

Stochastic Newton equation with absorbing area

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$X_t^\lambda, V_t^\lambda \in \mathbf{R}$: position and velocity of a particle at time t .

$$\begin{cases} dX_t^\lambda = V_t^\lambda dt \\ dV_t^\lambda = -bV_t^\lambda dt - \lambda \nabla g(X_t^\lambda) dt + \sigma(X_t^\lambda) dB_t, \\ (X_0^\lambda, V_0^\lambda) = (X_0, V_0), \end{cases} \quad (0.1)$$

with

$b > 0$: constant (coefficient of friction),

$\lambda \geq 1$: parameter (later, $\lambda \rightarrow \infty$),

$g \in C_0^\infty(\mathbf{R}; \mathbf{R})$: potential function,

$\sigma \in C^\infty(\mathbf{R}; \mathbf{R})$, positive uniformly

(X_0, V_0) : initial condition, $|X_0|$ big enough.

“Hamiltonian $H(x, v) = \frac{1}{2}|v|^2 + \lambda g(x)$ ” + friction + randomized

Problem: Behavior of the particle as $\lambda \rightarrow \infty$.

Motivation (reason for $\lambda \rightarrow \infty$): Mechanical model of Brownian motion.
Massive particle(s) in an ideal gas environment.

The Newtonian system with Hamiltonian: $H =$

$$\sum_{i=1}^N \frac{1}{2} M_i |V_i|^2 + \sum_{(x,v)} \frac{1}{2} m |v|^2 + \sum_{i=1}^N \sum_{(x,v)} U_i(X_i - x).$$

equivalently, (totally deterministic as long as the initial condition is given)

$$\left\{ \begin{array}{l} \frac{d}{dt} X_i(t, \omega) = V_i(t, \omega), \quad i = 1, \dots, N, \\ M_i \frac{d}{dt} V_i(t, \omega) = - \int_{\mathbf{R}^d \times \mathbf{R}^d} \mu_\omega(dx, dv) \\ \quad \cdot \nabla U_i(X_i(t, \omega) - x(t, x, v, \omega)), \\ (X_i(0, \omega), V_i(0, \omega)) = (X_{i,0}, V_{i,0}), \\ \frac{d}{dt} x(t, x, v, \omega) = v(t, x, v, \omega), \\ m \frac{d}{dt} v(t, x, v, \omega) = - \sum_{i=1}^N \nabla U_i(x(t, x, v, \omega) - X_i(t, \omega)), \\ (x(0, x, v, \omega), v(0, x, v, \omega)) = (x, v). \end{array} \right.$$

with initial condition (x, v) : Poisson point process, v and the density $\sim m^{-1/2}$,

Aim: Behavior of the massive particle(s) when $m \rightarrow 0$.

$P_m(d\omega)$: (probability on $Conf(\mathbb{R}^d \times \mathbb{R}^d)$),
the Poisson point process with intensity λ_m ,

$$\lambda_m(dx, dv) = m \frac{d-1}{2} \rho\left(\frac{m}{2}|v|^2 + \sum_{i=1}^N U_i(x - X_{i,0})\right) dx dv,$$

then (Kusuoka-L. (2010))

$$M_i V_i(t \wedge \sigma)$$

\approx initial + martingale + differentiable term

$$-m^{-1/2} \int_0^{t \wedge \sigma} \nabla_i \tilde{U}(\vec{X}(s)) ds,$$

with |the jump of the martingale term| $\leq C m^{1/2}$.

\tilde{U} : new potential for \vec{X} , (with no light particles as mediates)

Let $r_3 > 0$: $g(x) = 0 (\forall |x| \geq r_3)$.

→ In the domain $|x| > r_3$, λ has no effect

→ In the limit $\lambda \rightarrow \infty$, we get the same SDE:

$$\begin{cases} dX_t^\lambda = V_t^\lambda dt \\ dV_t^\lambda = -bV_t^\lambda dt + \sigma(X_t^\lambda)dB_t, \end{cases}$$

The behavior at $|X_t| = r_3$ (and $V_t \cdot X_t < 0$):

(i) if $\exists \varepsilon_0 > 0$ s.t., $g(x) > 0$ for $|x| \in (r_3 - \varepsilon_0, r_3)$

→ g gives a repulsion

→ after $\lambda \rightarrow \infty$, we get a reflective diffusion

→ Kusuoka (2004)

(ii) In this talk: $g(x) < 0$ right after entered $|x| \leq r_3$:

→ g gives an attraction

→ after $\lambda \rightarrow \infty$, $|V_t|$ becomes ∞ .

