[\psi \circ V(X_{\Gamma_{n+1}}) - \psi \circ V(X_{\Gamma_n}) \mid X_{\Gamma_n}] \leq \frac{\gamma_{n+1} \psi \circ V(X_{\Gamma_n})}{V(X_{\Gamma_n})} (\beta - \alpha \phi \circ V(X_{\Gamma_n})).$$

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## Example (Infinitesimal Estimation)

$$\forall \gamma \in (0, \bar{\gamma}], \forall f \in \mathcal{C}_K^3(\mathbb{R}^d), \forall x \in \mathbb{R}^d, \|\tilde{A}_\gamma f(x) - Af(x)\| \leq \Lambda_f(x, \gamma),$$

with

$$\tilde{A}_{\gamma_n} f(x) = \frac{1}{\gamma_n} \mathbb{E}[f(X_{\Gamma_n}) - f(X_{\Gamma_{n-1}}) \mid X_{\Gamma_{n-1}} = x].$$

$$Af(x) = \langle b(x), \nabla f(x) \rangle + \frac{1}{2} \text{Tr}(\sigma \sigma^*(x) \nabla^2 f(x))$$

# Basic Setting

The Ornstein–Uhlenbeck process under our consideration is

$$dX_t = 2(1 - X_t)dt + 2dW_t.$$

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About the time grid and the weight sequence, we take here both  $(\gamma_k)_k$  in order to simplify the argument.



# Euler Scheme

A general stochastic differential equation

$$dX_t = b(t, X_t)dt + \sigma(t, X_t)dW_t$$

can be discretized as

$$\overline{X_{\Gamma_{i+1}}} - \overline{X_{\Gamma_i}} = b(\Gamma_i, \overline{X_{\Gamma_i}})\gamma_{i+1} + \sigma(\Gamma_i, \overline{X_{\Gamma_i}})\sqrt{\gamma_{i+1}}Z_i,$$

where  $(Z_i)_{i \geq 0}$  are standard Gaussian.

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where  $(Z_i)_{i \geq 0}$  are standard Gaussian.

Hence, one can have

$$\nu_n^\gamma(f) := \frac{1}{\Gamma_n} \sum_{k=1}^n \gamma_k f(\overline{X_{\Gamma_{k-1}}}),$$

where we take  $\gamma_k = k^{-\alpha}$  and  $f(x) = x$  to show the convergence.

# Convergence Figures

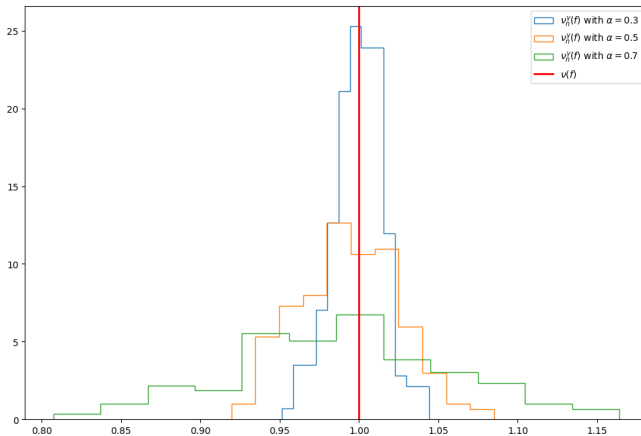


Figure – Histogram of  $\nu_n^\gamma(f)$  for different  $\alpha$

## Convergence Figures

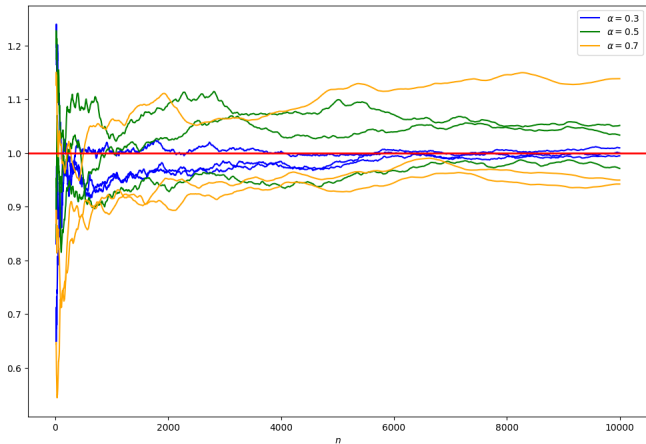


Figure – Values of  $\nu_n^\gamma(f)$  for different  $\alpha$

## Rate of Convergence

In order to...