$\exists r_1 \in (0, r_3)$ s.t.,

$$g(x) < 0 (x \in (r_1, r_3)),$$

$$g(r_1) = 0,$$

$$g'(r_1) < 0.$$

Related result: L. (2013):

the same potential (i.e., attracting) but with relative efficacy
→ resulting in a stochastic process with two phases: a diffusion phase (for $|\mathbf{X}_t| \geq r_3$) and a uniform motion phase (for $|\mathbf{X}_t| \in (r_1, r_3)$).

In this talk: without relative efficacy

(In the limit $\lambda \rightarrow \infty$),

- V_t becomes ∞ right after it enters the domain $|X_t| \in (r_1, r_3)$,
- it could then never leave this domain.

Indeed, let $H_t^\lambda := \frac{1}{2}|V_t^\lambda|^2 + \lambda g(X_t^\lambda)$, then by Ito's formula,

$$dH_t^\lambda = -b|V_t^\lambda|^2 dt + V_t^\lambda \sigma(X_t^\lambda) dt + \frac{1}{2} \sigma^2(X_t^\lambda) dt,$$

hence H_t^λ becomes negative before its first hitting time to r_1 .

So it is meaningless to consider the limit behavior of X_t^λ itself.

For any $f \in C_b(\mathbb{R})$, let

$$Y_t^{f,\lambda} := \int_0^t f(X_s^\lambda) ds$$

(Since f is bounded, $\left\{ \text{the distribution of } \left\{ Y_t^{f,\lambda}; t \geq 0 \right\}; \lambda \geq 1 \right\}$ is tight).

Aim: The limit of the distribution of $\left\{ Y_t^{f,\lambda}; t \in [0, \infty) \right\}$ as

$\lambda \rightarrow \infty$.

W.l.o.g., let $X_0 = r_3$ and $V_0 < 0$.

A related known result (Sugiyama): $g(x) = ax^2$ ($a > 0$), i.e.,

$$\begin{cases} dX_t^\lambda = V_t^\lambda dt \\ dV_t^\lambda = -bV_t^\lambda dt - 2\lambda a X_t^\lambda dt + \sigma(X_t^\lambda) dB_t, \\ (X_0^\lambda, V_0^\lambda) = (x_0, v_0), \end{cases}$$

Then

$$X_t^\lambda = \dots, \text{ (3 lines)} \quad V_t^\lambda = \dots, \text{ (5 lines)}$$

so for any $T > 0$ and $f \in C_b^1(\mathbb{R})$,

$$\lim_{\lambda \rightarrow \infty} E \left[\left| \int_0^T f(X_t^\lambda) dt - 2 \int_{-x_0}^{x_0} \int_0^{\frac{\log x_0 - \log |x|}{b} \wedge \frac{T}{2}} f(x) h(u, x) du dx \right| \right] = 0,$$

here

$$h(u, x) = \frac{1}{\pi \sqrt{x_0^2 e^{-2bu} - x^2}}.$$

Consider the non-random case:

$$\begin{cases} dx_t^\lambda = v_t^\lambda dt, \\ dv_t^\lambda = -bv_t^\lambda dt - \lambda \nabla g(x_t^\lambda) dt, \\ (x_0^\lambda, v_0^\lambda) = (r_3, v), \end{cases}$$

with $v < 0$.

Same as the random case, the particle could never leave (r_1, r_3) .

Problem: The range of the particle.

Let

$$h_t^\lambda := \frac{1}{2} |v_t^\lambda|^2 + \lambda g(x_t^\lambda), \quad j_t^\lambda := \lambda^{-1} h_t^\lambda.$$

Then $j_t := \lim_{\lambda \rightarrow \infty} j_t^\lambda$ gives us the range of the particle (of the limit process) around time t .

Assume that g in (r_1, r_3) is single-well, i.e., $\exists r_2 \in (r_1, r_3)$, s.t.,

- $g(x) \Big|_{x \in (r_1, r_2)}$ is strictly decreasing (write the inverse: $g^{-1,1}$),
- $g(x) \Big|_{x \in (r_2, r_3)}$ is strictly increasing (write the inverse: $g^{-1,2}$)

+ some technical condition.

Then

- $\dot{j}_t < 0$ for any $t > 0$,
- $dj_t = -2b \left(j_t - A^g g(j_t) \right) dt$, $j_0 = 0$, with

$$A^g f(x) := \frac{S_f(j)}{S_1(j)}, \quad x \in (-\|g\|_\infty, 0),$$

$$S_f(j) := S_f^g(j) := \sqrt{2} \int_{g^{-1,1}(j)}^{g^{-1,2}(j)} \frac{f(y)}{\sqrt{j - g(y)}} dy.$$

Idea: balance of “time for each trip” and “decay of energy during each trip”

Come back to the random case.

Let

$$J_t^\lambda := \lambda^{-1} H_t^\lambda = \frac{1}{2} \lambda^{-1} |V_t^\lambda|^2 + g(X_t^\lambda).$$

($J_t := \lim_{\lambda \rightarrow \infty} J_t^\lambda$ gives us the range of the particle (of the limit process) around time t).

Theorem. Under the above assumptions, for any $f \in C_b(\mathbb{R})$, we have that when $\lambda \rightarrow \infty$, $\left\{ (J_t^\lambda, Y_t^{f,\lambda}); t \in [0, \infty) \right\}$ converge to $\left\{ (j_t, \int_0^t A^g f(j_s) ds); t \in [0, \infty) \right\}$ weakly in $(W, dist)$.

Here $W = C([0, \infty); \mathbb{R}^2)$ and for $\forall w_1, w_2 \in W$,
 $dist(w_1, w_2) = \sum_{n=1}^{\infty} 2^{-n} \left(1 \wedge \left[\max_{t \in [0, n]} |w_1(t) - w_2(t)| \right] \right)$.

Remark: Non-random limit ONLY for $d = 1$

For $d \geq 2$: same limit for $|X_t|$, but random limit for the direction

Open problem. In our (motivating) mechanical model of Brownian Motion: **Is b negative-definite?**

The corresponding “limit” generator

$$\begin{aligned}
 L_1 = & \frac{1}{2} \sum_{k_1, k_2=1}^N \sum_{l_1, l_2=1}^d a_{k_1 l_1, k_2 l_2}(\vec{X}) \frac{\partial^2}{\partial V_{k_1}^{l_1} \partial V_{k_2}^{l_2}} + \\
 & \sum_{k_1, k_2=1}^N \sum_{l_1, l_2=1}^d b_{k_1 l_1, k_2 l_2}(\vec{X}) V_{k_2}^{l_2} \frac{\partial}{\partial V_{k_1}^{l_1}} + \\
 & \sum_{k=1}^N \sum_{i=1}^d V_k^i \frac{\partial}{\partial X_k^i},
 \end{aligned}$$

with

$$\begin{aligned}
 & \int_E \left(\int_{-\infty}^{\infty} \nabla^2 U_i(\psi^0(t, \mathbf{x}, \mathbf{v}, \vec{X}) - X_i) z(t, \mathbf{x}, \mathbf{v}, \vec{X}, \vec{V}, -t) dt \right) \\
 & \times \rho\left(\frac{1}{2} |\mathbf{v}|^2\right) \nu(d\mathbf{x}, d\mathbf{v}) = \sum_{\ell=1}^d \sum_{j=1}^N b_{i.,j\ell}(\vec{X}) V_j^\ell.
 \end{aligned}$$

$$E = \{(\mathbf{x}, \mathbf{v}) \in \mathbb{R}^d \times (\mathbb{R}^d \setminus \{0\}); \mathbf{x} \cdot \mathbf{v} = 0\},$$

$z(t; x, v, \vec{X}, \vec{V}, a) \in \mathbb{R}^d$ denotes the solution of the following standard differential equation.

$$\begin{cases} \frac{d^2}{dt^2} Z(t) = - \sum_{i=1}^N \nabla^2 U_i(\psi^0(t, x, v, \vec{X}) - X_i)(Z(t) - (t + a)V_i), \\ \lim_{t \rightarrow -\infty} Z(t) = \lim_{t \rightarrow -\infty} \frac{d}{dt} Z(t) = 0. \end{cases}$$