- reduce the constant when  $f$  has a large growth rate and  $\alpha$  is small, one can add  $n_0$  to  $\gamma_k$ , i.e.

$$\gamma_k = (k + n_0)^{-\alpha};$$

- increase the rate of convergence, theoretically, for  $f$  of form  $Ag$ , the best  $\alpha$  is  $1/3$ .

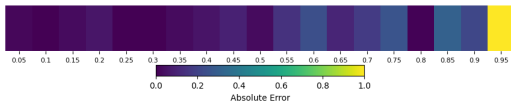


Figure – Normalized absolute errors of  $\nu_n^\gamma(f)$  for different  $\alpha$

# Recursive Formula of O-U process

Since the Ornstein-Uhlenbeck process has an exact solution, we can verify the result from the previous section :

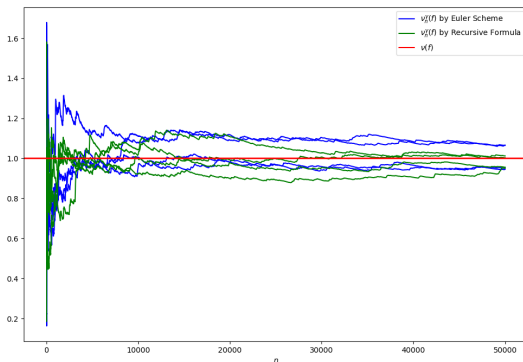


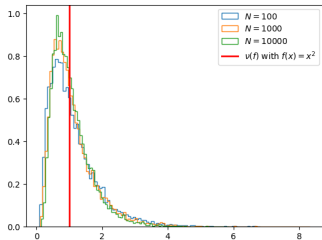
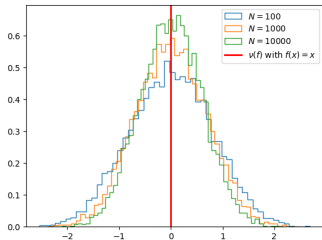
Figure – Values of  $\nu_n^\gamma(f)$  for different approach

## Almost Sure CLT

Let  $(U_n)$  be a sequence of i.i.d. square-integrable random variables, satisfying  $\mathbb{E}[U_1] = 0$  and  $\text{Var}(U_1) = 1$ , we have then

$$\frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} \delta_{(U_1 + \dots + U_k)/\sqrt{k}} \rightarrow \mathcal{N}(0, 1)$$

in distribution almost surely.



## Case of Non-Uniqueness

Inspired by ODE  $x' = -V'(x)$ , we define the following SDE

$$dX_t = -V'(X_t)dt + \sigma dW_t,$$

where  $\sigma$  can be changed.

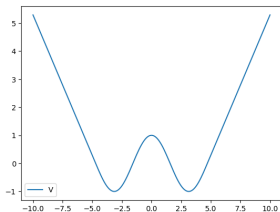


Figure – Image of  $V$

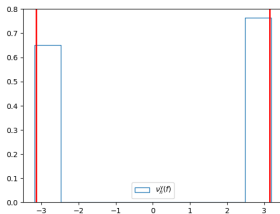


Figure – Case of  $\sigma = 0.5$

Its histogram can also be centered Gaussian when  $\sigma$  is big.



# Conclusion

Through our efforts in P1 , we ...

- comprehend a series of foundational works ;
- apply key theorems presented to derive new result ;
- verify the result through numerical simulations.

In the P2, we will ...

- consider the rate of convergence from theoretical perspectives ;
- explore more things.