(we have that $z(t; x, v, \vec{X}, \vec{V}, a)$ is a linear function of \vec{V}).

$$\psi(t, x, v; \vec{X}) := \lim_{s \rightarrow \infty} \varphi(t + s, x - sv, v; \vec{X}),$$

$$\begin{cases} \frac{d}{dt} \varphi^0(t, x, v; \vec{X}) = \varphi^1(t, x, v; \vec{X}) \\ \frac{d}{dt} \varphi^1(t, x, v; \vec{X}) = - \sum_{i=1}^N \nabla U_i(\varphi^0(t, x, v; \vec{X}) - X_i) \\ (\varphi^0(0, x, v; \vec{X}), \varphi^1(0, x, v; \vec{X})) = (x, v). \end{cases}$$

An important estimate for the proof: $\exists C > 0$, s.t.,

$$E \left[\sup_{t \in [0, T]} |V_t^\lambda|^4 \right]^{1/4} \leq C \lambda^{\frac{1}{2}}, \quad \lambda \geq 1.$$

Proof. By Ito's formula,

$$dH_t^\lambda = -b|V_t^\lambda|^2 dt + V_t^\lambda \sigma(X_t^\lambda) dB_t + \frac{1}{2} \sigma(X_t^\lambda)^2 dt, \text{ so}$$

$$\begin{aligned} |V_t^\lambda|^2 &= 2H_t^\lambda - 2\lambda g(X_t^\lambda) \\ &\leq 2H_0 + 2 \int_0^t V_s^\lambda \sigma(X_s^\lambda) dB_s + T \|\sigma\|_\infty^2 + 2\lambda \|g\|_\infty. \end{aligned}$$

Therefore,

$$\begin{aligned} &E \left[\sup_{t \in [0, T]} |V_t^\lambda|^4 \right] \\ &\leq 2(C_1 + C_2 \lambda)^2 + 2E \left[\sup_{t \in [0, T]} \left(2 \int_0^t V_s^\lambda \sigma(X_s^\lambda) dB_s \right)^2 \right] \\ &\leq 2(C_1 + C_2 \lambda)^2 + 32 \|\sigma\|_\infty^2 T E \left[\sup_{t \in [0, T]} |V_t^\lambda|^4 \right]^{1/2}. \end{aligned}$$

In general, for any $c_1, c_2 \in \mathbb{R}^+$, we have that

$$x^2 \leq c_1 + c_2 x \Rightarrow x \leq \frac{c_2 + \sqrt{c_2^2 + 4c_1}}{2} \leq c_2 + \sqrt{c_1}.$$

Therefore, we get that

$$\begin{aligned} & E \left[\sup_{t \in [0, T]} |V_t^\lambda|^4 \right]^{1/2} \\ & \leq 32 \|\sigma\|_\infty^2 T + \sqrt{2(C_1 + C_2 \lambda)^2} \\ & = 32 \|\sigma\|_\infty^2 T + \sqrt{2} C'_1 + C'_2 \lambda \end{aligned}$$

for any $\lambda \geq 1$.

Q.E.D.

Main idea for the dealing with randomness:

- For any $\{M(t)\}_{t \geq 0}$: continuous martingale, $\exists \{W(t)\}_{t \geq 0}$: BM, s.t., $M(t) = W(\langle M, M \rangle_t)$.
- For any standard BM $\{\bar{B}_t\}_{t \geq 0}$, we have the following:
 - (i) $\lim_{a \rightarrow \infty} P\left(\left\{\inf_{u \geq 0} (\epsilon u + \bar{B}_t) < -a\right\}\right) = 0$ for any $\epsilon > 0$,
 - (ii) $\lim_{a \rightarrow \infty} P\left(\left\{\bar{B}_s - \epsilon s \geq 0 \text{ for some } s \geq a\right\}\right) = 0$ for any $\epsilon > 0$,
 - (iii) $P\left(\limsup_{\epsilon \rightarrow 0} \left\{\sup_{0 \leq s \leq t+s \leq T, t \leq \epsilon} \frac{|\bar{B}_{t+s} - \bar{B}_s|}{\sqrt{2\epsilon \log c_1 \epsilon}}\right\} = 1\right) = 1$.

Thank you!